

Synchronizing Finite Automata

Lecture III: Complexity Issues

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1. Recap

Deterministic finite automata (DFA): $\mathcal{A} = \langle Q, \Sigma, \delta \rangle$.

- Q the state set
- Σ the input alphabet
- $\delta : Q \times \Sigma \rightarrow Q$ the transition function

\mathcal{A} is called **synchronizing** if there exists a word $w \in \Sigma^*$ whose action resets \mathcal{A} , that is, leaves the automaton in one particular state no matter which state in Q it started at: $\delta(q, w) = \delta(q', w)$ for all $q, q' \in Q$.

$|Q \cdot w| = 1$. Here $Q \cdot v = \{\delta(q, v) \mid q \in Q\}$.

Any w with this property is a **reset word** for \mathcal{A} .

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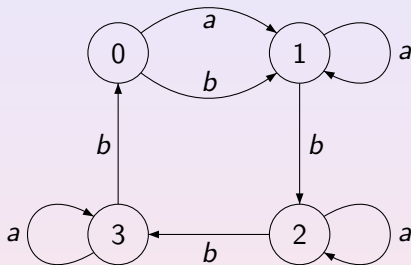
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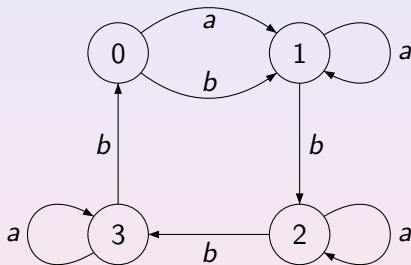
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3. Short Reset Words are Hard to Find

There is a algorithm that uses a natural greedy strategy and, when given a synchronizing automaton \mathcal{A} with n states, finds a reset word of length at most $\frac{n^3-n}{6}$ for \mathcal{A} spending polynomial time as a function of n . (In fact, time is $O(n^3)$).

However, it is known that the gap between the size of the solution found by this greedy algorithm question or any of its versions and the size of the optimal solution (i.e., a reset word of minimum length) can be arbitrarily large.

Now we aim to prove that under standard assumptions (like $NP \neq coNP$) no polynomial algorithm, **even non-deterministic**, can find the minimum length of reset words for synchronizing automata.

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4. 5-minute Tour in Complexity Theory

Recall what are P, NP, coNP, etc.

These are classes of **combinatorial decision problems**, i.e., problems whose input is a **finite object** (graph, formula, automaton, ...). and whose question is whether or not a given object possesses a certain **property** (which usually gives the name to the problem). The answer to each concrete instance of such a problem is either YES or NO.

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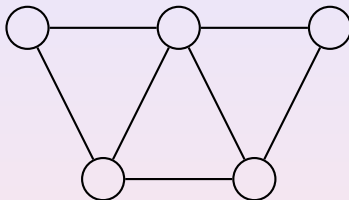
5. Example: Graph Coloring

The input of k -COLOR is a graph G .

The question is whether the vertices of G can be labeled with k colors so that adjacent vertices are assigned different colors.
For the above graph, the answer to 3-COLOR is YES
while the answer to 2-COLOR is NO.

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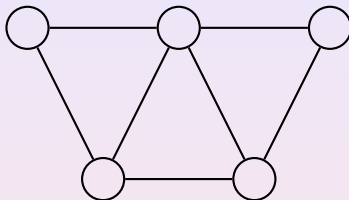
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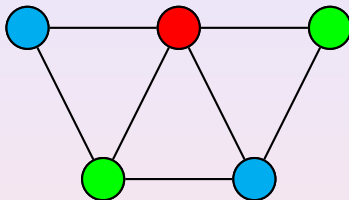


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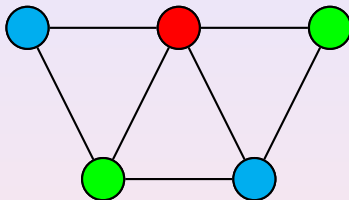
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Merlin, a wizard

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A problem is in P if Arthur can solve it in polynomial time (of the size of its input).

