BBM402-Lecture 12: NP-completeness

Lecturer: Lale Özkahya

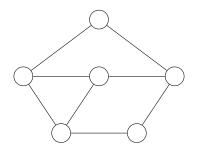
Resources for the presentation: http://ocw.mit.edu/courses/electrical-engineering-and-computer-science/6-045j-automata-computability-and-complexity-spring-2011/Syllabus/https://courses.engr.illinois.edu/cs498374/lectures.html

Given a graph G, a set of vertices V' is:

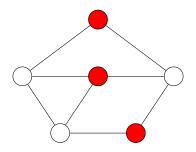
1 independent set: no two vertices of **V**' connected by an edge.

- lacktriangledown independent set: no two vertices of V' connected by an edge.
- $oldsymbol{\circ}$ clique: every pair of vertices in $oldsymbol{V}'$ is connected by an edge of $oldsymbol{G}$.

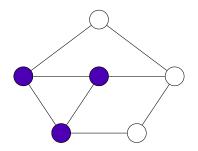
- lacktriangledown independent set: no two vertices of V' connected by an edge.
- $oldsymbol{\circ}$ clique: every pair of vertices in $oldsymbol{\mathsf{V}}'$ is connected by an edge of $oldsymbol{\mathsf{G}}$.



- lacktriangledown independent set: no two vertices of V' connected by an edge.
- $oldsymbol{\circ}$ clique: every pair of vertices in $oldsymbol{\mathsf{V}}'$ is connected by an edge of $oldsymbol{\mathsf{G}}$.



- lacktriangledown independent set: no two vertices of V' connected by an edge.
- $oldsymbol{\circ}$ clique: every pair of vertices in $oldsymbol{\mathsf{V}}'$ is connected by an edge of $oldsymbol{\mathsf{G}}$.



The Independent Set and Clique Problems

Problem: Independent Set

Instance: A graph G and an integer **k**.

Question: Does G has an independent set of size $\geq k$?

24

Spring 2015

The Independent Set and Clique Problems

Problem: Independent Set

Instance: A graph G and an integer **k**.

Question: Does G has an independent set of size $\geq k$?

Problem: Clique

Instance: A graph G and an integer **k**.

Question: Does G has a clique of size $\geq k$?

24

Recall

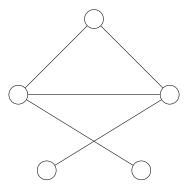
For decision problems X, Y, a reduction from X to Y is:

- An algorithm . . .
- that takes I_X, an instance of X as input . . .
- $oldsymbol{0}$ and returns $oldsymbol{I_Y}$, an instance of $oldsymbol{Y}$ as output . . .
- \odot such that the solution (YES/NO) to I_Y is the same as the solution to I_X .

An instance of **Independent Set** is a graph **G** and an integer **k**.

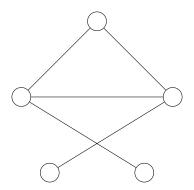
26

An instance of **Independent Set** is a graph **G** and an integer **k**.



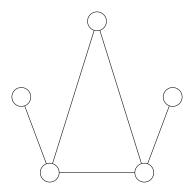
An instance of **Independent Set** is a graph **G** and an integer **k**.

Reduction given $< \mathbf{G}, \mathbf{k} >$ outputs $< \overline{\mathbf{G}}, \mathbf{k} >$ where $\overline{\mathbf{G}}$ is the complement of \mathbf{G} . $\overline{\mathbf{G}}$ has an edge (\mathbf{u}, \mathbf{v}) if and only if (\mathbf{u}, \mathbf{v}) is not an edge of \mathbf{G} .



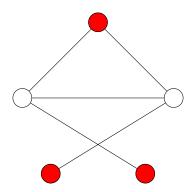
An instance of **Independent Set** is a graph **G** and an integer **k**.

Reduction given $< \underline{G}, k >$ outputs $< \overline{G}, k >$ where \overline{G} is the complement of G. \overline{G} has an edge (u, v) if and only if (u, v) is not an edge of G.



