Lower bounds for CSP refutation by SDP hierarchies

Ryuhei Mori¹ David Witmer²

¹Tokyo Institute of Technology

²Carnegie Mellon University

APPROX + RANDOM at Paris Sep 9, 2016

◆□▶ ◆□▶ ◆三▶ ◆三▶ 三三 のへで

Constraints satisfaction problem (CSP)

k-CSP

- ► An alphabet set [q] = {1, 2, ..., q}.
- Variables $(x_i \in [q])_{i=1,\ldots,n}$.
- Constraints $(C_j : [q]^k \rightarrow \{0, 1\})_{j=1,\dots m}$.

 $F = C_1(x_1, x_2, x_4) \land C_2(x_2, x_6, x_7) \land C_3(x_3, x_5, x_8)$



Random CSP with parameters *n*, *m* and *C*: $[q]^k \rightarrow \{0, 1\}$.

Each constraint is generated independently in the following way.

- Chose a set of k variables from $\binom{n}{k}$ candidates uniformly.
- Apply random permutations on [q] for all of the k variables.

Phase transition of the random k-CSP



Refutation of random k-CSP



 $\label{eq:main result:} \ensuremath{\mathsf{Not}}\xspace$ Not refutable in poly-time by SA+ nor LS+ if $m < n^{\chi(\mathcal{C})-\delta}$

Table of contents

- LP/SDP hierarchies
- Iocally consistent distribution [BGMT 2012]
- Sketch of the proof

MAX-k-CSP

$$\max_{\mathbf{x}\in[q]^n}:\sum_{j=1}^m C_j(\mathbf{x}_j)$$

if m > Optimum of relaxed MAX-k-CSP then

 $m > \text{Optimum of MAX-}k\text{-}\text{CSP} \implies \text{The CSP is UNSAT}$

LP/SDP relexation

Formulation of combinatorial optimization problem as linear program.

$$\max_{\mathbf{x}} : \sum_{j=1}^{m} C_j(\mathbf{x}_a)$$

s.t. : $\mathbf{x} \in [q]^n$

$$\begin{split} \max_{p} : \mathbb{E}_{p} \left[\sum_{j=1}^{m} C_{j}(\mathbf{X}_{a}) \right] \\ \text{s.t.} : p \in \mathcal{P}([q]^{n}) = \text{ The set of distributions on } [q]^{n} \end{split}$$

Then, relax the polytope $\mathcal{P}([q]^n)$.

Sherali-Adams LP hierarchy

The tight polytope

$$\begin{array}{l} \max_{p} : \mathbb{E}_{p} \left[\sum_{j=1}^{m} C_{j}(\mathbf{X}_{a}) \right] \\ \text{s.t.} : p \in \mathcal{P}([q]^{n}) = \ \text{The set of distributions on } [q]^{n} \end{array}$$

The r-round Sherali-Adams relaxation

$$\max_{\substack{(p_{S}:S \subseteq [n], |S| \le r)}} : \sum_{j=1}^{m} \mathbb{E}_{P_{V(C_{j})}} [C_{j}(\mathbf{x}_{j})]$$

s.t. : $p_{S} \in \mathcal{P}([q]^{k}), S \subseteq [n], |S| \le r$
All local distributions are locally-consistent

$$\mathsf{Tight} = \mathsf{SA}(n) \subseteq \mathsf{SA}(n-1) \subseteq \cdots \subseteq \mathsf{SA}(k)$$

Sherali-Adams+ SDP hierarchy

The *r*-round SA + = The *r*-round SA with PSDness condition for Σ

where Σ is variance-covariance matrix, i.e.,

$$\Sigma_{(i,x),(j,y)} := p_{\{i,j\}}(x,y) - p_{\{i\}}(x)p_{\{j\}}(y).$$

$$\max_{\substack{(p_S:S\subseteq [n], |S| \le r)}} : \sum_{j=1}^m \mathbb{E}_{P_{V(C_j)}} [C_j(\mathbf{x}_j)]$$

s.t.: $(p_S)_{S\subseteq [n], |S| \le r} \in SA(r)$
 $\Sigma \succeq 0$

Equivalence of PSDness Lemma (Schur complement)

$$\begin{bmatrix} 1 & p^T \\ p & B \end{bmatrix} \text{ is PSD } \Longleftrightarrow B - pp^T \text{ is PSD.}$$

