

CMPSC 465

Data Structures and Algorithms

Spring 2022

Instructor: Chunhao Wang

NP and Computational Hardness

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Polynomial-time reduction

(Kleinberg-Tardos, Section 8.1, 8.2)

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A **k-CNF** is a CNF where each clause contains exactly k literals

The Satisfiability Problem

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The k -Satisfiability Problem (k -SAT)

Instance: A k -CNF Φ

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3-SAT and Independent Set

Theorem

$3\text{-SAT} \leq_P \text{Independent Set}$

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Diagram illustrating the formula Φ with annotations:

- Red arrows pointing down to x_1 and \bar{x}_3 are labeled "bad".
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Diagram illustrating the formula Φ with annotations for conflict checking:

- Red arrows pointing down to x_1 and \bar{x}_3 are labeled "bad", indicating a conflict between these two literals.
- Green arrows pointing up to x_1 and x_3 are labeled "good", indicating that these literals are consistent.

We encode a CNF as a graph, and encode an assignment as independent sets (to keep track of the conflicts)

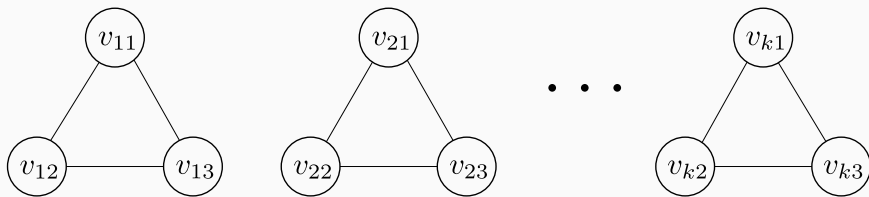
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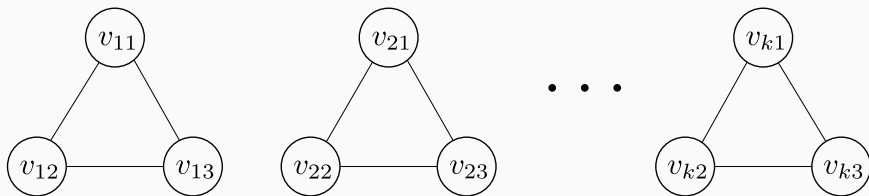
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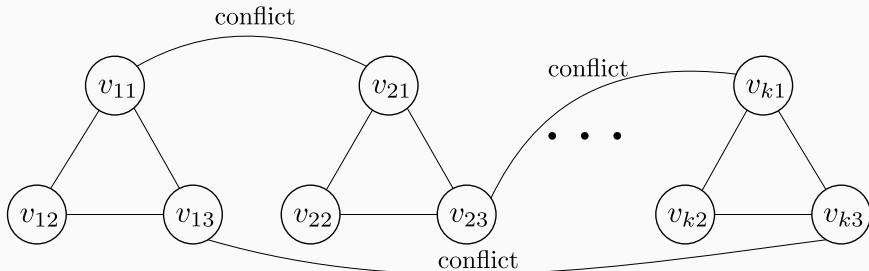
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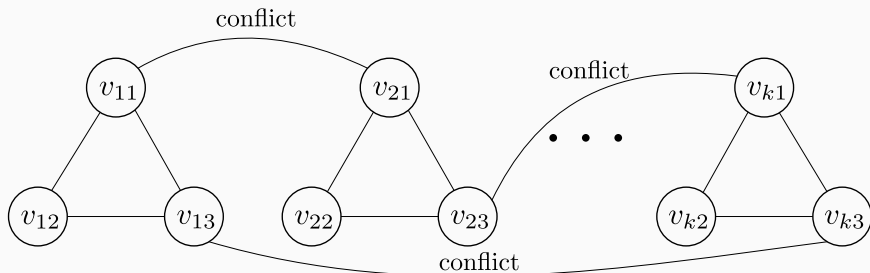
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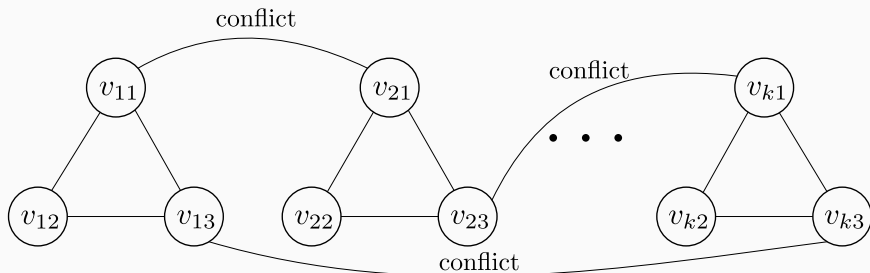


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At most one vertex in each triangle can be in an independent set, so the size of an independent set cannot be larger than k

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So the 3-CNF has a satisfying assignment if and only if G has an independent set of size k

