

# Intractability: P and NP

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Programming, Data Structures and Algorithms using Python

Week 11

# Checking algorithms

- Checking algorithm  $C$  for problem
- Takes in an input instance  $I$  and a solution “certificate”  $S$  for  $I$
- $C$  outputs yes if  $S$  represents a valid solution for  $I$ , no otherwise

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- Non-deterministic Turing machines ...

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- If we can solve one efficiently, we can solve them all!

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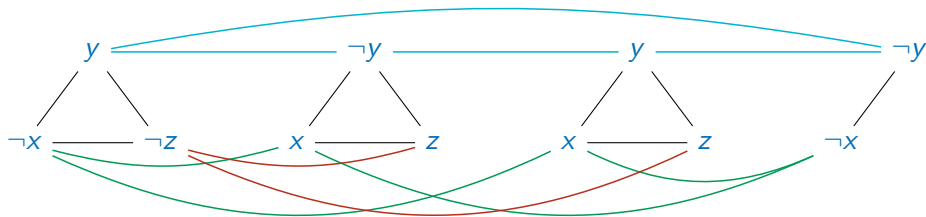
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- Repeat till all clauses are of size 3 or less  $(v \vee \neg w \vee a) \wedge (\neg a \vee x \vee b) \wedge (\neg b \vee \neg y \vee z)$
- If SAT is hard, so is 3-SAT

# 3-SAT to independent set

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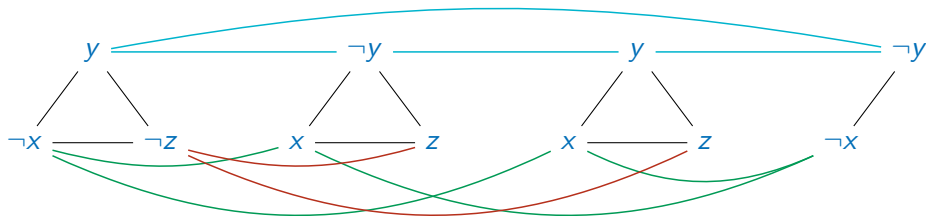
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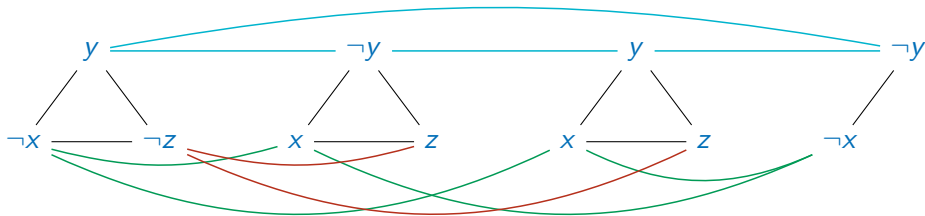


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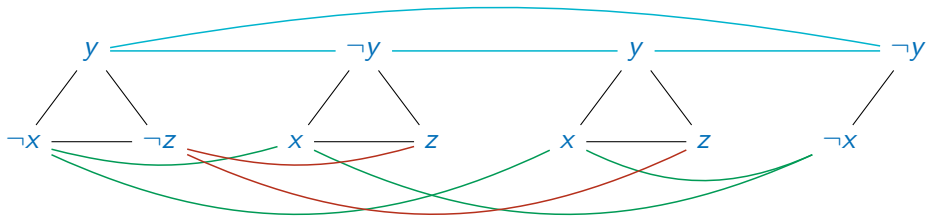
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- Edges enforce consistency across clauses
- Ask for size of independent set = number of clauses

# Reductions within NP

- $\text{SAT} \rightarrow 3\text{-SAT}$ ,  $3\text{-SAT} \rightarrow \text{independent set}$ ,  $\text{independent set} \leftrightarrow \text{vertex cover}$
- Reduction is transitive, so  $\text{SAT} \rightarrow \text{vertex cover}$ , ...
- Other inter-reducible NP problems
  - Travelling salesman, integer linear programming ... All these problems are “equally” hard

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- In general, to show P is NP-complete, reduce some existing NP-complete problem to P

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- Empirical evidence that NP is different from P
- But a formal proof is elusive, and worth \$1 million!