

NP-completeness

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UNSW

Outline

- 1 Overview
- 2 Turing Machines, P, and NP
- 3 Reductions and NP-completeness
- 4 NP-complete problems
- 5 Further Reading

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Polynomial time

Polynomial-time algorithm

Polynomial-time algorithm:

There exists a constant $c \in \mathbb{N}$ such that the algorithm has (worst-case) running-time $O(n^c)$, where n is the size of the input.

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Example

Polynomial: n ; $n^2 \log_2 n$; n^3 ; n^{20}

Super-polynomial: $n^{\log_2 n}$; $2^{\sqrt{n}}$; 1.001^n ; 2^n ; $n!$

Tractable problems

Central Question

Which computational problems have polynomial-time algorithms?

Million-dollar question

Intriguing class of problems: NP-complete problems.

NP-complete problems

It is unknown whether NP-complete problems have polynomial-time algorithms.

- A polynomial-time algorithm for one NP-complete problem would imply polynomial-time algorithms for all problems in NP.

Gerhard Woeginger's P vs NP page:

<http://www.win.tue.nl/~gwoegi/P-versus-NP.htm>

Polynomial vs. NP-complete

Polynomial

- SHORTEST PATH: Given a graph G , two vertices a and b of G , and an integer k , does G have a simple a - b -path of length at most k ?
- EULER TOUR: Given a graph G , does G have a cycle that traverses each edge of G exactly once?
- 2-CNF SAT: Given a propositional formula F in 2-CNF, is F satisfiable?

A k -CNF formula is a conjunction (AND) of clauses, and each clause is a disjunction (OR) of at most k literals, which are negated or unnegated Boolean variables.

NP-complete

- LONGEST PATH: Given a graph G and an integer k , does G have a simple path of length at least k ?
- HAMILTONIAN CYCLE: Given a graph G , does G have a simple cycle that visits each vertex of G ?
- 3-CNF SAT: Given a propositional formula F in 3-CNF, is F satisfiable?

Example:

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

What's next?

- Formally define P , NP , and NP -complete (NPC)
- (New) skill: show that a problem is NP -complete

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Decision problems and Encodings

<Name of Decision Problem>

Input: <What constitutes an instance>

Question: <Yes/No question>

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We want to know which decision problems can be solved in polynomial time – polynomial in the **size of the input** n .

- Assume a “reasonable” encoding of the input
- Many encodings are polynomial-time equivalent; i.e., one encoding can be computed from another in polynomial time.
- Important exception: unary versus binary encoding of integers.
 - An integer x takes $\lceil \log_2 x \rceil$ bits in binary and $x = 2^{\log_2 x}$ bits in unary.

Formal-language framework

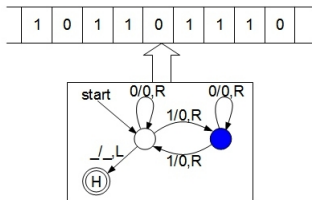
We can view decision problems as languages.

- Alphabet Σ : finite set of symbols. W.l.o.g., $\Sigma = \{0, 1\}$
- Language L over Σ : set of strings made with symbols from Σ : $L \subseteq \Sigma^*$
- Fix an encoding of instances of a decision problem Π into Σ
- Define the language $L_\Pi \subseteq \Sigma^*$ such that

$$x \in L_\Pi \Leftrightarrow x \text{ is a Yes-instance for } \Pi$$

Non-deterministic Turing Machine (NTM)

- **input word** $x \in \Sigma^*$ placed on an **infinite tape** (memory)
- read-write head initially placed on the first symbol of x
- computation step: if the machine is in state s and reads a , it can move into state s' , writing b , and moving the head into direction $D \in \{L, R\}$ if $((s, a), (s', b, D)) \in \delta$.



- Q : finite, non-empty set of states
- Γ : finite, non-empty set of tape symbols
- $_ \in \Gamma$: blank symbol (the only symbol allowed to occur on the tape infinitely often)
- $\Sigma \subseteq \Gamma \setminus \{b\}$: set of input symbols
- $q_0 \in Q$: start state
- $A \subseteq Q$: set of accepting (final) states
- $\delta \subseteq (Q \setminus A \times \Gamma) \times (Q \times \Gamma \times \{L, R\})$: transition relation, where L stands for a move to the left and R for a move to the right.

Definition 1

A NTM **accepts** a word $x \in \Sigma^*$ if there exists a sequence of computation steps starting in the start state and ending in an accept state.

Definition 2

The language **accepted** by an NTM is the set of words it accepts.