Proof.

$$\begin{bmatrix} 1 & p^{T} \\ p & B \end{bmatrix} \text{ is PSD}$$

$$\iff \left(\begin{bmatrix} x_{0} x \end{bmatrix} \begin{bmatrix} 1 & p^{T} \\ p & B \end{bmatrix} \begin{bmatrix} x_{0} x \end{bmatrix}^{T} \ge 0 \text{ for any } x_{0} \in \mathbb{R}, x \in \mathbb{R}^{nq} \right)$$

$$\iff \left(x_{0}^{2} + 2\langle p, x \rangle x_{0} + \langle Bx, x \rangle \ge 0 \text{ for any } x_{0} \in \mathbb{R}, x \in \mathbb{R}^{nq} \right)$$

$$\iff \left((x_{0} + \langle p, x \rangle)^{2} - \langle p, x \rangle^{2} + \langle Bx, x \rangle \ge 0 \text{ for any } x_{0} \in \mathbb{R}, x \in \mathbb{R}^{nq} \right)$$

$$\iff \left(-\langle p, x \rangle^{2} + \langle Bx, x \rangle \ge 0 \text{ for any } x \in \mathbb{R}^{nq} \right)$$

$$\iff \left(x(B - pp^{T})x^{T} \text{ for any } x \in \mathbb{R}^{nq} \right)$$

$$\iff B - pp^{T} \text{ is PSD.}$$

11 / 24

Traditional description of SA+

$$\max_{\substack{(p_S:S \subseteq n, |S| \le r)}} : \sum_{j=1}^m \mathbb{E}_{p_j} [C_j(\mathbf{x}_j)]$$

s.t.:
$$(p_S)_{S \subseteq [n], |S| \le r} \in \mathsf{SA}(r)$$
$$\langle \mathbf{v}_{i,a}, \mathbf{v}_{j,b} \rangle = p_{\{i,j\}}(a, b) \quad \forall i \neq j \in [n], a, b \in [q]$$
$$\langle \mathbf{v}_{i,a}, \mathbf{v}_{i,b} \rangle = 0 \qquad \forall i \in [n], a \neq b \in [q]$$
$$\|\mathbf{v}_{i,a}\|^2 = \langle \mathbf{v}_{i,a}, \mathbf{v}_{\emptyset} \rangle = p_{\{i\}}(a) \qquad \forall i \in [n], a \in [q]$$
$$\|\mathbf{v}_{\emptyset}\|^2 = 1$$

$$\exists V, \text{s.t. } V^{T}V = \begin{bmatrix} 1 & p^{T} \\ p & B \end{bmatrix} \iff \begin{bmatrix} 1 & p^{T} \\ p & B \end{bmatrix} \text{ is PSD}$$

Other SDP hierarchies

- Lovász-Schrijver+ (LS+): SA with PSDness condition for conditional variance-covariance matrix [Tulsiani and Worah 2012]
- Lasserre/SOS: SA with PSDness condition for variance-covariance matrix for higher order statistics.

LP/SDP in Computational Complexity

- If poly-size SA cannot refute then any poly-size LP relaxation cannot refute [Chan, Lee, Raghavendra, and Steurer 2013]
- If poly-size SOS cannot refute then any poly-size SDP relaxation cannot refute [Lee, Raghavendra, and Steurer 2015]

Known facts and Main results

- Worst-case lower bound of SA+ for pairwise uniform MAX-k-CSP with linearly many constraints [Benabbas, Georgiou, Magen, and Tulsiani 2012]
- Average-case lower bound of LS+ for pairwise uniform random MAX-k-CSP with linearly many constraints [Tulsiani and Worah 2013]
- Average-case lower bound of SA+ for (t − 1)-wise uniform random MAX-k-CSP with n^{t/2−δ} constraints with small modification [O'Donnell and Witmer 2014]

Theorem ([This Work])

Poly-size SA+/LS+ cannot refute random (t - 1)-wise uniform CSP with $n^{t/2-\delta}$ constraints with high probability.

