

COMP26120: Tractability and NP Completeness (2018/19)

Lucas Cordeiro

lucas.cordeiro@manchester.ac.uk

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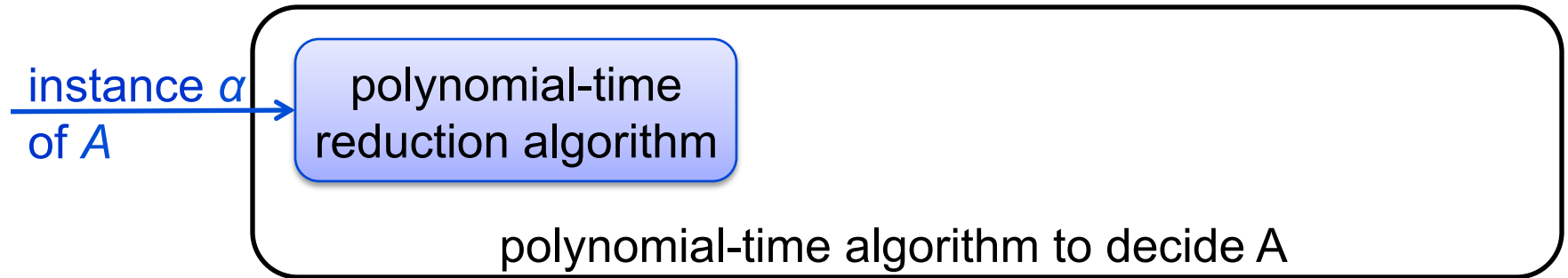
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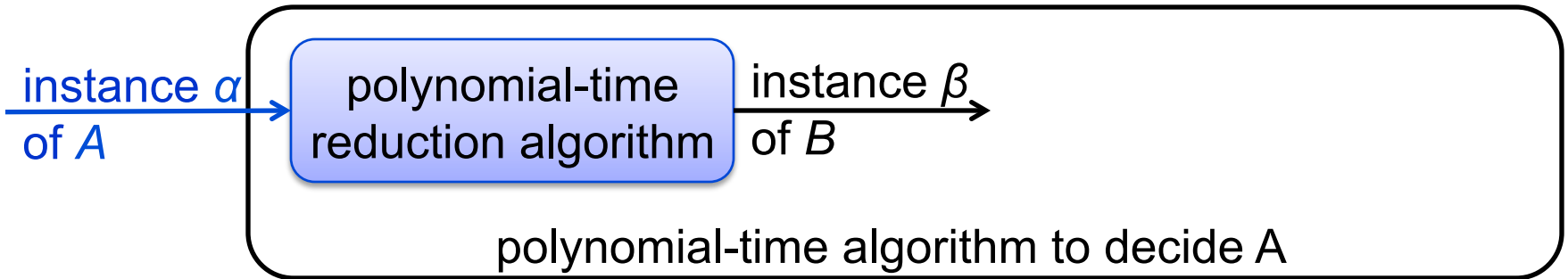
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 - We are trying to prove that **no efficient algorithm is likely to exist**

Polynomial-time Reduction



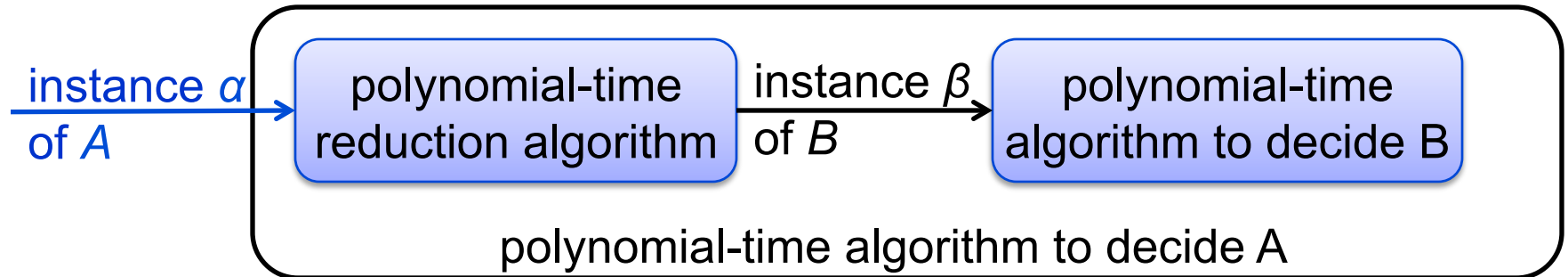
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Polynomial-time Reduction



