

Probabilistically Checkable Proofs (PCP) and Hardness of Approximations

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Outline

- 1 Introduction on Approximation Algorithms
- 2 PCP & PCP Theorem ($NP = PCP(O(\log n), O(1))$)
- 3 PCP Theorem \Leftrightarrow MAX-3SAT is hard to Approximate
- 4 From Label-Cover to MAX-Clique

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Approximation Algorithms

Π a maximization Problem

c -Approximation for Π

$1 < c$ constant or depends on input length

- deterministic polynomial-time algorithm \mathcal{A}
- for any input I , \mathcal{A} returns a solution with value at least $OPT(I)/c$.

Example: MAX-3SAT

input: a 3CNF formula Φ

output: maximum (over all assignments A) number of clauses of Φ satisfied by A

remark: MAX-3SAT is NP-hard

Algorithm: Random assignment

$\Rightarrow \frac{7}{8}$ -approximation for MAX-3SAT

each clause is satisfied with probability $\frac{7}{8}$

Application of PCP Theorem : above algorithm (derandomized) is optimal unless $P = NP$.

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Hierarchy of NP-hard Problems

Fully Polynomial Time Approximation Scheme (FPTAS)

For any ϵ , $\exists (1 + \epsilon)$ -approximation algorithm, polynomial in both n and $1/\epsilon$
 e.g., Knapsack

Polynomial Time Approximation Scheme (PTAS)

For any ϵ , $\exists (1 + \epsilon)$ -approximation algorithm, polynomial in n (typically $n^{f(\epsilon)}$)
 but no FPTAS (unless $P=NP$)
 e.g., Euclidean TSP

$\Theta(1)$ -approximation but no PTAS (unless $P=NP$)

e.g., Vertex Cover, MAX-3SAT

$O(\log n)$ -approximation but no constant approximation (unless $P=NP$)

e.g., Set Cover

only $n^{\Theta(1)}$ -approximation, no $n^{1-\epsilon}$ -approximation for any $\epsilon > 0$ (unless $P=NP$)

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Reminder: Definition of NP

$L \in NP$ iff \exists a deterministic Verifier V s.t.

for any input (word) w

- V polynomial-time in $|w|$
- and
 - if $w \in L$ then \exists proof (certificate) x with $V(w, x) = 1$
 - if $w \notin L$ then \forall proof x , $V(w, x) = 0$

examples:

3-SAT: proof= assignment of variables

3-Coloriage: proof= Labelling of vertices

Verifier checks (1) that certificate is consistent (2) that the answer is correct

Definition of the class PCP(r, q)

$L \in \text{PCP}(r, q)$ iff \exists a randomized Verifier V s.t.

for any input w

- V polynomial-time in $|w|$, uses a string ρ of $O(r(|w|))$ bits of randomness and only checks $O(q(|w|))$ bits of the proof, and
 - if $w \in L$ then \exists proof x with $\Pr_{\rho}(V(w, x, \rho) = 1) = 1$ (**completeness**).
 - if $w \notin L$ then \forall proof x , $\Pr_{\rho}(V(w, x, \rho) = 1) \leq 1/2$ (**soundness**).

proof x has size $\leq q(|w|) \cdot 2^{r(|w|)}$ and only $O(q(|w|))$ bits of x are checked.

If the proof x is correct, then w is accepted with proba 1 (**completeness**).
 Otherwise (if proof x is a fake), w is rejected with probability at least $1/2$ (**soundness**).

tradeoff between randomness (r) and number of bits (q) checked

Relationships between PCP and NP

$NP \subseteq PCP(0, poly(n))$

idea: NP-proof has polynomial size and the PCP-Verifier can check it completely

$PCP(O(\log n), O(1)) \subseteq NP$

idea: NP-Verifier: apply PCP-Verifier for each of the $2^{O(\log n)}$ possible random strings
i.e., “check the full proof” in polynomial-time

