

طراحی الگوریتم ها

مبحث نوزدهم: پیچیدگی محاسبات

سجاد شیرعلی شمرضا

بهار 1402

سه شنبه، 9 خرداد 1402

اطلاع رسانی

- آخرین مبحث سال!

زمان حل چند جمله ای

Idea

- Well-known algorithm rule: Polynomial good, exponential bad!
 - The latter is obvious, the former may need some explanation
 - We say that polynomial-time problems are **tractable**
 - I.e., exponential problems are **intractable**

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- It's not exponential!
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 - If $f(t)$ and $g(t)$ are polynomials, so is $f(g(t))$
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- Almost all of the algorithms we have studied in this course had polynomial time

Decision Problems

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 - **Instance description:** defines the information expected in the input
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 - Refers to variables that are defined in the instance description
- Almost every optimization problem can be expressed in decision problem form

Decision Problem Example 1

- **Definition:**
 - In a graph $G=(V,E)$, a **clique** E is a subset of V such that for all u and v in E , the edge (u,v) is in E .

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- **k-Clique Decision problem**
 - **Instance:** an undirected graph $G=(V,E)$.
 - Note: k is some constant, independent of the problem
 - **Question:** Does G contain a clique of k vertices?

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- **Graph Coloring Decision Problem**
 - **Instance:** an undirected graph $G=(V,E)$ and an integer $k>0$.
 - **Question:** Is there a coloring of G that uses no more than k colors?

Decision Problem Example 3

- **Definition:**
 - Suppose we have an unlimited number of bins, each with capacity 1.0, and n objects with sizes s_1, \dots, s_n , where $0 < s_i \leq 1$ (all s_i rational)

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- **Bin Packing Decision Problem**
 - **Instance:** s_1, \dots, s_n as described above, and an integer k .
 - **Question:** Can the n objects be packed into k bins?



سوال؟

تقليل

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- If there is an algorithm for q , create an algorithm for p :
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- If T is a function with polynomially bounded running time:
 - We say: p is **polynomially reducible** to q
 - We write: $p \leq_p q$
- For now, reducible means polynomially reducible

Class P

- **Definition:** An algorithm is polynomially bounded if its worst-case complexity is big-O of a polynomial function of the input size n .
 - I.e. there is a single polynomial p such that for each input of size n , the algorithm terminates after at most $p(n)$ steps
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 - Informally (with slight abuse of notation), we can say that polynomially bounded optimization problems are in P

Example of a problem in P

- Minimum Spanning Tree (MST)
- **Input:** A weighted graph $G=(V,E)$ with n vertices [each edge e is labeled with a non-negative weight $w(e)$], and a number k .
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 - Find the MST and check whether its cost is less than k

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 - No. The size to represent k is $\log k$

Class NP

- NP: Nondeterministic Polynomial time
- First stage: assumes a “guess” of a possible solution
- Can we verify in polynomial time whether the proposed solution is a correct solution?
 - I.e., do we have a verifier that verifies an answer in a polynomial time?

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 - E.g., a coloring that
 - May or may not have only k colors
 - May or may not have distinct colors for adjacent nodes
- The problem is in NP **if and only if** there is a polynomial-time (in N) algorithm that can check a proposed solution to see if it really is a solution

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 - The nondeterministic “**guessing**” phase: produce the proposed solution
 - The deterministic **verifying** phase: check the proposed solution to see if it is indeed a solution
 - Output yes if it is a solution
- NP Class: class of decision problems for which there is a polynomially bounded nondeterministic algorithm
 - Examples: Graph coloring, Bin packing, Clique



سوال؟

رابطه بین P و NP

Problem Class Containment

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- Then $P \subseteq NP \subseteq \text{Exp}$
- $P \subseteq NP$
 - Directly solve the problem to find the answer
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- $NP \subseteq \text{Exp}$
 - Generate all possible solutions
 - Will be exponential in terms of the input
 - Check them to see if one of them is the answer
 - Will be deterministic

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- Try to solve it as an extra assignment
 - Will give you extra mark for course if you do it!

