

COMS3261: Computer Science Theory

Fall 2013

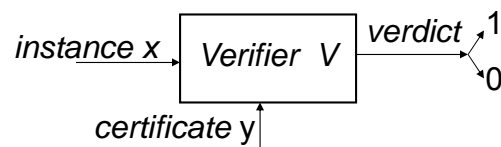
Mihalis Yannakakis

Lecture 25, 12/4/13

NP = short, verifiable certificates

- **Verifier V**: Polynomial time algorithm which checks that y is a certificate for x

$$L = \{x \in A^* \mid \exists y \in B^*, |y| \leq |x|^c, V(x, y) = 1\}$$

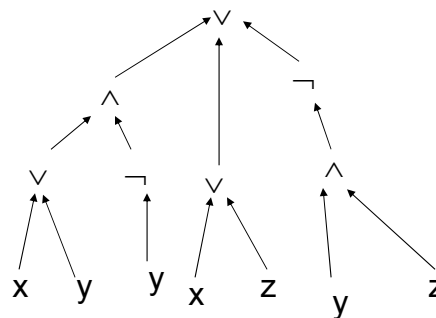


- $x \in L \Rightarrow \exists y \in B^*, |y| \leq |x|^c \text{ s.t. } V(x, y) = 1$
- $x \notin L \Rightarrow \forall y \in B^*, |y| \leq |x|^c. V(x, y) = 0$

Example: (Boolean Formula) Satisfiability

- Input: Boolean formula using AND, OR, NOT connectives (operators) with several variables
- Output: Yes, if there is an assignment to the variables of True/False (1/0) that makes the formula true

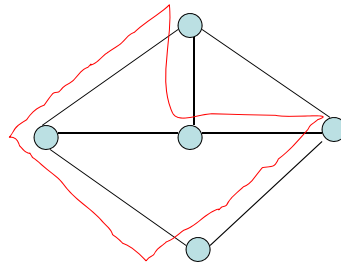
Example: $((x \vee y) \wedge (\neg y)) \vee (x \vee z) \vee (\neg(y \wedge z))$



$x=y=z=0$ satisfies the formula

Examples

- **Graph Hamiltonicity:** Instance: Graph G
certificate y = cycle that goes through every node exactly once (Hamiltonian cycle)
 - Eulerian graph (go through every edge once) in P



Traveling Salesman problem (TSP): cities, distances, TSP tour

Optimization problems

Optimization → Decision Problem

- It is common to define complexity classes like P and NP as classes of *decision problems* - problems with Yes/No (0/1) answer.
- **Optimization problem Π :** Every **instance I** has a set of **solutions**, every solution has a **value** (profit or cost). The problem is to compute the maximum or minimum possible value and a solution that achieves it.
- Corresponding **Decision problem for Π :**
Input: instance I of Π , bound b on value
Output: Yes iff optimum value $\geq b$ for maximization problem (resp. optimum value $\leq b$ for minimization problem), else No

Examples of Decision versions

- **Minimum Coloring Problem:**
Input: Graph G , bound b
Output: Yes if G can be colored with (\leq) b colors, else No
- **Traveling Salesman problem**
Input: Distances between n cities, bound b
Output: Yes, if \exists TSP tour of length $\leq b$
- **Clique Problem**
Input : Graph G , bound b
Output: Yes, if \exists clique with $\geq b$ (mutually adjacent) nodes (mutually compatible entities)

Optimization vs. Decision Problem

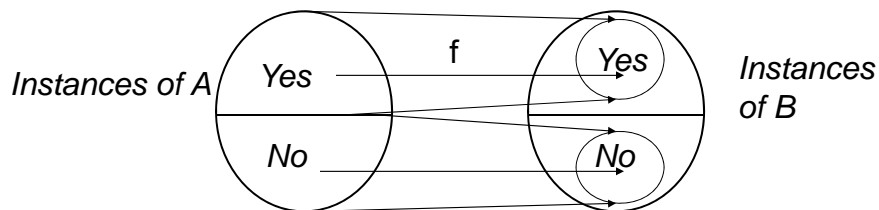
- Optimization problem Π is at least as hard as the corresponding Decision problem: solving opt. Π gives immediately answer for decision
- In most cases, it is possible (but harder) to go also the other way using a routine for the Decision problem repeatedly (but only a polynomial number of times).
- Step 1. Find the optimal value
- Step 2. Find an optimal solution, one piece at a time