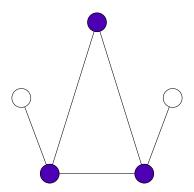
An instance of **Independent Set** is a graph **G** and an integer **k**.

Reduction given $< \underline{G}, k >$ outputs $< \overline{G}, k >$ where \overline{G} is the complement of G. \overline{G} has an edge (u, v) if and only if (u, v) is not an edge of G.



An instance of **Independent Set** is a graph **G** and an integer **k**.

Reduction given $< \underline{G}, k >$ outputs $< \overline{G}, k >$ where \overline{G} is the complement of G. \overline{G} has an edge (u, v) if and only if (u, v) is not an edge of G.



Correctness of reduction

Lemma

G has an independent set of size **k** if and only if $\overline{\mathbf{G}}$ has a clique of size **k**.

Proof.

Need to prove two facts:

G has independent set of size at least **k** implies that $\overline{\mathbf{G}}$ has a clique of size at least **k**.

 $\overline{\mathbf{G}}$ has a clique of size at least \mathbf{k} implies that \mathbf{G} has an independent set of size at least \mathbf{k} .

Easy to see both from the fact that $\mathbf{S} \subseteq \mathbf{V}$ is an independent set in

G if and only if **S** is a clique in $\overline{\mathbf{G}}$.

• Independent Set \leq Clique.

28

- Independent Set \leq Clique.
 - What does this mean?
- If have an algorithm for Clique, then we have an algorithm for Independent Set.

- Independent Set ≤ Clique. What does this mean?
- If have an algorithm for Clique, then we have an algorithm for Independent Set.
- Clique is at least as hard as Independent Set.

- Independent Set ≤ Clique. What does this mean?
- If have an algorithm for Clique, then we have an algorithm for Independent Set.
- Clique is at least as hard as Independent Set.
- Also... Independent Set is at least as hard as Clique.

Assume you can solve the Clique problem in T(n) time. Then you can solve the Independent Set problem in

- (A) O(T(n)) time.
- (B) $O(n \log n + T(n))$ time.
- (C) $O(n^2T(n^2))$ time.
- (D) $O(n^4T(n^4))$ time.
- (E) $O(n^2 + T(n^2))$ time.
- (F) Does not matter all these are polynomial if T(n) is polynomial, which is good enough for our purposes.

Given a graph G = (V, E), a set of vertices S is:

39

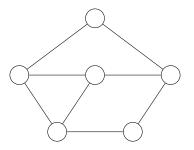
Given a graph G = (V, E), a set of vertices S is:

1 A **vertex cover** if every $e \in E$ has at least one endpoint in **S**.

39

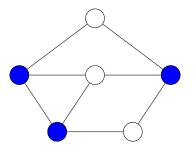
Given a graph G = (V, E), a set of vertices S is:

1 A **vertex cover** if every $e \in E$ has at least one endpoint in **S**.



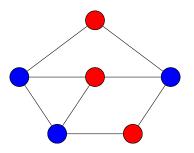
Given a graph G = (V, E), a set of vertices S is:

1 A **vertex cover** if every $e \in E$ has at least one endpoint in **S**.



Given a graph G = (V, E), a set of vertices S is:

1 A **vertex cover** if every $e \in E$ has at least one endpoint in **S**.



The Vertex Cover Problem

Problem (Vertex Cover)

Input: A graph G and integer **k**.

Goal: Is there a vertex cover of size $\leq k$ in G?

40

The Vertex Cover Problem

Problem (Vertex Cover)

Input: A graph G and integer **k**.

Goal: Is there a vertex cover of size < k in G?

Can we relate **Independent Set** and **Vertex Cover**?

Relationship between...

Vertex Cover and Independent Set

Proposition

Let G = (V, E) be a graph. S is an independent set if and only if $V \setminus S$ is a vertex cover.

Proof.