Example: 2-COLOR is in P since Arthur can check in polynomial time whether or not all simple cycles of a given graph are of even length.

A problem is in NP if, **whenever the answer to its instance is YES**, Merlin can convince Arthur that the answer is YES in polynomial time (of the size of the input).

Example: 3-COLOR is in NP since, given a 3-colorable graph, Merlin can exhibit its 3-coloring, and Arthur can check in polynomial time that this coloring is correct.

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8. Classes P, NP, and coNP

Clearly $P \subseteq NP$ and $P \subseteq \text{coNP}$.

Is any of the inclusions strict? In other words, is it true that $P \neq NP$?

This is a VERY BIG PROBLEM
which is worth \$1000000 (before tax).

According to the present paradigm, we assume that
 $P \neq NP \neq \text{coNP}$.

An NP-hard problem is a problem to which any problem from NP
can be reduced in polynomial time.

An NP-complete problem is a problem in NP that at the same time
is NP-hard.

Example: 3-COLOR is NP-complete (Leonid Levin, 1973).

How can one prove that a problem is NP-hard? Via a polynomial
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Consider the following decision problem:

SHORT-RESET-WORD: Given a synchronizing automaton $\mathcal{A} = \langle Q, \Sigma, \delta \rangle$ and a positive integer ℓ , is it true that \mathcal{A} has a reset word of length ℓ ?

Clearly, **SHORT-RESET-WORD** belongs to NP: Merlin can non-deterministically guess a word $w \in \Sigma^*$ of length ℓ and then Arthur can check if w is a reset word for \mathcal{A} in time $\ell|Q|$.

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10. Reduction from SAT

Given an instance ψ of SAT with n variables x_1, \dots, x_n and m clauses c_1, \dots, c_m , one constructs $\mathcal{A}(\psi)$ with 2 input letters a and b and the state set $\{z, q_{i,j} \mid 1 \leq i \leq m, 1 \leq j \leq n+1\}$.

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$$q_{i,j} \cdot a = \begin{cases} z & \text{if } x_j \text{ occurs in } c_i, \\ q_{i,j+1} & \text{otherwise} \end{cases} \quad \text{for } 1 \leq i \leq m, 1 \leq j \leq n;$$

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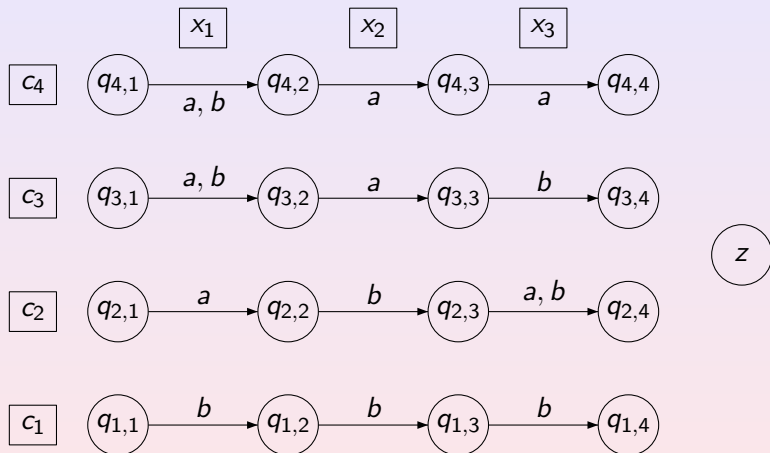
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It is easy to see that $\mathcal{A}(\psi)$ is reset by every word of length $n + 1$ and is reset by a word of length n if and only if ψ is satisfiable.

Thus, assigning the instance $(\mathcal{A}(\psi), n)$ of SHORT-RESET-WORD to an arbitrary n -variable instance ψ of SAT, one gets a polynomial reduction which is in fact parsimonious.