$NP \subseteq PCP(n^{O(1)}, O(1))$

idea: Increase the size of NP-proof s.t. if it is a fake, the mistakes are “everywhere”

PCP Theorem: $NP \subseteq PCP(O(\log n), O(1))$ Corollary: $NP = PCP(O(\log n), O(1))$

Initial proof: [Arora,Safra FOCS'92] [Arora,Lund,Motwani,Sudan,Szegedy FOCS'92]
[Arora FOCS'95]

improvements: [Raz STOC'95] [Hastad JACM'01]

“easier” proof: [Dinur STOC'06, JACM'07]

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PCP Theorem

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

[Arora et al.'92]

Consequence: for any $L \in NP$, \exists a Verifier V s.t.

for any input w of size n

- V polynomial-time in $|w|$, uses a string ρ of $O(\log n)$ bits of randomness and only checks $O(1)$ bits of the proof, and
 - if $w \in L$ then \exists proof x with $Pr_{\rho}(V(w, x, \rho) = 1) = 1$ (completeness).
 - if $w \notin L$ then \forall proof x , $Pr_{\rho}(V(w, x, \rho) = 1) \leq 1/2$ (soundness).

Remark 1: $PCP(O(\log n), O(1))$ is stable by polynomial-time reductions.

i.e., if $L' \prec_{poly} L$ and $L \in PCP(O(\log n), O(1))$ then $L' \in PCP(O(\log n), O(1))$

$L' \prec_{poly} L$ means \exists a poly-time reduction from L to L' (i.e., L not "easier" than L').

Remark 2: V may have extra/restricted properties [Arora et al.'92, Raz'95, Hastad'01]

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MAX-3SAT is hard to Approximate (1/2)

Goal: prove that $\exists c_0$ s.t. $\forall \epsilon$, there is no $(c_0 - \epsilon)$ -approx of MAX-3SAT unless $P = NP$

MAX-3SAT

input: a 3CNF formula Φ

output: maximum (over all assignments A) number of clauses of Φ satisfied by A

GAP-3SAT_s

 $0 < s < 1$

input: a 3CNF formula Φ with m clauses

output:

- YES if Φ is satisfiable
- NO if at most $s \cdot m$ clauses are satisfiable (for all assignments)
- anything otherwise

Lemma

If $\exists s < 1$ s.t. GAP-3SAT_s is NP-hard

then there is no s -approx of MAX-3SAT unless $P = NP$

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PCP \Leftrightarrow Hardness of Approximation

$NP \subseteq PCP(O(\log n), O(1)) \Leftrightarrow \exists s < 1$ s.t. GAP-3SAT_s is NP-hard

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PCP \Leftrightarrow Hardness of Approximation

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\Leftarrow roughly, prove that GAP-3SAT_s $\in PCP(O(\log n), O(1))$

For any $L \in NP$ there is Φ polynomial-time s.t.

- for any word w , $\Phi(w)$ is an instance of GAP-3SAT_s
- if $w \in L$ then $\Phi(w)$ satisfiable; else at most a fraction s of the clauses can be satisfied

Verifier:

- 1 Compute $\Phi(w)$, m its number of clauses
- 2 use $\lceil \log m \rceil$ random bits to choose a clause
- 3 read the value of the literals in this clause
- 4 answer 1 if the clause is satisfied, 0 otherwise

If $w \notin L$ (i.e., at most $s \cdot m$ clauses can be satisfied), it is detected with proba $\geq 1 - s$.
Repeat constant number of times to achieve the desired probability

MAX-3SAT is hard to Approximate (2/2)

PCP \Leftrightarrow Hardness of Approximation

$NP \subseteq PCP(O(\log n), O(1)) \Leftrightarrow \exists s < 1$ s.t. GAP-3SAT_s is NP-hard

\Rightarrow

reduce $L \in NP$ to GAP-kSAT _{$1 - \frac{1}{2^{k+1}}$}

L has a Verifier V using $O(\log n)$ random bits and read $k = O(1)$ bits in the proof.
 given a word w and a random string ρ , $V(w, \rho, x) \in \{0, 1\}$ reads k bits of the proof x
 $f_{w, \rho} : x \rightarrow V(w, \rho, x)$ expressed as k -SAT formula with $\leq 2^k$ clauses.

$$\text{Let } \Phi_w = \bigwedge_{\rho \in \{0,1\}^{O(\log n)}} f_{w, \rho}.$$

if $w \in L$ then Φ_w satisfied by x
 otherwise at most a fraction $1 - \frac{1}{2^{k+1}}$ of clauses can be satisfied.

