

The Approximability of constraint Satisfaction Problems

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Defn] (α, β) -approximability CSP(Γ): if $\text{Opt}(\text{cl}) \geq \beta$
alg's assignment has value $\geq \alpha$.

Basic LP] $\max_{\text{cl}} \text{LP Val}(\mathcal{L}) := \text{avg}_{C=(R,D) \in \text{cl}} \{ \Pr [L(S) \in R] \}$
 $L \sim \lambda_C$
 s.t. λ_C is prob. dist. on local assignments
 $S \rightarrow D \quad \forall C=(R,S), \mu_v$ is prob. dist. on $D \quad \forall v \in V$,
 (*) $\Pr [L(v) = l] = \mu_v[l] \quad \forall C \quad \forall v \in S \quad \forall l \in D$
 $L \sim \lambda_C$

Thm] Given cl for max-(k)Sat
 • Solve LP optimally, get $\mathcal{L} = ((\mu_v), (\lambda_C))$
 • Output $F: V \rightarrow \{0,1\}$: $F(v) \sim \mu_v$ independently
 $\forall v \in V$
 Then this $((1-\epsilon)\beta, \beta)$ -approximates $\forall \beta$.

eg] $C = x \vee y \vee z$. $\Pr_F [F \text{ satisfies } C] = 1 - \mu_x[0] - \mu_y[0] - \mu_z[0]$

Generally say $C = (R, S)$
 $b_C: S \rightarrow \{0,1\}$ for the bad (forbidden) assign.
 for C .

$$\Pr_F [F \text{ satisfies } C] = 1 - \prod_{v \in S} \mu_v [b_C(v)] \quad (*)$$

$$P_C := \Pr_{L \sim \lambda_C} [L(S) \text{ satisfies } R] \leq \sum_{v \in S} \Pr_{L \sim \lambda_C} [L(v) \neq b_C(v)]$$

$$= \sum_{v \in S} (1 - \mu_v [b_C(v)]) \quad \text{"by (*)"}.$$

$$= |S| \text{AM}_{v \in S} [1 - \mu_v [b_C(v)]]$$

$$(*) = 1 - \text{GM}_{v \in S} \{ \mu_v [b_C(v)] \}^{|S|}$$

$$\geq 1 - \text{AM}_{v \in S} \{ \mu_v [b_C(v)] \}^{|S|}$$

$$= 1 - (1 - \text{AM}_{v \in S} [1 - \mu_v [b_C(v)]])^{|S|}$$

$$\geq 1 - (1 - P_C / |S|)^{|S|} \geq (1 - \frac{1}{e}) P_C$$

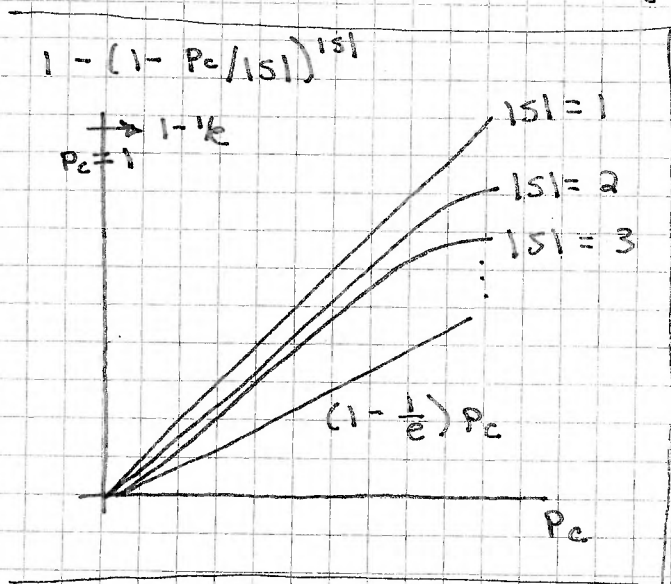
$$\therefore \Pr [F \text{ satisfies } C] \geq \left(1 - \frac{1}{e}\right) p_c$$

$$\text{LP Val}(\mathcal{L}) = \text{avg} \{ p_c \}$$

$$\begin{aligned} \therefore \mathbb{E}_F [\text{Val}(F)] &\geq \text{avg} \left\{ \left(1 - \frac{1}{e}\right) p_c \right\} \\ &\geq \left(1 - \frac{1}{e}\right) \text{LP Val}(\mathcal{L}) \end{aligned}$$

□

Cor1 For Max-2 Sat, $\left(\frac{3}{4} \beta, \beta\right)$ - approx.



Cor1 For Max-2 Sat $\left(\frac{3}{4} \beta, \beta\right)$ - approx. with trivial random alg.

$\left(\frac{7}{8} + \epsilon, 1\right)$ - hard for E3-Sat.

Cor1 For \mathcal{L} , instance of Max-Sat
 $\text{Opt}(\mathcal{L}) \leq \text{LP Opt}(\mathcal{L}) \leq \frac{4}{3} \text{Opt}(\mathcal{L})$.

Defn (α, β) - gap instance for CSP(Γ) is \mathcal{L} s.t.
 $\text{Opt}(\mathcal{L}) \leq \alpha$, but $\text{LP Opt}(\mathcal{L}) \geq \beta$

⊕ If such \mathcal{L} exists basic LP is not (α, β) -dist. alg.

