

NP Completeness

Lecture 3

September 3, 2015

3.1: Definition of NP

Problems

- | | |
|--------------------------|--------------------|
| ① Clique | ① Set Cover |
| ② Independent Set | ② SAT |
| ③ Vertex Cover | ③ 3SAT |

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Relationship

Independent Set \leq_P Clique

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Relationship

Independent Set \leq_P **Clique** \leq_P **Independent Set**

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Independent Set \approx_P **Clique**

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Independent Set \approx_P **Clique**

Independent Set \leq_P **Vertex Cover**

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Relationship

Independent Set \approx_P **Clique**

Independent Set \leq_P **Vertex Cover** \leq_P **Independent Set**

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Vertex Cover \approx_P **Independent Set** \approx_P **Clique**
3SAT \leq_P **SAT**

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Vertex Cover \approx_P Independent Set \approx_P Clique
 $3SAT \leq_P SAT \leq_P 3SAT$

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Vertex Cover \approx_P Independent Set \approx_P Clique
3SAT \approx_P SAT

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Relationship

Vertex Cover \approx_P Independent Set \approx_P Clique

3SAT \approx_P SAT

3SAT \leq_P Independent Set

3.1.1: Preliminaries

3.1.1.1: Problems and Algorithms

Problems and Algorithms: Formal Approach

Decision Problems

- 1 **Problem Instance:** Binary string s , with size $|s|$
- 2 **Problem:** Set X of strings s.t. answer is “yes”: members of X are **YES instances** of X .
Strings not in X are **NO instances** of X .

Definition

- 1 **alg:** algorithm for problem X if $\text{alg}(s) = \text{“yes”} \iff s \in X$.
- 2 **alg** have **polynomial running time** $\exists p(\cdot)$ polynomial s.t. $\forall s$, **alg**(s) terminates in at most $O(p(|s|))$ steps.

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Polynomial Time

Definition

Polynomial time (denoted by **P**): class of all (decision) problems that have an algorithm that solves it in polynomial time.

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Example

Problems in **P** include

- 1 Is there a shortest path from s to t of length $\leq k$ in G ?
- 2 Is there a flow of value $\geq k$ in network G ?
- 3 Is there an assignment to variables to satisfy given linear constraints?

Efficiency Hypothesis

Efficiency hypothesis.

A problem X has an efficient algorithm

$\iff X \in \mathbf{P}$, that is X has a polynomial time algorithm.

1 Justifications:

- 1 Robustness of definition to variations in machines.
- 2 A sound theoretical definition.
- 3 Most known polynomial time algorithms for “natural” problems have small polynomial running times.

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Problems that are hard...

...with no known polynomial time algorithms

Problems

- 1 **Independent Set**
- 2 **Vertex Cover**
- 3 **Set Cover**
- 4 **SAT**
- 5 **3SAT**

- 1 undecidable problems are way harder (no algorithm at all!)
- 2 ...but many problems want to solve: similar to above.
- 3 **Question:** What is common to above problems?

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Efficient Checkability

- 1 Above problems have the property:

Checkability

For any **YES** instance I_X of X :

- (A) there is a proof (or certificate) C .
- (B) Length of certificate $|C| \leq \text{poly}(|I_X|)$.
- (C) Given C, I_x : efficiently check that I_X is **YES** instance.

- 2 Examples:

- 1 **SAT** formula φ : proof is a satisfying assignment.
- 2 **Independent Set** in graph G and k :
Certificate: a subset S of vertices.

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- 1 **SAT** formula φ : proof is a satisfying assignment.
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3.1.2: Certifiers/Verifiers

Definition

Algorithm $C(\cdot, \cdot)$ is **certifier** for problem X : $\forall s \in X$ there $\exists t$ such that $C(s, t) = \text{"YES"}$, and conversely, if for some s and t , $C(s, t) = \text{"yes"}$ then $s \in X$.

t is the **certificate** or **proof** for s .

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Definition (Efficient Certifier.)