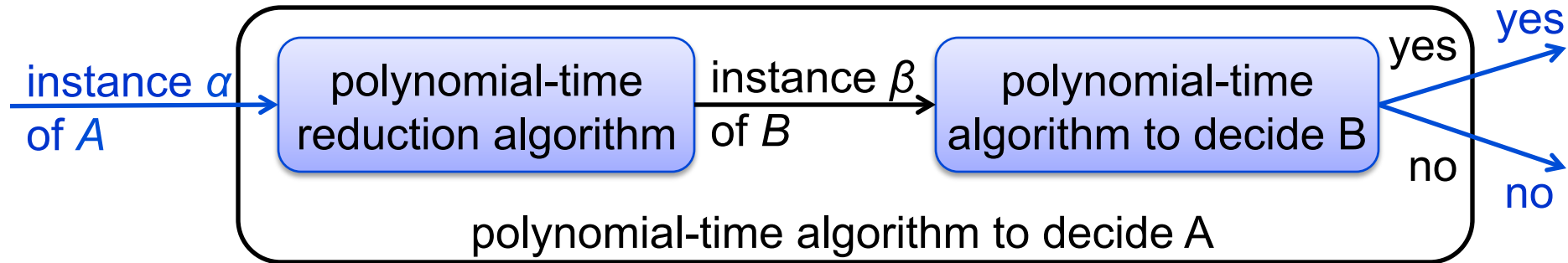
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- 4) Use the answer for β as the answer for **α**

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 4. Prove that the function f satisfies $x \in L'$ iff $f(x) \in L$ for all $x \in \{0, 1\}^*$
 5. Prove that the algorithm computing f runs in **polynomial-time**

1.5 hours

**UNIVERSITY OF MANCHESTER
SCHOOL OF COMPUTER SCIENCE**

COMP26120

January 2018

Time: 00:00 – 00:00

Marking Scheme Included

Do not publish

Answer both questions

The use of electronic calculators is permitted provided they are not programmable and do not store text.

1. Tractability and NP Completeness

- a) (**NP Completeness**) A clique in an undirected graph $G = (V, E)$ is a subset $V' \subseteq V$ of vertices, each pair of which is connected by an edge in E , i.e., a clique is a complete subgraph of G . The size of a clique is the number of vertices it contains. The problem of finding a clique of maximum size in a graph is NP-complete. The decision problem CLIQUE has the corresponding language:

$$CLIQUE = \{\langle G, k \rangle : G \text{ is a graph containing a clique of size } k\}$$

Sketch a proof of NP-completeness for the decision problem CLIQUE.

No marks will be given for simply stating that CLIQUE is NP-complete.

(10 marks)

Model Answer:**1. Prove CLIQUE \in NP.**

For a given $G = (V, E)$, we use the set $V' \subseteq V$ of vertices in the clique as a certificate for G . We can check whether V' is a clique in polynomial time by checking whether for each pair $u, v \in V'$, the edge (u, v) belongs to E .