- (\Rightarrow) Let **S** be an independent set
 - Consider any edge $uv \in E$.
 - 2 Since S is an independent set, either $\mathbf{u} \not\in \mathbf{S}$ or $\mathbf{v} \not\in \mathbf{S}$.
 - **3** Thus, either $\mathbf{u} \in \mathbf{V} \setminus \mathbf{S}$ or $\mathbf{v} \in \mathbf{V} \setminus \mathbf{S}$.
 - **◊ V \ S** is a vertex cover.
- (⇐) Let **V** \ **S** be some vertex cover:
 - Consider $\mathbf{u}, \mathbf{v} \in \mathbf{S}$
 - **2** uv is not an edge of G, as otherwise $V \setminus S$ does not cover uv.
 - \longrightarrow S is thus an independent set.

Independent Set \leq_{P} Vertex Cover

G: graph with n vertices, and an integer k be an instance of the Independent Set problem.

Independent Set \leq_P Vertex Cover

- G: graph with n vertices, and an integer k be an instance of the Independent Set problem.
- ② **G** has an independent set of size \geq **k** iff **G** has a vertex cover of size \leq **n k**

Independent Set \leq_{P} Vertex Cover

- G: graph with n vertices, and an integer k be an instance of the Independent Set problem.
- ② **G** has an independent set of size \geq **k** iff **G** has a vertex cover of size \leq **n k**
- **3** (G, k) is an instance of **Independent Set**, and (G, n k) is an instance of **Vertex Cover** with the same answer.

Independent Set \leq_P Vertex Cover

- G: graph with n vertices, and an integer k be an instance of the Independent Set problem.
- ② **G** has an independent set of size \geq **k** iff **G** has a vertex cover of size \leq **n k**
- **3** (G, k) is an instance of Independent Set, and (G, n k) is an instance of Vertex Cover with the same answer.
- Therefore, Independent Set ≤_P Vertex Cover. Also Vertex Cover ≤_P Independent Set.

Proving Correctness of Reductions

To prove that $X \leq_P Y$ you need to give an algorithm A that:

- **1** Transforms an instance I_X of X into an instance I_Y of Y.
- ② Satisfies the property that answer to I_X is YES iff I_Y is YES.
 - typical easy direction to prove: answer to I_Y is YES if answer to I_X is YES
 - typical difficult direction to prove: answer to I_X is YES if answer to I_Y is YES (equivalently answer to I_X is NO if answer to I_Y is NO).
- Runs in polynomial time.

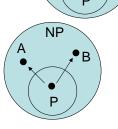
NP-Completeness

 ≤_p allows us to relate problems in NP, saying which allow us to solve which others efficiently.

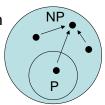
 Even though we don't know whether all of these problems are in P, we can use ≤_p to impose some structure on the class NP:

• $A \rightarrow B$ here means $A \leq_p B$.

 Sets in NP – P might not be totally ordered by ≤_p: we might have A, B with neither A ≤_p B nor B ≤_p A:



- Some languages in NP are hardest, in the sense that every language in NP is ≤_p-reducible to them.
- Call these NP-complete.
 - Definition: Language B is NP-complete if both of the following hold:
 - (a) $B \in NP$, and
 - (b) For any language $A \in NP$, $A \leq_p B$.



- Sometimes, we consider languages that aren't, or might not be, in NP, but to which all NP languages are reducible.
- Call these NP-hard.
- Definition: Language B is NP-hard if, for any language A
 ∈ NP, A ≤_n B.

- Today, and next time, we'll:
 - Give examples of interesting problems that are NPcomplete, and
 - Develop methods for showing NP-completeness.
- Theorem: ∃B, B is NP-complete.
 - There is at least one NP-complete problem.
 - We'll show this later.
- Theorem: If A, B, are NP-complete, then $A \leq_p B$.
 - Two NP-complete problems are essentially equivalent (up to \leq_p).
- Proof: A ∈ NP, B is NP-hard, so A ≤_p B by definition.

- Theorem: If some NP-complete language is in P, then P = NP.
 - That is, if a polynomial-time algorithm exists for any NPcomplete problem, then the entire class NP collapses into P.
 - Polynomial algorithms immediately arise for all problems in NP.