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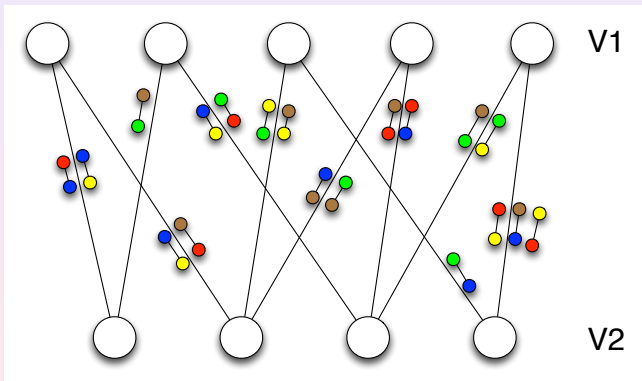
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LABEL COVER

(Maximization version)

regular bipartite graph $G = (V_1 \cup V_2, E)$ and, $\forall e \in E$, partial function

$$\Pi_e : [1, M] \rightarrow [1, M]$$



Let $c : V \rightarrow [1, M]$ be a coloring of V .

an edge $e = \{u, v\} \in E$, $u \in V_1, v \in V_2$, is covered if $\Pi_e(c(u)) = c(v)$.

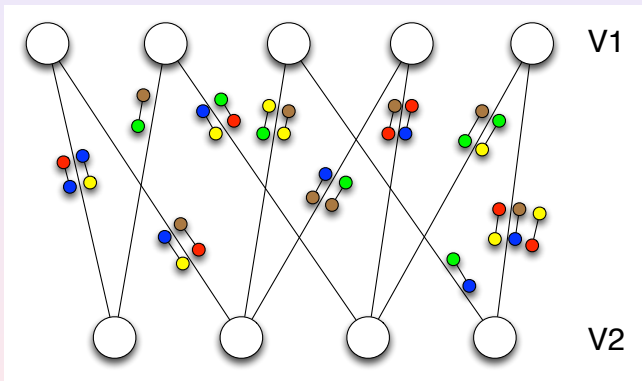
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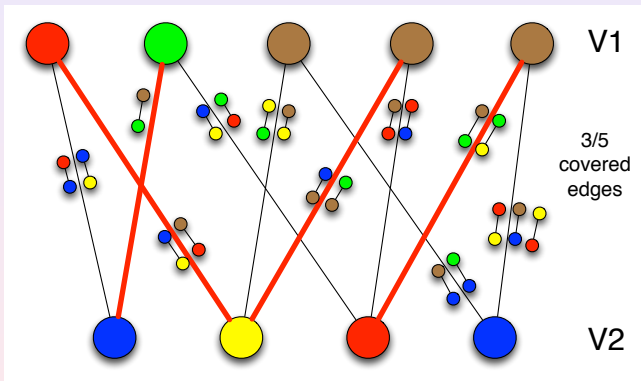
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Goal: Find a coloring maximizing the fraction of covered edges.

LABEL COVER hard to approximate

\exists reduction f from 3SAT to LABEL COVER s.t.

for any instance I of 3SAT and instance $f(I)$ of LABEL COVER

if X clauses of I can be satisfied, $\geq X$ edges of $f(I)$ can be covered

if Y edges of $f(I)$ can be covered, $\geq Y/3$ clauses of I can be satisfied

Gap-Preserving Reduction

Corollary

$\exists \epsilon$ s.t. finding an $(1 + \epsilon)$ -approximation to Label Cover is NP-hard.