Summary for LP/SDP hierarchy

All LP/SDP hierarchies can be understood by the idea of local distributions.

All SDP hierarchies have PSDness condition of variance-covariance matrix.

Table of contents

- LP/SDP hierarchies
- Iocally consistent distribution [BGMT 2012]
- Sketch of the proof

The local distribution

Definition ((t-1)-wise uniform constraint)

A constraint $C: [q]^k \to \{0, 1\}$ is said to be (t - 1)-wise uniform if there exists a distribution μ on the support of C such that the marginal distribution μ_T for $T \subseteq [k]$ is uniform distribution on $[q]^{|T|}$ if $|T| \le t - 1$.

Specific choice of local distribution [BGMT 2012].

$$D'_{S}(x_{S}) := \frac{1}{Z_{S}} \prod_{C \in \mathcal{C}(S)} \mu_{C}(x_{\Gamma C})$$
$$D_{S}(x_{S}) := \sum_{x_{\bar{S} \setminus S}} D'_{\bar{S}}(x_{\bar{S}})$$

 \bar{S} : The closure of S.

Local consistency

Theorem ([BGMT 2012] [O'Donnell and Witmer 2014]) For $m = O(n^{t/2-\delta})$ and $r = O(n^{\frac{\delta}{t-2}})$, $(D_S)_{|S| \le r}$ is locally consistent with high probability.

Summary and our goal

Sherali-Adams+ Relaxation.

$$\Sigma_{(i,x),(j,y)} := p_{\{i,j\}}(x,y) - p_{\{i\}}(x)p_{\{j\}}(y).$$

$$\max_{\substack{(p_S:S \subseteq [n], |S| \le r)}} : \sum_{j=1}^m \mathbb{E}_{p_{V(C_j)}} [C_j(\mathbf{x}_j)]$$

s.t.: $(p_S)_{S \subseteq [n], |S| \le r} \in SA(r)$
 $\Sigma \succeq 0$

From [BGMT 2012] and [O'Donnell and Witmer 2014], $(p_S = D_S)_{S \subseteq [n], |S| \le r} \in SA(r).$

• Our goal is to show that Σ for $(D_S)_{S \subseteq [n], |S| \leq r}$ is PSD.

Table of contents

- LP/SDP hierarchies
- Iocally consistent distribution [BGMT 2012]
- Sketch of the proof

Proof strategy

Goal is to show

 Σ for $(D_S)_{S \subseteq [n], |S| \leq r}$ is PSD.

- In fact, Σ is not diagonal.
- We will not find the vectors explicitly.
- We will show the PSDness in more implicit way.

Correlation graph

• Vertex set: V = [n].

► Edge set: $E = \{(i,j) \mid \exists (x,y) \in [q]^2 \text{ s.t. } \Sigma_{(i,x),(j,y)} \neq 0\}.$

Theorem ([This Work])

Every connected components of the correlation graph for D has size O(1).

By rearranging the indices of Σ , we obtain

$$\begin{bmatrix} \Sigma_1 & 0 & 0 & 0 \\ 0 & \Sigma_2 & 0 & 0 \\ 0 & 0 & \Sigma_3 & 0 \\ 0 & 0 & 0 & \textit{I} \end{bmatrix}$$

Since each block is PSD, the whole matrix Σ is PSD.

Summary

Main result:

Poly-size SA+/LS+ cannot refute random (t-1)-wise uniform CSP with $n^{t/2-\delta}$ constraints with high probability.

Proof technique:

- We regard the PSDness condition as $\Sigma \succeq 0$.
- We show that the connected components in the correlation graph are small.
- This immediately implies the PSDness of Σ.