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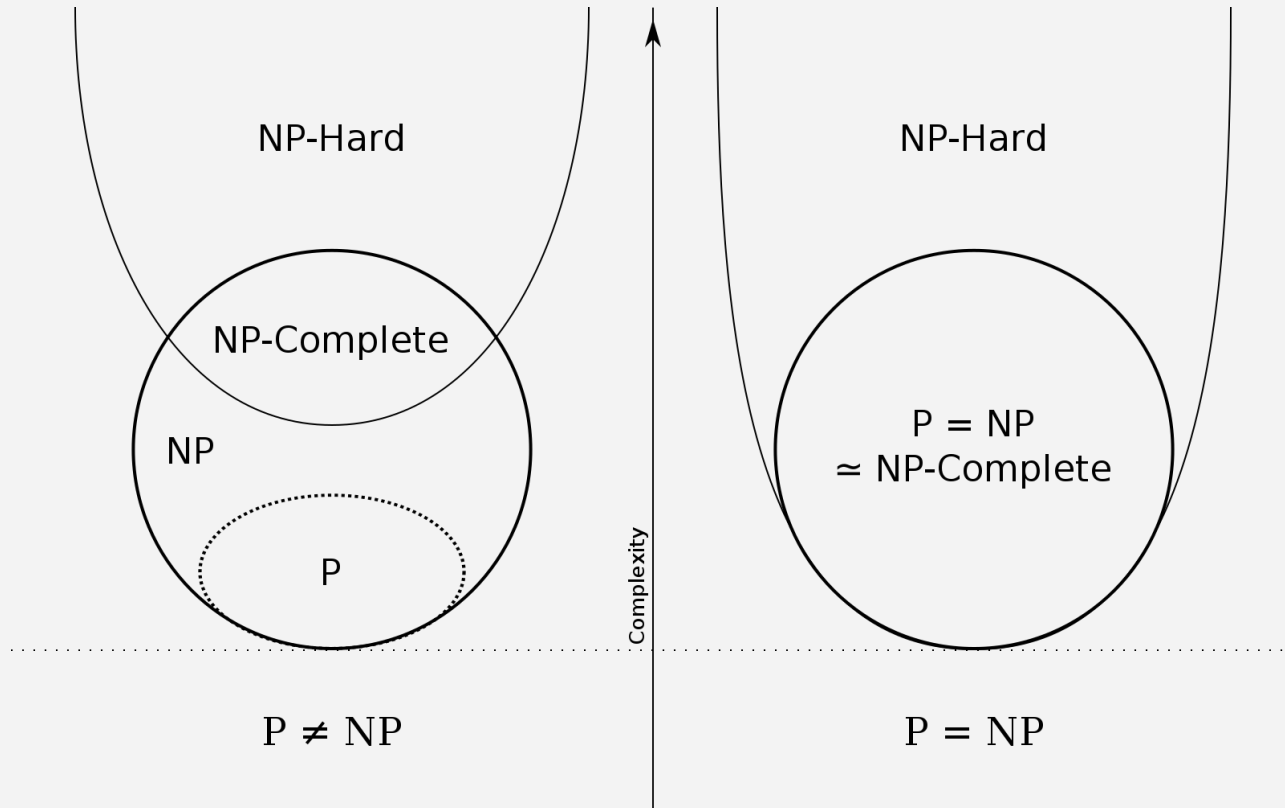
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- NP – H includes all NP - C problems

Relation between P, NP, NP-H, NP-C



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 - Example: 3-satisfiability
- If p is NP-hard, and $p \leq_p q$, then q is NP-hard.
 - Most NP-complete problems are shown to be NP-C by showing that 3-satisfiability (or some other known NP-complete problem) reduces to them

3-SAT

- 3-Satisfiability problem:
- A CNF (conjunctive normal form) formula is in 3-CNF if every clause has exactly three literals
- **Instance:** A 3CNF propositional formula f (containing n different variables)
- **Question:** Is there a truth assignment that satisfies f ?

$$(\neg a \vee b \vee c) \wedge (\neg b \vee a \vee c) \wedge (\neg c \vee a \vee b) \wedge (\neg d \vee a \vee b) \wedge (\neg e \vee a \vee b) \wedge (\neg a \vee b \vee d) \wedge (\neg b \vee a \vee d)$$



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