Acceptance in polynomial time

Definition 3

A language L is **accepted in polynomial time** by an NTM M if

- L is accepted by M , and
- there is a constant k such that for any word $x \in L$, the NTM M accepts x in $O(|x|^k)$ computation steps.

Deterministic Turing Machine

Definition 4

A **Deterministic Turing Machine (DTM)** is a Non-deterministic Turing Machine where the transition relation contains at most one tuple $((s, a), (\cdot, \cdot, \cdot))$ for each $s \in Q \setminus A$ and $a \in \Gamma$.

The transition relation δ can be viewed as a function

$$\delta : Q \setminus A \times \Gamma \rightarrow Q \times \Gamma \times \{L, R\}.$$

\Rightarrow For a given input word $x \in \Sigma^*$, there is exactly one sequence of computation steps starting in the start state.

Many computational models are polynomial-time equivalent to DTMs:

- Random Access Machine (RAM, used for algorithms in the textbook)
- variants of Turing machines (multiple tapes, infinite only in one direction, ...)
- ...

Definition 5 (P)

$P = \{L \subseteq \Sigma^* : \text{there is a DTM accepting } L \text{ in polynomial time}\}$

Definition 6 (NP)

$NP = \{L \subseteq \Sigma^* : \text{there is a NTM accepting } L \text{ in polynomial time}\}$

Definition 7 (coNP)

$coNP = \{L \subseteq \Sigma^* : \Sigma^* \setminus L \in NP\}$

Theorem 8

If $L \in P$, then there is a polynomial-time DTM that halts in an accepting state on every word in L and it halts in a non-accepting state on every word not in L .

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Proof sketch.

Suppose $L \in \mathbf{P}$. By the definition of \mathbf{P} , there is a DTM M that accepts L in polynomial time.

Idea: design a DTM M' that simulates M for $c \cdot n^k$ steps, where $c \cdot n^k$ is the running time of M and transitions to a non-accepting state if M does not halt in an accepting state.

(Note that this proof is nonconstructive: we might not know the running time of M .) □

Non-deterministic choices

A NTM for an NP-language L makes a polynomial number of non-deterministic choices on input $x \in L$.

We can encode these non-deterministic choices into a certificate c , which is a polynomial-length word.

Now, there exists a DTM, which, given x and c , verifies that $x \in L$ in polynomial time.

Thus, $L \in \text{NP}$ iff there is a DTM V and for each $x \in L$ there exists a polynomial-length certificate c such that $V(x, c) = 1$, but $V(y, \cdot) = 0$ for each $y \notin L$.

CNF-SAT is in NP

- A **CNF formula** is a propositional formula in conjunctive normal form: a conjunction (AND) of clauses; each clause is a disjunction (OR) of literals; each literal is a negated or unnegated Boolean variable.
- An assignment $\alpha : \text{var}(F) \rightarrow \{0, 1\}$ satisfies a clause C if it sets a literal of C to true, and it satisfies F if it satisfies all clauses in F .

CNF-SAT

Input: CNF formula F

Question: Does F have a satisfying assignment?

Example: $(x \vee \neg y \vee z) \wedge (\neg x \vee z) \wedge (\neg y \vee \neg z)$.

Lemma 9

CNF-SAT \in **NP**.

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Proof.

Certificate: assignment α to the variables.

Given a certificate, it can be checked in polynomial time whether all clauses are satisfied. □

Brute-force algorithms for problems in NP

Theorem 10

Every problem in NP can be solved in exponential time.

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Proof.

Let Π be an arbitrary problem in NP. [Use certificate-based definition of NP]

We know that \exists a polynomial p and a polynomial-time verification algorithm V such that:

- for every $x \in \Pi$ (i.e., every YES-instance for Π) \exists string $c \in \{0, 1\}^*$, $|c| \leq p(|x|)$, such that $V(x, c) = 1$, and
- for every $x \notin \Pi$ (i.e., every NO-instance for Π) and every string $c \in \{0, 1\}^*$, $V(x, c) = 0$.

Brute-force algorithms for problems in NP

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Every problem in **NP** can be solved in exponential time.

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Let Π be an arbitrary problem in **NP**. [Use certificate-based definition of **NP**] We know that \exists a polynomial p and a polynomial-time verification algorithm V such that:

- for every $x \in \Pi$ (i.e., every **YES**-instance for Π) \exists string $c \in \{0, 1\}^*$, $|c| \leq p(|x|)$, such that $V(x, c) = 1$, and
- for every $x \notin \Pi$ (i.e., every **NO**-instance for Π) and every string $c \in \{0, 1\}^*$, $V(x, c) = 0$.