Functional problems \rightarrow Decision problems

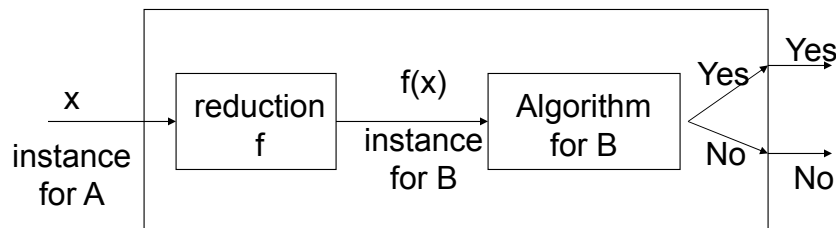
- **Factoring – decision version**
Input: Number N , numbers a, b
Output: Yes, if N has a prime factor between a and b , else No
- Can factor N with polynomially many calls to an algorithm for the decision problem. (Use binary search to find a prime factor of N , divide N by the factor, and repeat).
- Any “functional” problem Π (problem with output) can be reduced to a series of calls to a decision problem:
Input: Instance x of Π , index i
Output: Yes, if the i -th bit of the output of Π on input x is 1, No otherwise

Reductions between Decision Problems

- Polynomial time Reduction from language L to L' (notation $L \leq_P L'$): polynomial time computable function
 $f: \{0,1\}^* \rightarrow \{0,1\}^*$ s.t. $\forall x \in \{0,1\}^*, x \in L \text{ iff } f(x) \in L'$
- Polynomial time Reduction from decision problem A to B (notation $A \leq_P B$): reduction from L_A to L_B (where L_A = set of Yes instances of A and L_B = set of Yes instances of B)
 f maps Yes instances of A to Yes instances of B and No instances of A to No instances of B



Reductions between Decision Problems



- If $B \in P$ then also $A \in P$
- If $A \notin P$ then also $B \notin P$
- If $B \in NP$ then also $A \in NP$
- If $A \notin NP$ then also $B \notin NP$

Reductions compose: $A \leq_P B$ and $B \leq_P C \Rightarrow A \leq_P C$

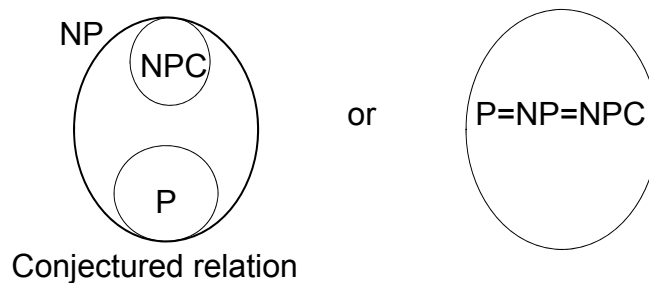
Complementation: $A \leq_P B \Rightarrow \bar{A} \leq_P \bar{B}$

NP-hardness and NP-completeness

- A language L (decision problem) is **NP-hard** if every language L' in NP reduces to it: $\forall L' \in \text{NP}, L' \leq_P L$
- A language L (decision problem) is **NP-complete** if
 1. L is in NP, and
 2. L is NP-hard
- If A is NP-hard and $A \leq_P B$ then B is also NP-hard
- All NP-complete problems reduce to each other

NP-complete problems and P vs NP

- If any NP-hard problem L is in P then $P=NP$
Proof: For every $A \in \text{NP}$, $A \leq_P L$ and $L \in P$ imply $A \in P$
- Either all NP-complete problems are in P (and $P=NP$)
or no NP-complete problem is in P (and $P \neq \text{NP}$)



NP-complete Problems

- (Boolean Formula) Satisfiability
- Graph 3-colorability
- Graph Hamiltonicity
- Travelling Salesman problem
- Clique
-
- And thousands and thousands of other problems from all disciplines are NP-complete or NP-hard.
- Widespread phenomenon!

- We only have exponential time algorithms for solving NP-complete problems.
- It is widely believed that it is impossible to solve them in polynomial time

NP-hard problems for automata and regular expressions

- Is a given regular expression $\equiv (0+1)^*$?
- Is the language of a given NFA $= (0+1)^*$?
- Is $L(A) = L(B)$ for two given NFA A, B ?
- Is $L(R) = L(S)$ for two given regular expressions R, S ?

- All are NP-hard (at least as hard as all problems in NP) but they are not known and are not believed to be in NP
 - Shortest counterexample string to the equality may be exponentially long in the size of the regular expressions or the NFA.