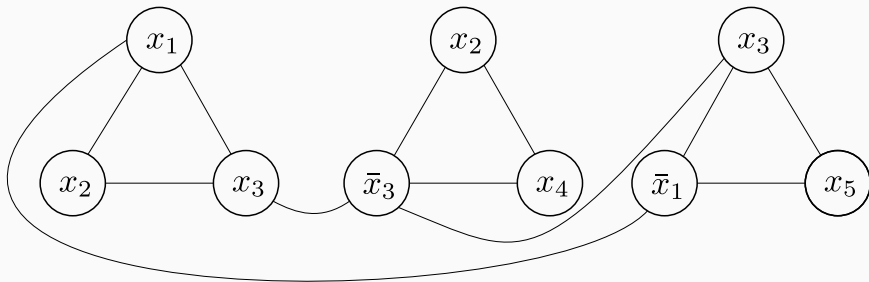


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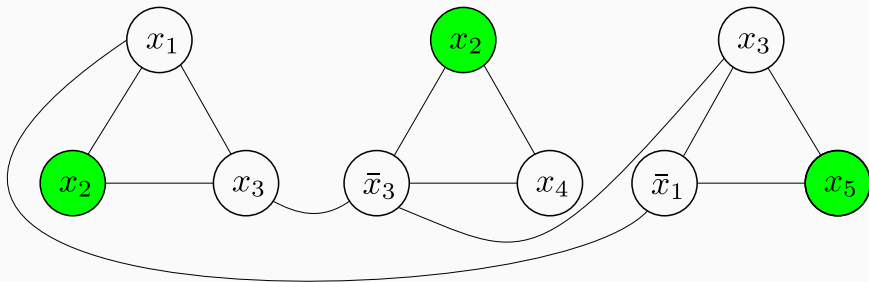
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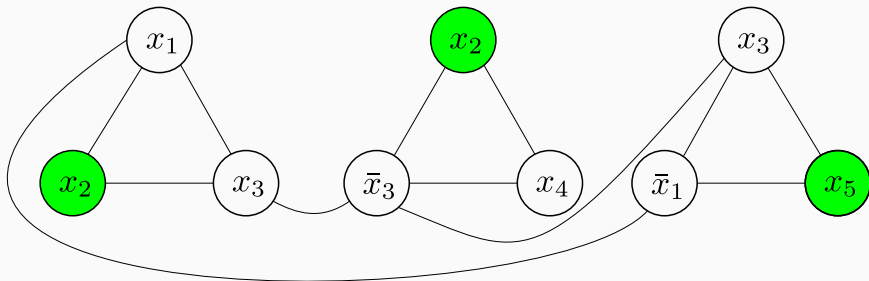
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Satisfying assignment: $x_1 = 0, x_2 = 1, x_3 = 0, x_4 = 0, x_5 = 1$

NP and Computational Hardness

P, NP, and NP-completeness
(Kleinberg-Tardos, Section 8.3, 8.4)

Problems and algorithms

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Computational class

P : the class of all problems for which there exists a polynomial-time algorithm

Checking vs solving

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But there might be exponentially many possible t 's

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Proof.

For any problem in **P** with algorithm A , we construct a certifier B that just returns $A(s)$ with empty certificate t □

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If an NP-complete problem can be solved in polynomial time, then
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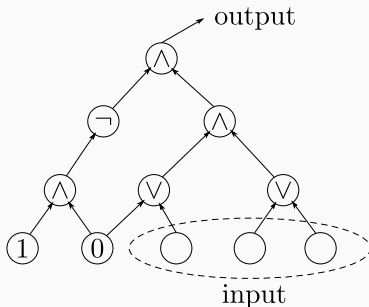
Which problems are NP-complete?

A first **NP**-complete problem: Circuit Satisfiability

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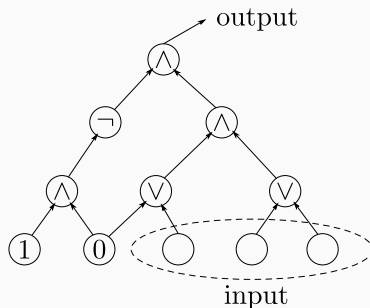


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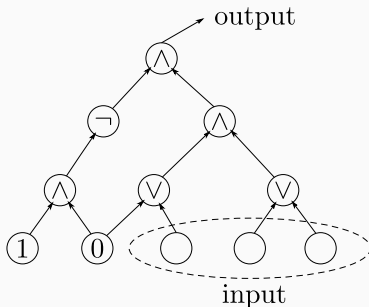


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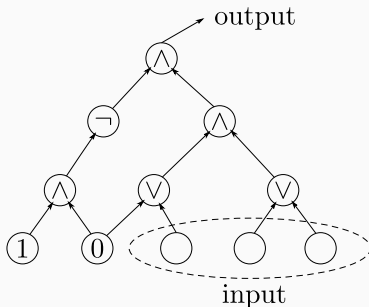


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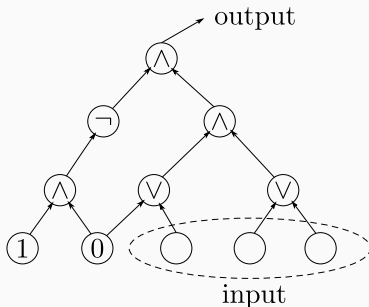


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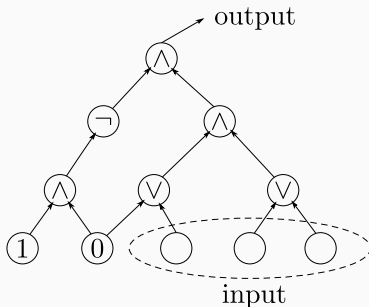


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The Circuit Satisfiability Problem (circuit-SAT)

Instance: A circuit C

Objective: Decide if C is satisfiable

The Cook-Levin Theorem

Theorem (Cook-Levin)

circuit-SAT is NP-complete

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Proof sketch.

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We use the fact that X has a polynomial-time certifier $B(\cdot, \cdot)$

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To decide if $s \in X$, we check if there exists a string t of length $p(|S|)$ s.t. $B(s, t) = \text{yes}$.

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If no, there's such t that $B(s, t) = \text{yes}$. So $s \notin X$.

□