Proof:

- Suppose B is NP-complete and B ∈ P.
- Let A be any language in NP; show $A \in P$.
- We know $A \leq_n B$ since B is NP-complete.
- Then $A \in P$, since $B \in P$ and "easiness propagates downward".
- Since every A in NP is also in P, NP \subseteq P.
- Since $P \subseteq NP$, it follows that P = NP.

- Theorem: The following are equivalent.
 - 1. P = NP.
 - 2. Every NP-complete language is in P.
 - 3. Some NP-complete language is in P.
- Proof:
 - $1 \Rightarrow 2$:
 - Assume P = NP, and suppose that B is NP-complete.
 - Then $B \in NP$, so $B \in P$, as needed.
 - $2 \Rightarrow 3$:
 - Immediate because there is at least NP-complete language.
 - 3 ⇒ 1:
 - · By the previous theorem.

Beliefs about P vs. NP

- Most theoretical computer scientists believe P ≠ NP.
- Why?
- Many interesting NP-complete problems have been discovered over the years, and many smart people have tried to find fast algorithms; no one has succeeded.
- The problems have arisen in many different settings, including logic, graph theory, number theory, operations research, games and puzzles.
- Entire book devoted to them [Garey, Johnson].
- All these problems are essentially the same since all NPcomplete problems are polynomial-reducible to each other.
- So essentially the same problem has been studied in many different contexts, by different groups of people, with different backgrounds, using different methods.

Beliefs about P vs. NP

- Most theoretical computer scientists believe P ≠ NP.
- Because many smart people have tried to find fast algorithms and no one has succeeded.
- That doesn't mean P ≠ NP; this is just some kind of empirical evidence.
- The essence of why NP-complete problems seem hard:
 - They have NP structure:

```
x \in L \text{ iff } (\exists \ c, \ |c| \leq p(|x|) \ ) \ [ \ V(\ x, \ c \ ) \text{ accepts } ], where V is poly-time.
```

- Guess and verify.
- Seems to involve exploring a tree of possible choices, exponential blowup.
- However, no one has yet succeeded in proving that they actually are hard!
 - We don't have sharp enough methods.
 - So in the meantime, we just show problems are NP-complete.

- SAT = { < φ > | φ is a satisfiable Boolean formula }
- Definition: (Boolean formula):
 - Variables: x, x₁, x₂, ..., y,..., z,...
 - Can take on values 1 (true) or 0 (false).
 - **Literal**: A variable or its negated version: $x, \neg x, \neg x_1, ...$
 - Operations: ∧ ∨ ¬
 - Boolean formula: Constructed from literals using operations, e.g.:

$$\phi = X \wedge ((y \wedge z) \vee (\neg y \wedge \neg z)) \wedge \neg (x \wedge z)$$

- Definition: (Satisfiability):
 - A Boolean formula is satisfiable iff there is an assignment of 0s and 1s to the variables that makes the entire formula evaluate to 1 (true).

- SAT = { < ∅ > | ∅ is a satisfiable Boolean formula }
- Boolean formula: Constructed from literals using operations, e.g.:

```
\phi = X \wedge ((y \wedge z) \vee (\neg y \wedge \neg z)) \wedge \neg (X \wedge Z)
```

- A Boolean formula is satisfiable iff there is an assignment of 0s and 1s to the variables that makes the entire formula evaluate to 1 (true).
- Example: φ above
 - Satisfiable, using the assignment x = 1, y = 0, z = 0.
 - So ϕ ∈ SAT.
- Example: x ∧ ((y ∧ z) ∨ (¬y ∧ z)) ∧ ¬(x ∧ z)
 - Not in SAT.
 - x must be set to 1, so z must = 0.