Certifier C is **efficient certifier** for X if there is a polynomial $p(\cdot)$ s.t. for every string s :

- ★ $s \in X$ if and only if
- ★ there is a string t :
 - ① $|t| \leq p(|s|)$,
 - ② $C(s, t) = \text{"yes"}$,
 - ③ and C runs in polynomial time.

Example: Independent Set

- ① **Problem:** Does $\mathbf{G} = (\mathbf{V}, \mathbf{E})$ have an independent set of size $\geq k$?
 - ① **Certificate:** Set $\mathbf{S} \subseteq \mathbf{V}$.
 - ② **Certifier:** Check $|\mathbf{S}| \geq k$ and no pair of vertices in \mathbf{S} is connected by an edge.

3.1.3: Examples

Example: Vertex Cover

- ① **Problem:** Does G have a vertex cover of size $\leq k$?
- ① **Certificate:** $S \subseteq V$.
- ② **Certifier:** Check $|S| \leq k$ and that for every edge at least one endpoint is in S .

Example: SAT

- ① **Problem:** Does formula φ have a satisfying truth assignment?
 - ① **Certificate:** Assignment a of **0/1** values to each variable.
 - ② **Certifier:** Check each clause under a and say “yes” if all clauses are true.

Composite

Instance: A number s .

Question: Is the number s a composite?

- 1 **Problem: Composite.**
 - 1 **Certificate:** A factor $t \leq s$ such that $t \neq 1$ and $t \neq s$.
 - 2 **Certifier:** Check that t divides s .

3.2: *NP*

3.2.1: Definition

Nondeterministic Polynomial Time

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Example

Independent Set, **Vertex Cover**, **Set Cover**, **SAT**, **3SAT**, and **Composite** are all examples of problems in **NP**.

Why is it called...

Nondeterministic Polynomial Time

- 1 Certifier is algorithm $C(I, c)$ with two inputs:
 - 1 I : instance.
 - 2 c : proof/certificate that the instance is indeed a **YES** instance of the given problem.
- 2 C “algorithm” for original problem, if:
 - 1 Given I , the algorithm guess (non-deterministically, and who knows how) the certificate c .
 - 2 Algorithm verifies certificate c for the instance I .
- 3 Usually **NP** is described using Turing machines (gag).

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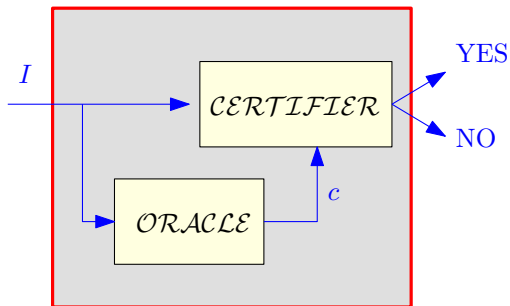
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Certifiers as algorithms...

...with a little help from an oracle friend.



- 1 Oracle: Guesses certificate c for given instance I .
- 2 Certifier: Polynomial time, given I and c , verify that indeed c proves that I is a YES instance.

Asymmetry in Definition of NP

① Only **YES** instances have a short proof/certificate. NO instances need not have a short certificate.

② For example...

Example

SAT formula φ . No easy way to prove that φ is NOT satisfiable!

③ More on this and **co-NP** later on.

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3.2.2: Intractability

P versus NP

Proposition

P \subseteq **NP**.

For a problem in **P** no need for a certificate!

Proof.

Consider problem $X \in \mathbf{P}$ with algorithm **alg**. Need to demonstrate that X has an efficient certifier:

- 1 Certifier C (input s, t):
runs **alg**(s) and returns its answer.
- 2 C runs in polynomial time.
- 3 If $s \in X$, then for every t , $C(s, t) = \text{"YES"}$.
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Exponential Time

Definition

Exponential Time (denoted **EXP**) set of all problems with algorithm that runs in exponential time.

For input s : Running time is $O(2^{\text{poly}(|s|)})$.