2. Select Satisfiability of boolean formulas in 3-CNF as a known NP-complete language called 3-CNF-SAT.**3. Describe an algorithm that computes a function f mapping every instance of 3-CNF-SAT to an instance of CLIQUE.**

Let $\phi = C_1 \wedge C_2 \wedge \dots \wedge C_k$ be a boolean formula in 3-CNF with k clauses. For $r = 1, 2, \dots, k$, each clause C_r has exactly three distinct literals l_1^r, l_2^r , and l_3^r . For each clause $C_r = (l_1^r \vee l_2^r \vee l_3^r)$ in ϕ , we place a triple of vertices v_1^r, v_2^r , and v_3^r into V . We put an edge between two vertices v_i^r and v_j^s if both of the following hold:

- v_i^r and v_j^s are in different triples, i.e., $r \neq s$, and
- their corresponding literals are consistent, i.e., l_i^r is not the negation of l_j^s .

4. Show that this transformation of ϕ into G is a reduction.

First, suppose that ϕ has a satisfying assignment. Then each clause C_r contains at least one literal l_i^r that is assigned 1, and each such literal corresponds to a vertex v_i^r . Picking one such “true” literal from each clause yields a set V' of k vertices. We claim that V' is a clique. For any two vertices $v_i^r, v_j^s \in V'$, where $r \neq s$ both corresponding literals l_i^r and l_j^s map to 1 by the given satisfying assignment, and thus the literals cannot be complements. Thus, by the construction of G , the edge (v_i^r, v_j^s) belongs to E .

5. Show that the algorithm computing f runs in polynomial time.

We can easily build this graph from ϕ in polynomial time. As an example of this construction, if we have

$\phi = (x_1 \vee \neg x_2 \vee \neg x_3) \wedge (\neg x_1 \vee x_2 \vee x_3) \wedge (x_1 \vee x_2 \vee x_3)$ then G is the graph shown below.

Distribution of Marks: 2 marks for each step of the proof. No marks for answer without explanation.

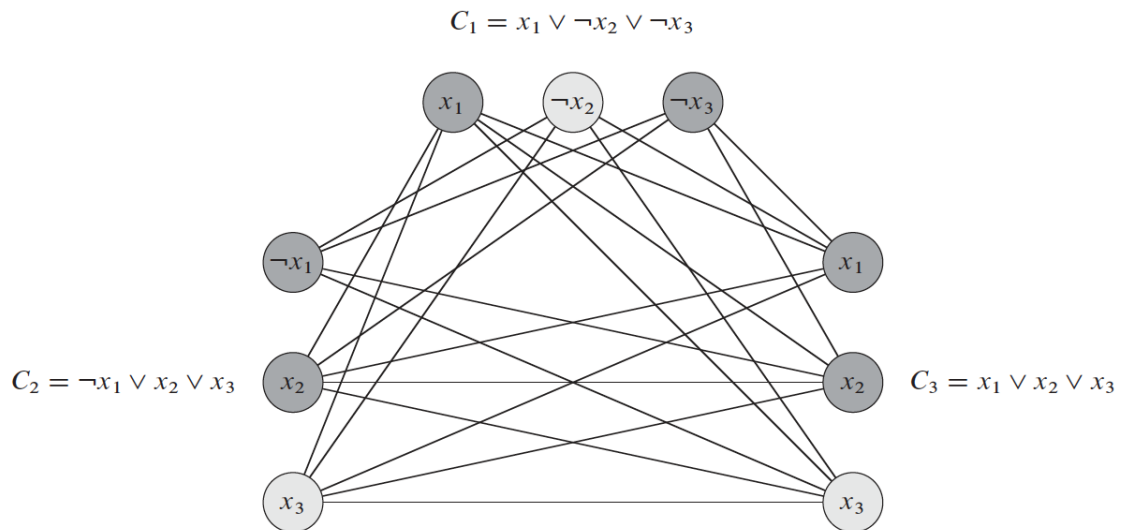


Figure 1: The graph G derived from the 3-CNF formula ϕ . A SAT assignment has $x_2 = 0$, $x_3 = 1$, and x_1 either 0 or 1. This assignment satisfies C_1 with $\neg x_2$, and it satisfies C_2 and C_3 with x_3 , corresponding to the clique with lightly shaded vertices.