- SAT = $\{ \langle \phi \rangle | \phi \text{ is a satisfiable Boolean formula } \}$
- Theorem: SAT is NP-complete.
- Lemma 1: SAT ∈ NP.
- Lemma 2: SAT is NP-hard.
- Proof of Lemma 1:
 - Recall: L ∈ NP if and only if (\exists V, poly-time verifier) (\exists p, poly) x ∈ L iff (\exists c, |c| ≤ p(|x|)) [V(x, c) accepts]
 - So, to show SAT \in NP, it's enough to show (\exists V) (\exists p)
 - $\phi \in SAT \text{ iff } (\exists \ c, \ |c| \le p(|x|) \) \ [\ V(\ \phi, \ c \) \ accepts \]$
 - We know: $\phi \in SAT$ iff there is an assignment to the variables such that ϕ with this assignment evaluates to 1.
 - So, let certificate c be the assignment.
 - Let verifier V take a formula φ and an assignment c and accept exactly if φ with c evaluates to true.

- Lemma 2: SAT is NP-hard.
- Proof of Lemma 2:
 - Need to show that, for any $A \in NP$, $A \leq_n SAT$.
 - $Fix A \in NP$.
 - Construct a poly-time f such that

 $w \in A$ if and only if $f(w) \in SAT$.

A formula, write it as ϕ_w .

- By definition, since $A \in NP$, there is a nondeterministic TM M that decides A in polynomial time.
- Fix polynomial p such that M on input w always halts, on all branches, in time $\leq p(|w|)$; assume $p(|w|) \geq |w|$.
- w ∈ A if and only if there is an accepting computation history (CH) of M on w.

- Lemma 2: SAT is NP-hard.
- Proof, cont'd:
 - − Need w ∈ A if and only if f(w) (= ϕ_w) ∈ SAT.
 - $w \in A$ if and only if there is an accepting CH of M on w.
 - So we must construct formula ϕ_w to be satisfiable iff there is an accepting CH of M on w.
 - Recall definitions of computation history and accepting computation history from Post Correspondence Problem:
 # C₀ # C₁ # C₂ ...
 - Configurations include tape contents, state, head position.
 - We construct ϕ_w to describe an accepting CH.
 - Let M = (Q, Σ , Γ , δ , q_0 , q_{acc} , q_{rei}) as usual.
 - Instead of lining up configs in a row as before, arrange in (p(|w|) + 1) row × (p(|w|) + 3) column matrix:

Proof that SAT is NP-hard

- ϕ_w will be satisfiable iff there is an accepting CH of M on w.
- Let M = (Q, Σ , Γ , δ , q_0 , q_{acc} , q_{rei}).
- Arrange configs in $(p(|w|) + 1) \times (p(|w|) + 3)$ matrix:

- · Successive configs, ending with accepting config.
- Assume WLOG that each computation takes exactly p(|w|) steps, so we use p(|w|) + 1 rows.
- p(|w|) + 3 columns: p(|w|) for the interesting portion of the tape, one for head and state, two for endmarkers.

Proof that SAT is NP-hard

- ϕ_w is satisfiable iff there is an accepting CH of M on w.
- Entries in the matrix are represented by Boolean variables:
 - − Define $C = Q \cup \Gamma \cup \{\#\}$, alphabet of possible matrix entries.
 - Variable x_{i,i,c} represents "the entry in position (i, j) is c".
- Define ϕ_w as a formula over these $x_{i,j,c}$ variables, satisfiable if and only if there is an accepting computation history for w (in matrix form).
- Moreover, an assignment of values to the $x_{i,j,c}$ variables that satisfies ϕ_w will correspond to an encoding of an accepting computation.
- Specifically, $\phi_w = \phi_{cell} \wedge \phi_{start} \wedge \phi_{accept} \wedge \phi_{move}$, where:
 - ϕ_{cell} : There is exactly one value in each matrix location.
 - ϕ_{start} : The first row represents the starting configuration.
 - ϕ_{accept} : The last row is an accepting configuration.
 - ϕ_{move} : Successive rows represent allowable moves of M.

ϕ_{cel}

- For each position (i,j), write the conjunction of two formulas:
 - $\bigvee_{c \in C} x_{i,j,c}$. Some value appears in position (i,j).
 - $\bigwedge_{c, d \in C, c \neq d} (\neg x_{i,j,c} \lor \neg x_{i,j,d})$: Position (i,j) doesn't contain two values.
- φ_{cell}: Conjoin formulas for all positions (i,j).
- Easy to construct the entire formula φ_{cell} given w input.
- Construct it in polynomial time.
- Sanity check: Length of formula is polynomial in |w|:
 - $O((p(|w|)^2))$ subformulas, one for each (i,j).
 - Length of each subformula depends on C, O($|C|^2$).