LABEL COVER *very* hard to approximate

Theorem

$\forall 0 < \epsilon < 1/2$ and $c(n) \geq 2^{\log^{0.5-\epsilon} n}$, and there is a reduction f in time $n^{\text{poly}(\log n)}$, from any NP problem Π to Label Cover such that

- for any $w \in \Pi$, $f(w)$ has value 1
- for any $w \notin \Pi$, $f(w)$ has value at most $1/c(n)$.

Corollary: if $\exists 0 < \epsilon < 1/2$ and a $2^{\log^{0.5-\epsilon} n}$ -approximation algorithm for Label Cover, then any problem in NP can be solved in time $n^{\text{poly}(\log n)}$.

Conjecture: Exponential Time Hypothesis (ETH)

There is no sub-exponential-time algorithm to solve 3-SAT unless $P = NP$

Corollary 1: Assuming ETH

$\forall \epsilon$, there is no $2^{\log^{0.5-\epsilon} n}$ -approximation algorithm for Label Cover unless $P = NP$

Corollary 2: Assuming ETH

$\forall \epsilon$, there is no $2^{\log^{0.5-\epsilon} n}$ -approximation algorithm for Max Clique unless $P = NP$

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Proof of Cor. 2: “Max Clique hard to approx.”

Gap preserving reduction from Label Cover

Reduce any instance of Label Cover to instance of Max Clique s.t. both optima coincide

Let $G = (V_1 \cup V_2, E)$ and, $\forall e \in E, \Pi_e : [1, M] \rightarrow [1, M]$

- for any $e \in E$ and any $a, b \in [1, M]$ s.t., $\Pi_e(a) = b$
add a vertex (e, a, b)
- 2 nodes are adjacent if they are consistent
e.g., (uv, a, b) and (uw, a', b') are adjacent iff $a = a'$

Proof of Theorem

$L \in \text{RPCP}(r, s, \rho)$ iff \exists a **restricted** Verifier V s.t.

- V polynomial-time in $|w|$, uses a string ρ of $O(r(|w|))$ bits of randomness
- the proof uses an alphabet with $2^{O(s(|w|))}$ symbols
- the proof consists of 2 Tables T_1, T_2
- V chooses 1 location (chosen uniformly at random) in each table, and read only the corresponding 2 symbols a_1 and a_2
- V confirms that T_1 coherent with T_2 : for any a_1 at most one symbol a_2 makes V accepting
 - if $w \in L$ then \exists proof x with $\Pr_{\rho}(V(w, x, \rho) = 1) = 1$ (**completeness**).
 - if $w \notin L$ then \forall proof x , $\Pr_{\rho}(V(w, x, \rho) = 1) \leq 2^{-\rho(|w|)}$ (**soundness**).

Theorem [Feige, Lovász'92]

improved by Raz'95

For any integer $k \geq 2$,

$$NP \subseteq \text{RPCP}(\log^{2k+2} n, \log^{k+2} n, \log^k n)$$

For any $L \in NP$, use a restricted PCP-Verifier of it to build an instance of Label Cover.

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References

- Sanjeev Arora, Probabilistic Checking of Proofs and Hardness of Approximation Problems, Ph.D. thesis, 1994
- Sanjeev Arora, Carsten Lund, Rajeev Motwani, Madhu Sudan, Mario Szegedy: Proof Verification and the Hardness of Approximation Problems. J. ACM 45(3): 501-555 (1998)
- Irit Dinur: The PCP theorem by gap amplification. J. ACM 54(3): 12 (2007)
- Johan Hastad: Some optimal inapproximability results. J. ACM 48(4): 798-859 (2001)
- Ran Raz: A Parallel Repetition Theorem. SIAM J. Comput. 27(3): 763-803 (1998)

lecture of Nicolas Schabanel on proof PCP theorem (video online)

lecture of Luca Trevisan on PCP (video online)