## Introduction to Approximation Algorithms

Mong-Jen Kao (高孟駿)

Friday 13:20 – 15:10

## The Complexity Class NP

&

**Proof Checking** 

The classic point of view

## The Complexity Class NP

A language L is in NP if there is a <u>nondeterministic Turing machine</u> (NTM) M that decides it in polynomial-time.

For any string x,

- If  $x \in L$ , then there exists a computation path of M that accepts x.
- If  $x \notin L$ , then all computation paths of M reject x.

An equivalent point of view

## The Complexity Class NP

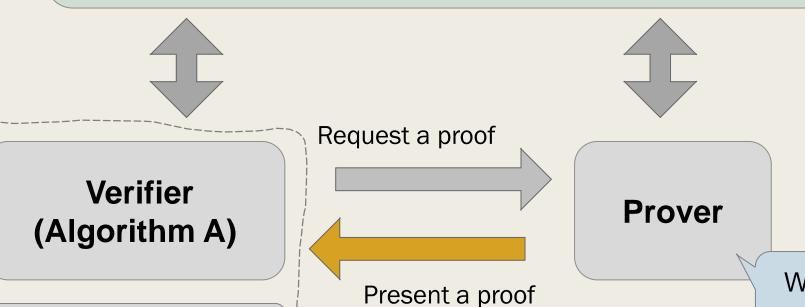
- A problem Π is in NP,
   if there is a *polynomial-time algorithm A* such that
   for any instance *I* of Π,
  - If I is a "Yes"-instance, then there is a **proof**  $\pi \in \{0,1\}^{poly(n)}$  such that A accepts on  $(I,\pi)$ .
  - If *I* is a "No"-instance, then *A* rejects  $(I, \pi)$  for all  $\pi \in \{0,1\}^{poly(n)}$ .

An equivalent point of view

## The Complexity Class NP

- A problem Π is in NP if there is a <u>proof system</u>
  for its <u>yes answers</u> to be verified efficiently in polynomial-time.
  - (Completeness)
     For each "yes"-instance, there is a proof that leads to accept.
  - (Soundness)
     For each "no"-instance, no proof leads to accepts.

Assume a *valid proof system* for  $\Pi$  that can be <u>efficiently verified</u> by an **algorithm** A.



With <u>limited</u> computation power

With <u>unlimited</u> computation power

The classic point of view

## The Complexity Class P

A language L is in P if there is a deterministic Turing machine M that decides it in polynomial-time.

For any string x,

- If  $x \in L$ , then M accepts x in polynomial-time.
- If  $x \notin L$ , then M rejects x in polynomial-time.

A *Turing machine* is actually an *algorithm*, so...

The classic point of view

## The Complexity Class P

A problem Π is in P
 if there is a polynomial-time algorithm A that decides it.

For any instance *I*,

A answers "Yes" if I is a "Yes"-instance, and
 "No" if I is a "No"-instance.

The complexity class P consists of
 problems that can be solved efficiently in polynomial-time.

## The Complexity Classes P vs NP

- From the proof-verifying perspective,
  - Problems in P are those, whose proof can be computed (composed) efficiently in polynomial-time.
- Obviously,  $P \subseteq NP$ .
- Whether or not  $NP \subseteq P$  is <u>a major open problem</u> in CS.
  - Is writing proofs as easy as verifying them?

# Probabilistically Checkable Proofs (PCP)

## The Complexity Class PCP(r(n),q(n))

- A language L is in PCP(r(n),q(n)) if there is a polynomial-time randomized algorithm V such that on any input string  $x \in \{0,1\}^n$ ,
  - (Efficiency)  $V \text{ uses } O\big(r(n)\big) \text{ random bits,}$   $\text{makes } O\big(q(n)\big) \text{ queries to a given proof } \pi \in \{\mathbf{0},\mathbf{1}\}^*, \text{ and accepts / rejects.}$

Assume a *valid proof system* for  $\Pi$  that can be <u>efficiently verified</u> by a randomized **algorithm** A.



Verifier (Algorithm A)

With *limited* computation power

**Query** a part of the proof

Present the part queried



With <u>unlimited</u> computation power

**Prover** 



Possesses a proof that can be very, very, very long.

## The Complexity Class PCP(r(n),q(n))

- A language L is in PCP(r(n),q(n))

  if there is a polynomial-time randomized algorithm V such that on any input string  $x \in \{0,1\}^n$ ,
  - (Completeness)

    If  $x \in L$ , then there exists a proof  $\pi \in \{0,1\}^*$  such that  $\Pr[V^{\pi}(x) \text{ accepts }] = 1$ .
  - (Soundness) If  $x \notin L$ , then for any  $\pi \in \{0,1\}^*$ ,  $\Pr[V^{\pi}(x) \text{ accepts }] \leq 1/2$ .