Now, we can prove there exists an exponential-time algorithm for Π with input x :

- For each string $c \in \{0, 1\}^*$ with $|c| \leq p(|x|)$, evaluate $V(x, c)$ and return **YES** if $V(x, c) = 1$.
- Return **NO**.

Running time: $2^{p(|x|)} \cdot n^{O(1)} \subseteq 2^{O(2 \cdot p(|x|))} = 2^{O(p(|x|))}$, but non-constructive. \square

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Polynomial-time reduction

Definition 11

A language L_1 is **polynomial-time reducible** to a language L_2 , written $L_1 \leq_P L_2$, if there exists a polynomial-time computable function $f : \Sigma^* \rightarrow \Sigma^*$ such that for all $x \in \Sigma^*$,

$$x \in L_1 \Leftrightarrow f(x) \in L_2.$$

A polynomial time algorithm computing f is a **reduction algorithm**.

New polynomial-time algorithms via reductions

Lemma 12

If $L_1, L_2 \in \Sigma^$ are languages such that $L_1 \leq_P L_2$, then $L_2 \in \mathbf{P}$ implies $L_1 \in \mathbf{P}$.*

Definition 13 (NP-hard)

A language $L \subseteq \Sigma^*$ is **NP-hard** if

$$L' \leq_P L \text{ for every } L' \in \text{NP}.$$

Definition 14 (NP-complete)

A language $L \subseteq \Sigma^*$ is **NP-complete** (in **NPC**) if

- 1 $L \in \text{NP}$, and
- 2 L is **NP-hard**.

A first NP-complete problem

Theorem 15

CNF-SAT is NP-complete.

Proved by encoding NTMs into SAT (Cook, 1971; Levin, 1973) and then CNF-SAT (Karp, 1972).

Proving NP-completeness

Lemma 16

*If L is a language such that $L' \leq_P L$ for some $L' \in \text{NPC}$, then L is NP-hard.
If, in addition, $L \in \text{NP}$, then $L \in \text{NPC}$.*

Proving NP-completeness

Lemma 16

If L is a language such that $L' \leq_P L$ for some $L' \in \text{NPC}$, then L is NP-hard.
If, in addition, $L \in \text{NP}$, then $L \in \text{NPC}$.

Proof.

For all $L'' \in \text{NP}$, we have $L'' \leq_P L' \leq_P L$.

By transitivity, we have $L'' \leq_P L$.

Thus, L is NP-hard. □

Proving NP-completeness (2)

Method to prove that a language L is NP-complete:

- ① Prove $L \in \text{NP}$
- ② Prove L is NP-hard.
 - Select a known NP-complete language L' .
 - Describe an algorithm that computes a function f mapping every instance $x \in \Sigma^*$ of L' to an instance $f(x)$ of L .
 - Prove that $x \in L' \Leftrightarrow f(x) \in L$ for all $x \in \Sigma^*$.
 - Prove that the algorithm computing f runs in polynomial time.

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3-CNF SAT is in **NP**, since it is a special case of CNF-SAT.

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To show that 3-CNF SAT is NP-hard, we give a polynomial reduction from CNF-SAT.

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Let F be a CNF formula. The reduction algorithm constructs a 3-CNF formula F' as follows. For each clause C in F :

- If C has at most 3 literals, then copy C into F' .
- Otherwise, denote $C = (\ell_1 \vee \ell_2 \vee \dots \vee \ell_k)$.

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- If C has at most 3 literals, then copy C into F' .
- Otherwise, denote $C = (\ell_1 \vee \ell_2 \vee \dots \vee \ell_k)$. Create $k - 3$ new variables y_1, \dots, y_{k-3} , and add the clauses $(\ell_1 \vee \ell_2 \vee y_1), (\neg y_1 \vee \ell_3 \vee y_2), (\neg y_2 \vee \ell_4 \vee y_3), \dots, (\neg y_{k-3} \vee \ell_{k-1} \vee \ell_k)$.

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Show that F is satisfiable $\Leftrightarrow F'$ is satisfiable.

Show that F' can be computed in polynomial time (trivial; use a RAM). □

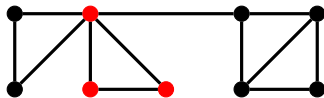
Clique

A **clique** in a graph $G = (V, E)$ is a subset of vertices $S \subseteq V$ such that every two vertices of S are adjacent in G .

CLIQUE

Input: Graph G , integer k

Question: Does G have a clique of size k ?



Theorem 18

CLIQUE is **NP-complete**.