Example: $O(2^n)$, $O(2^{n \log n})$, $O(2^{n^3})$, ...

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NP versus EXP

Proposition

NP \subseteq **EXP**.

Proof.

Let $X \in \mathbf{NP}$ with certifier C . Need to design an exponential time algorithm for X .

- 1 For every t , with $|t| \leq p(|s|)$ run $C(s, t)$; answer “yes” if any one of these calls returns “yes”.
- 2 The above algorithm correctly solves X (exercise).
- 3 Algorithm runs in $O(q(|s| + |p(s)|)2^{p(|s|)})$, where q is the running time of C . □

Examples

- ① **SAT**: try all possible truth assignment to variables.
- ② **Independent Set**: try all possible subsets of vertices.
- ③ **Vertex Cover**: try all possible subsets of vertices.

Is **NP** efficiently solvable?

We know **P** \subseteq **NP** \subseteq **EXP**.

Is **NP** efficiently solvable?

We know $\mathbf{P} \subseteq \mathbf{NP} \subseteq \mathbf{EXP}$.

Big Question

Is there are problem in **NP** that **does not** belong to **P**? Is $\mathbf{P} = \mathbf{NP}$?

If $P = NP$...

Or: If pigs could fly then life would be sweet.

- 1 Many important optimization problems can be solved efficiently.
- 2 The **RSA** cryptosystem can be broken.
- 3 No security on the web.
- 4 No e-commerce ...
- 5 Creativity can be automated! Proofs for mathematical statement can be found by computers automatically (if short ones exist).

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P versus NP

Status

Relationship between **P** and **NP** remains one of the most important open problems in mathematics/computer science.

Consensus: Most people feel/believe **P** \neq **NP**.

Resolving **P** versus **NP** is a Clay Millennium Prize Problem. You can win a million dollars in addition to a Turing award and major fame!

3.3: NP Completeness

Definition

An algorithm $C(\cdot, \cdot)$ is a **certifier** for problem X if for every $s \in X$ there is some string t such that $C(s, t) = \text{"yes"}$, and conversely, if for some s and t , $C(s, t) = \text{"yes"}$ then $s \in X$.

The string t is called a **certificate** or **proof** for s .

Definition (Efficient Certifier.)

A certifier C is an **efficient certifier** for problem X if there is a polynomial $p(\cdot)$ such that for every string s , we have that

- ★ $s \in X$ if and only if
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NP-Complete Problems

Definition

A problem X is said to be **NP-Complete** if

- 1 $X \in \mathbf{NP}$, and
- 2 (**Hardness**) For any $Y \in \mathbf{NP}$, $Y \leq_P X$.

Solving **NP-Complete** Problems

Proposition

Suppose X is **NP-Complete**. Then X can be solved in polynomial time if and only if **P = NP**.

Proof.

\Rightarrow Suppose X can be solved in polynomial time

- ① Let $Y \in \mathbf{NP}$. We know $Y \leq_P X$.
- ② We showed that if $Y \leq_P X$ and X can be solved in polynomial time, then Y can be solved in polynomial time.
- ③ Thus, every problem $Y \in \mathbf{NP}$ is such that $Y \in P$;
 $\mathbf{NP} \subseteq P$.
- ④ Since $\mathbf{P} \subseteq \mathbf{NP}$, we have **P = NP**.

\Leftarrow Since **P = NP**, and $X \in \mathbf{NP}$, we have a polynomial time algorithm for X . □

NP-Hard Problems

- 1 Formal definition:

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A problem X is said to be **NP-Hard** if

- 1 (**Hardness**) For any $Y \in \mathbf{NP}$, we have that $Y \leq_P X$.
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- 3 **Example:** Halting problem is **NP-Hard** (why?) but not **NP-Complete**.

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Consequences of proving **NP-Completeness**

① If X is **NP-Complete**

- ① Since we believe $P \neq NP$,
- ② and solving X implies $P = NP$.

X is **unlikely** to be efficiently solvable.

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