■ The PCP theorem says that,

$$NP = PCP(\log n, 1).$$

Every language in NP <u>has a proof system</u> that can be verified probabilistically using  $O(\log n)$  random bits and O(1) queries to the proof.

- The PCP theorem has several equivalent formulations.
  - Probabilistically checkable proofs,
    - Graph version,
    - Error-correcting code version, etc.

#### **Definition.** (q-CSP)

An instance of q-CSP consist of a set of alphabet  $\Sigma$ , a set of variables  $X = \{X_1, ..., X_n\}$  with  $X_i \in \Sigma$ , and a set of constraints  $\phi_1, ..., \phi_m$ , where  $\phi_i : X \to \{0,1\}$  depends on at most q variables.

The <u>value</u> of the instance is the <u>maximum fraction</u> of the constraints that can be satisfied by any assignment.

For example, vertex cover is a 2-CSP problem.

- There exist  $q \in \mathbb{N}$  and  $|\Sigma| > 1$  such that, given a q-CSP instance I over alphabet  $\Sigma$ , it is NP-hard to distinguish between the two cases:
  - val(I) = 1, or
  - val(I) < 1/2.

■ Then, the ratio of the gap corresponds to the hardness of approximating the q-CSP problem.

- Definition. (ρ-Gap q-CSP)
   Given an instance of q-CSP problem, distinguish between the following two cases:
  - val(I) = 1, or
  - $val(I) < \rho$ .
- There exists  $q \in \mathbb{N}$  and  $\rho \in (0,1)$  such that  $\rho$ -Gap q-CSP is NP-hard.

#### ■ **Definition.** (Label Cover)

An instance of label cover consist of  $(G = (V_1, V_2, E), \Sigma, \Pi)$ , where

- G is a bipartite graph.
- For any edge  $e \in E$ , there is a constraint  $\Pi_e: \Sigma \to \Sigma$ .

A labelling of the vertices  $\sigma: V \to \Sigma$  is said to satisfy an edge e = (u, v) with  $u \in V_1, v \in V_2$  if and only if

$$\Pi_e(\sigma(u)) = \sigma(v).$$

The <u>value</u> of the instance is the <u>maximum fraction</u> of edges that can be satisfied by any labelling.

■ **Definition.** (GapLabelCover<sub>1, $\epsilon$ </sub>( $\Sigma$ ))

Given an instance of *I* of Label Cover, distinguish between the following two cases:

- val(I) = 1, or
- $val(I) < \epsilon$ .

■ For any  $\epsilon > 0$ , there exists a constant  $|\Sigma|$  such that  $\text{GapLabelCover}_{1,\epsilon}(\Sigma)$  is NP-hard.

## Equivalent Views of PCP Theorem

■ We have defined the language class PCP(r(n), q(n)).

Theorem 1. (PCP Theorem, proof verifying view)

$$NP = PCP(O(\log n), O(1)).$$

#### Theorem 2. (PCP Theorem, hardness of approximation view)

There exists  $\rho$  < 1 such that,

for every language  $L \in NP$ , there is a polynomial-time mapping

$$f: \{0,1\}^* \mapsto 3CNFs$$

such that

$$x \in L \Rightarrow \operatorname{val}(f(x)) = 1$$

$$x \notin L \Rightarrow \operatorname{val}(f(x)) < \rho$$
.

We have defined the gap version of CSP problems.

Theorem 2. (PCP Theorem, CSP view)

There exists  $q \in \mathbb{N}$ ,  $\rho \in (0,1)$  such that  $\rho$ GAPqCSP is NP-hard.

#### Theorem $1 \Longrightarrow \text{Theorem } 3$

- Suppose that  $NP = PCP(O(\log n), O(1))$ .
- It suffices to construct a  $\rho$ GAPqCSP instance from a PCP verifier V of an NP language, say, 3-SAT.
  - Formulate the execution of V as a CSP constraint.
  - V uses O(log n) random bits.
     So, at most poly(n) different constraints.
  - V makes q = O(1) random bits. Each constraint has arity q.

#### Theorem $1 \Longrightarrow \text{Theorem } 3$

- It suffices to construct a  $\rho$ GAPqCSP instance from a PCP verifier V of an NP language, say, 3-SAT.
  - Number of variables =  $q \cdot poly(n) = poly(n)$ .
  - Hence, the CSP instance has polynomial size.
  - The instance has completeness 1 and soundness  $\rho = 1/2$ .
- Since 3-SAT is NP-hard, the gap instance is NP-hard to decide.