Clique (2)

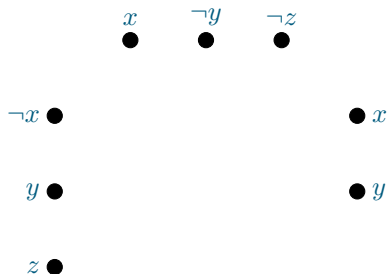
- CLIQUE is in NP

Clique (2)

- CLIQUE is in NP
- Let $F = C_1 \wedge C_2 \wedge \dots \wedge C_k$ be a 3-CNF formula
- Construct a graph G that has a clique of size k iff F is satisfiable

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

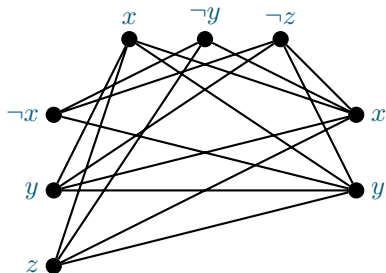
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- CLIQUE is in **NP**
- Let $F = C_1 \wedge C_2 \wedge \dots \wedge C_k$ be a 3-CNF formula
- Construct a graph G that has a clique of size k iff F is satisfiable
- For each clause $C_r = (\ell_1^r \vee \dots \vee \ell_w^r)$, $1 \leq r \leq k$, create w new vertices v_1^r, \dots, v_w^r

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

Clique (2)



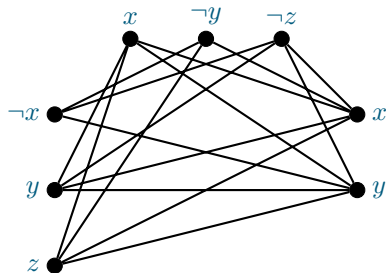
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- Add an edge between v_i^r and v_j^s if

$$r \neq s \quad \text{and} \quad \ell_i^r \neq \neg \ell_j^s \quad \text{where } \neg \neg x = x.$$

- Check correctness and polynomial running time

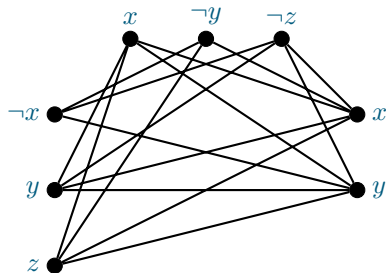
Clique (2)



- Correctness: F has a satisfying assignment iff G has a clique of size k .

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

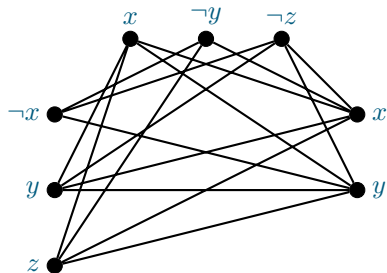
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- Correctness: F has a satisfying assignment iff G has a clique of size k .
- (\Rightarrow): Let α be a sat. assignment for F . For each clause C_r , choose a literal ℓ_i^r with $\alpha(\ell_i^r) = 1$, and denote by s^r the corresponding vertex in G . Now, $\{s^r : 1 \leq r \leq k\}$ is a clique of size k in G since $\alpha(x) \neq \alpha(\neg x)$.

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

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

- Correctness: F has a satisfying assignment iff G has a clique of size k .
- (\Rightarrow) : Let α be a sat. assignment for F . For each clause C_r , choose a literal ℓ_i^r with $\alpha(\ell_i^r) = 1$, and denote by s^r the corresponding vertex in G . Now, $\{s^r : 1 \leq r \leq k\}$ is a clique of size k in G since $\alpha(x) \neq \alpha(\neg x)$.
- (\Leftarrow) : Let S be a clique of size k in G . Then, S contains exactly one vertex $s_r \in \{v_1^r, \dots, v_w^r\}$ for each $r \in \{1, \dots, k\}$. Denote by l^r the corresponding literal. Now, for any r, r' , it is not the case that $l_r = \neg l_{r'}$. Therefore, there is an assignment α to $\text{var}(F)$ such that $\alpha(l_r) = 1$ for each $r \in \{1, \dots, k\}$ and α satisfies F .

Vertex Cover

A **vertex cover** in a graph $G = (V, E)$ is a subset of vertices $S \subseteq V$ such that every edge of G has an endpoint in S .

VERTEX COVER

Input: Graph G , integer k

Question: Does G have a vertex cover of size k ?

Theorem 19

VERTEX COVER is **NP**-complete.

The proof is left as an exercise.

Hamiltonian Cycle

A **Hamiltonian Cycle** in a graph $G = (V, E)$ is a cycle visiting each vertex exactly once.

(Alternatively, a permutation of V such that every two consecutive vertices are adjacent and the first and last vertex in the permutation are adjacent.)