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If we change ψ to $\{x_1 \vee x_2, \neg x_1 \vee x_2, \neg x_2 \vee x_3, \neg x_2 \vee \neg x_3\}$, it becomes unsatisfiable and $\mathcal{A}(\psi)$ is reset by no word of length 3.

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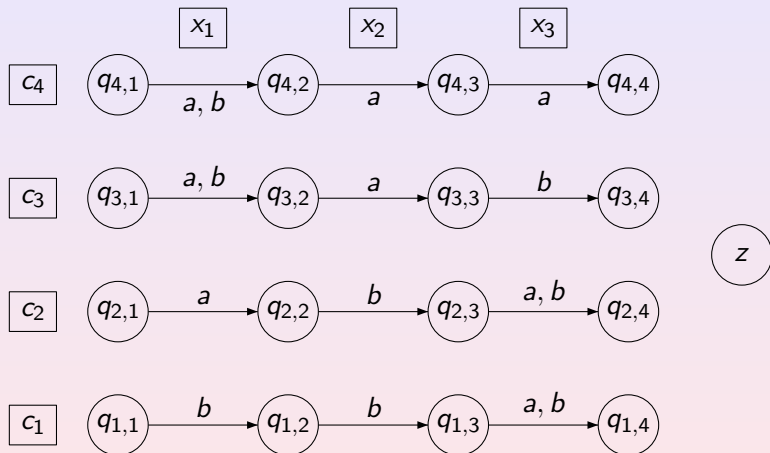
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Now consider the following decision problem:

SHORTEST-RESET-WORD: *Given a synchronizing automaton \mathcal{A} and a positive integer ℓ , is it true that the minimum length of a reset word for \mathcal{A} is equal to ℓ ?*

Assigning the instance $(\mathcal{A}(\psi), n + 1)$ of SHORTEST-RESET-WORD to an arbitrary system ψ of clauses on n variables, one sees that the answer to the instance is “Yes” if and only if ψ is **not** satisfiable. This is a polynomial reduction from the **negation** of SAT to SHORTEST-RESET-WORD whence the latter problem is coNP-hard. As a corollary, SHORTEST-RESET-WORD cannot belong to NP unless $\text{NP} = \text{coNP}$.

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15. Computing is Harder than Deciding

$P^{NP[\log]}$ is the class of all problems that can be solved by a deterministic polynomial-time Turing machine that has an access to an oracle for an NP-complete problem, with the number of queries being logarithmic in the size of the input.

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16. Non-approximability: Constant Factor

However, all these results were consistent with the existence of very good polynomial approximation algorithms for the problem!

Mikhail Berlinkov has shown that under $NP \neq P$, for no k , there may exist a polynomial algorithm that, given a synchronizing automaton with two input letters, produces a reset word whose length is less than $k \times$ minimum possible length of a reset word.

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Gerbush and Heeringa have observed that SET COVER admits a transparent reduction to the problem of finding a reset word of minimum length for a given synchronizing automaton. Recall that an instance of SET COVER consists of a set S , a family $\{C_i\}_{i \in I}$ of subsets of S and a positive integer N . The question is whether or not there exists a subset $J \subseteq I$ such that $|J| \leq N$ and $\bigcup_{j \in J} C_j = S$. Using a difficult result on SET COVER by Alon, Moshkovitz and Safra, Gerbush and Heeringa have deduced that the minimum length of reset words for synchronizing automata with n states and **unbounded** alphabet cannot be approximated within the factor $c \log n$ for some constant $c > 0$ unless $P = NP$. Berlinkov has obtained a similar result for synchronizing automata with only 2 input letters.

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18. Non-approximability: Sublinear Factor

Very recently, Gawrychowski and Straszak have shown that for every $\varepsilon > 0$ it is not possible to approximate the length of the shortest reset word for synchronizing automata with n states within a factor of $n^{1-\varepsilon}$ in polynomial time, unless $P = NP$.