#### Theorem $3 \Rightarrow$ Theorem 1

- It suffices to construct a PCP verifier for  $\rho$ GAPqCSP.
  - The verifier expects the proof to be the assignment of the variables.
  - Pick a constant  $c \ge 1$  such that  $\rho^c \le 1/2$ .
  - Pick c random constraints and test them.
  - Number of random bits =  $c \cdot \log m$ . Number of queries = cq = O(1).
  - The verifier has completeness 1 and soundness 1/2.

## Mapping of Concepts between Different Views

Proof verifying view

<u>CSP view</u> (hardness of approx.)

PCP verifier V



CSP instance  $\phi$ 

**Execution of Verifier** 



**CSP** constraint

Probability that *V* accepts



Value of  $\phi$ 

Number of random bits r



Logarithm of number of constraints  $\log m$ 

#### Proof verifying view

<u>CSP view</u> (hardness of approx.)

Length of proof (to be accessed)



Number of variables

PCP proof  $\pi$ 



Assignment to variables

Number of queries q



Arity of constraints *q* 

Soundness parameter (usually 1/2)



Maximum value of any No instance

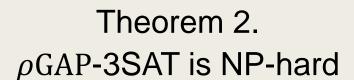
#### Proof verifying view

<u>CSP view</u> (hardness of approx.)

Theorem 1.  $NP = PCP(O(\log n), O(1))$ 



Theorem 3.  $\rho$ GAPqCSP is NP-hard



Corollary.  $(\rho - \epsilon)$ -approximation for Max-3SAT is NP-hard

## Theorem $3 \Longrightarrow$ Theorem 2

- Suppose that  $\rho$ GAP-3SAT is NP-hard.
- 3SAT is a qCSP problem with q = 3.
  - An algorithm that decides  $\rho$ GAPqCSP can be used to decide  $\rho$ GAP-3SAT.
- Hence,  $\rho$ GAPqCSP must also be NP-hard to decide.

#### Theorem $2 \Longrightarrow$ Theorem 3

- Now suppose that  $\rho$ GAPqCSP is NP-hard.
- Given an instance of  $\rho$ GAPqCSP, we construct an instance of  $\rho'$ GAP-3SAT with  $\rho' = \rho/(q2^q)$ .
  - Then,  $\rho'$ GAP-3SAT must be NP-hard to decide.

#### Theorem $2 \Longrightarrow$ Theorem 3

- First, each CSP constraint, say,  $\phi_i = \phi_i(y_1, y_2, ..., y_q)$ , can be transformed to an equivalent q-CNF with at most  $2^q$  clauses.
  - Collect all configurations of  $y_1, y_2, ..., y_q$  that make  $\phi_i$  false.
  - This corresponds to a q-DNF with at most  $2^q$  clauses.
  - Taking negation, we get a q-CNF as claimed.

## Theorem $2 \Rightarrow$ Theorem 3

- Next, we can apply the Cook-Levin technique to transform the *q*-CNF into an equivalent 3-CNF.
- Repeat the following two steps until we have a 3-CNF.
  - Pick a clause with size at least 4, say,  $y_1 \vee y_2 \vee \phi'$ , where  $|\phi'| \geq 2$ .

The number of literals is decreased by 1.

Add a new variable z and replace the clause with

$$(y_1 \lor y_2 \lor z) \land (\bar{z} \lor \phi')$$
.

The number of variables and clauses are increased by 1.

- Repeat the following two steps until we have a 3-CNF.
  - Pick a clause with size at least 4, say,  $\phi' = y_1 \vee y_2 \vee \phi''$ , where  $|\phi'| \geq 2$ .

The number of literals is decreased by 1.

- Introduce a new variable z and replace  $\phi'$  with

$$(y_1 \lor y_2 \lor z) \land (\bar{z} \lor \phi'')$$
.

If  $\phi'$  is satisfied, then there exists  $z \in \{0,1\}$  such that  $(y_1 \lor y_2 \lor z) \land (\bar{z} \lor \phi'')$  is satisfied.

The number of variables and clauses are increased by 1.

If  $\phi'$  is not satisfied, then no  $z \in \{0,1\}$  can simultaneously satisfy  $(y_1 \lor y_2 \lor z)$  and  $(\bar{z} \lor \phi'')$ .

#### Theorem $2 \Longrightarrow$ Theorem 3

- Next, we can apply the Cook-Levin technique to transform the *q*-CNF into an equivalent 3-CNF.
- From the q-CNF with n variables and  $2^q m$  clauses, we obtain a 3-CNF with n+qm variables and  $q2^q m$  clauses.
  - The completeness is 1.
  - Each unsatisfied clause in *q*-CNF results in at least one unsatisfied clause in 3-CNF.
  - The soundness is  $\rho' = \rho/(q2^q)$ .