ϕ_{start}

• The right symbols appear in the first row:

```
# q<sub>0</sub> w<sub>1</sub> w<sub>2</sub> w<sub>3</sub> ... w<sub>n</sub> -- -- ... -- #
```

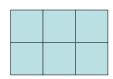
$$\phi_{accept}$$

• For each j, $2 \le j \le p(|w|) + 2$, write the formula:

$$\mathbf{X}_{p(|w|)+1,j,qacc}$$

- q_{acc} appears in position j of the last row.
- φ_{accept}: Take disjunction (or) of all formulas for all j.
- That is, q_{acc} appears in some position of the last row.

- As for PCP, correct moves depend on correct changes to local portions of configurations.
- It's enough to consider 2 × 3 rectangles:
- If every 2 x 3 rectangle is "good", i.e., consistent with the transitions, then the entire matrix represents an accepting CH.
- For each position (i,j), 1 ≤ i ≤ p(|w|), 1 ≤ j ≤ p(|w|)+1, write a formula saying that the rectangle with upper left at (i,j) is "good".
- Then conjoin all of these, O(p(|w|)²) clauses.
- Good tiles for (i,j), for a, b, c in Γ:



а	b	С
а	b	С

#	а	b
#	а	b

а	b	#
а	b	#

- Other good tiles are defined in terms of the nondeterministic transition function δ.
- E.g., if δ(q₁, a) includes tuple (q₂, b, L), then the following are good:
 - Represents the move directly; for any c:
 - Head moves left out of the rectangle; for any c, d:
 - Head is just to the left of the rectangle; for any c, d:
 - Head at right; for any c, d, e:
 - And more, for #, etc.
- Analogously if δ(q₁, a) includes (q₂, b, R).
- Since M is nondeterministic, δ(q₁, a) may contain several moves, so include all the tiles.

С	q ₁	а
q_2	С	b

q_1	а	С
d	b	C

а	С	d
b	C	d

d	С	q ₁
d	q_2	С

е	d	С
е	а	q_2

- The good tiles give partial constraints on the computation.
- When taken together, they give enough constraints so that only a correct CH can satisfy them all.
- The part (conjunct) of ϕ_{move} for (i,j) should say that the rectangle with upper left at (i,j) is good:
- It is simply the disjunction (or), over all allowable tiles, of the subformula:

$$X_{i,j,a1} \wedge X_{i,j+1,a2} \wedge X_{i,j+2,a3} \wedge X_{i+1,j,b1} \wedge X_{i+1,j+1,b2} \wedge X_{i+1,j+2,b3}$$

• Thus, ϕ_{move} is the conjunction over all (i,j), of the disjunction over all good tiles, of the formula just above.

- φ_{move} is the conjunction over all (i,j), of the disjunction over all good tiles, of the given sixterm conjunctive formula.
- Q: How big is the formula ϕ_{move} ?
- O(p(|w|)²) clauses, one for each (i,j) pair.
- Each clause is only constant length, O(1).
 - Because machine M yields only a constant number of good tiles.
 - And there are only 6 terms for each tile.
- Thus, length of ϕ_{move} is polynomial in |w|.
- $\phi_w = \phi_{cell} \wedge \phi_{start} \wedge \phi_{accept} \wedge \phi_{move}$, length also poly in |w|.

- $\phi_w = \phi_{cell} \wedge \phi_{start} \wedge \phi_{accept} \wedge \phi_{move}$, length poly in |w|.
- More importantly, can produce φ_w from w in time that is polynomial in |w|.
- $w \in A$ if and only if M has an accepting CH for w if and only if ϕ_w is satisfiable.
- Thus, $A \leq_{n} SAT$.
- Since A was any language in NP, this proves that SAT is NP-hard.
- Since SAT is in NP and is NP-hard, SAT is NP-complete.