HAMILTONIAN CYCLE

Input: Graph G

Question: Does G have a Hamiltonian Cycle?

Theorem 20

HAMILTONIAN CYCLE is **NP**-complete.

Proof sketch.

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- HAMILTONIAN CYCLE is in **NP**: the certificate is a Hamiltonian Cycle of G .

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- Let us show: VERTEX COVER \leq_P HAMILTONIAN CYCLE

...



Hamiltonian Cycle (2)

Theorem 21

HAMILTONIAN CYCLE is NP-complete.

Proof sketch (continued).

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Hamiltonian Cycle (2)

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Proof sketch (continued).

- Let us show: VERTEX COVER \leq_P HAMILTONIAN CYCLE
- Let $(G = (V, E), k)$ be an instance for VERTEX COVER (VC).
- We will construct an equivalent instance G' for HAMILTONIAN CYCLE (HC).

Hamiltonian Cycle (2)

Theorem 21

HAMILTONIAN CYCLE is NP-complete.

Proof sketch (continued).

- Let us show: VERTEX COVER \leq_P HAMILTONIAN CYCLE
- Let $(G = (V, E), k)$ be an instance for VERTEX COVER (VC).
- We will construct an equivalent instance G' for HAMILTONIAN CYCLE (HC).
- Intuition: Non-deterministic choices
 - for VC: which vertices to select in the vertex cover
 - for HC: which route the cycle takes

...



Hamiltonian Cycle (3)

Theorem 22

HAMILTONIAN CYCLE is NP-complete.

Proof sketch (continued).

- Add k vertices s_1, \dots, s_k to G' (*selector vertices*)

Hamiltonian Cycle (3)

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Proof sketch (continued).

- Add k vertices s_1, \dots, s_k to G' (*selector vertices*)
- Each edge of G will be represented by a gadget (subgraph) of G'
- s.t. the set of edges covered by a vertex x in G corresponds to a partial cycle going through all gadgets of G' representing these edges.

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- Attention: we need to allow for an edge to be covered by both endpoints

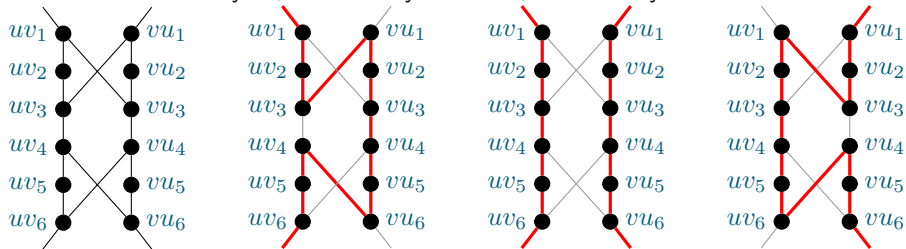
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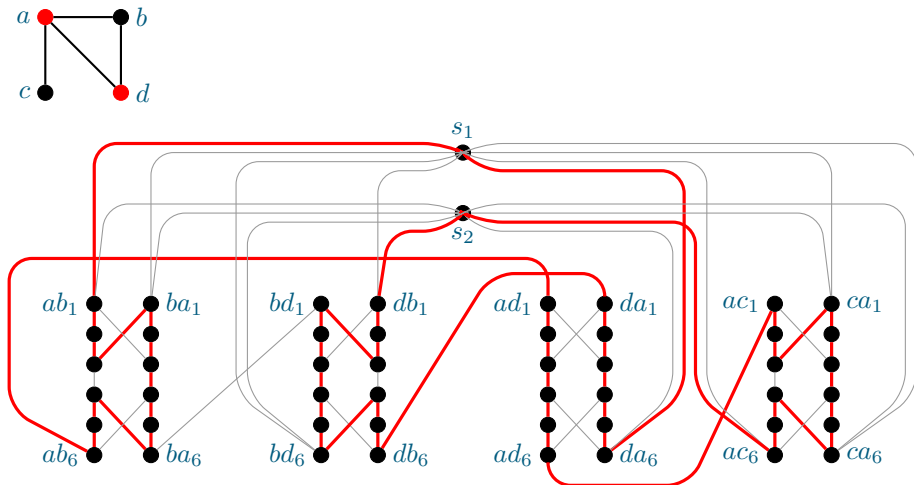
Hamiltonian Cycle (4)

Gadget representing the edge $\{u, v\} \in E$

Its states: 'covered by u ', 'covered by u and v ', 'covered by v '



Hamiltonian Cycle (5)



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Further Reading

- Chapter 34, **NP-Completeness**, in (Cormen et al., 2009)
- Garey and Johnson's influential reference book (Garey and Johnson, 1979)

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