Prop This is a $(\frac{3}{4}, 1)$ -gap instance for E2-Sat
 $\bullet x_1 y_1, \bar{x}_1 y_1, x_2 y_2, \bar{x}_2 y_2$

check: $\text{opt} \leq \frac{3}{4}$ \checkmark

LP Sol: $\mu_x: \frac{1}{2} / \frac{1}{2}$ on $0, 1$ μ_y same

$\lambda_{c_2}: \frac{1}{2}$ prob. on $(0, 0), \frac{1}{2}$ prob. on $(1, 1)$

Similarly for c_1, c_3, c_4
 Get $\text{LP opt} = 1$.

Max-Cut $D = \{0, 1\}$ $\Gamma = \{ \neq \}$
 $\forall G$ (G a graph) $\text{LP opt}(G) = 1$

$\forall v \in V$ $\mu_v: \frac{1}{2}$ on $0, \frac{1}{2}$ on 1

$\forall e \in E$ $\lambda_e: \frac{1}{2}$ on $(0, 1), \frac{1}{2}$ on $(1, 0)$

(GW)

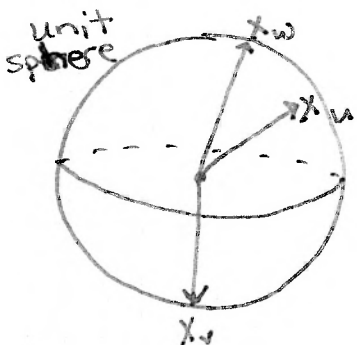
Semidefinite Programming (SDP) Relaxation for instance $G = (V, E)$ of Max-Cut

Maximize $\text{SDP Val}(\mathcal{X}) = \text{avg}_{(u,v) \in E} \left\{ \frac{1}{2} - \frac{1}{2} \langle X_u, X_v \rangle \right\}$

s.t. $\mathcal{X} = (X_v)_{v \in V}$ a collection of n -dim. vectors

with $\langle X_u, X_u \rangle = \|X_u\|_{\mathbb{R}^n}^2 = 1$ ($n = \#$ vertices)

$X_v \in \mathbb{R}^n$



Prop $\text{Opt}(G) \leq \text{SDP Opt}(G) \leq \text{LP opt}(G)$

PF Let $F: V \rightarrow \{0, 1\}$ be optimal cut.

\mathcal{X} : Define $X_v = (F(v), \underbrace{0, 0, \dots, 0}_{n-1})$

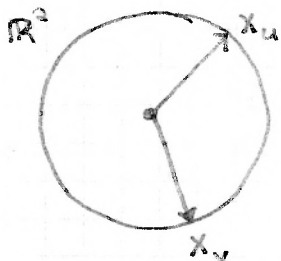
$\text{SDP Val}(\mathcal{X}) = \text{avg}_{(u,v) \in E} \left(\frac{1}{2} - \frac{1}{2} F(u)F(v) \right)$

$= \text{avg}_{(u,v) \in E} \begin{cases} 0 & \text{if } F(u) = F(v) \\ 1 & \text{if } F(u) \neq F(v) \end{cases}$

$= \text{Val}_G(F)$

GW algorithm: Pick a uniformly random hyperplane through 0 (the origin). (Pick its normal \vec{z} from a rotationally symmetric distribution.) Define $F(v) = \text{sgn}(\langle \vec{z}, X_v \rangle)$.

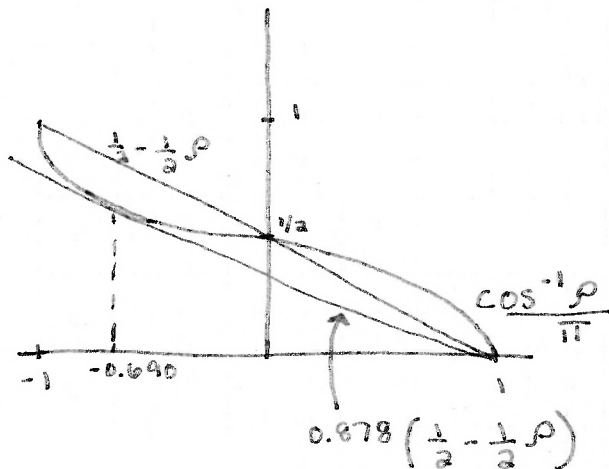
Fix $e = (u, v)$? $\Pr_{\vec{z}} [\text{alg "cuts" } e]$



$$= \frac{\text{angle}(X_u, X_v)}{\pi}$$

$$= \frac{\cos^{-1}(\langle X_u, X_v \rangle)}{\pi} \geq 0.878 \left(\frac{1}{2} - \frac{1}{2} \langle X_u, X_v \rangle \right)$$

$$\rho_{uv} := \langle X_u, X_v \rangle$$



$$E_{\text{GW}} [\text{Val}_G(F)] = \text{avg}_{(u,v) \in E} \{ \Pr[(u,v) \text{ cut}] \}$$

$$\geq \text{avg}_{(u,v)} \left\{ 0.878 \left(\frac{1}{2} - \frac{1}{2} \langle X_u, X_v \rangle \right) \right\}$$

$$= 0.878 \text{SDP Val}(G)$$

Thm) $E[\text{Val}] \geq 0.878 \text{SDP Opt}(G) \geq 0.878 \text{Opt}(G)$

$\therefore (0.878\beta, \beta)$ - approx $\forall \beta$

~~$(1 - 0.5\epsilon, 1 - \epsilon)$~~
 $(1 - 0.5\epsilon, 1 - \epsilon)$

Thm) $\forall \beta \geq 0.845 \exists$ a graph G_n with

$$\text{SDP Opt}(G_n) \geq \beta - o_n(1)$$

$$\text{and } \text{Opt}(G_n) \leq 0.878\beta + o_n(1)$$

Let G be graph whose vertices

$$V = \{x \in \mathbb{R}^n \mid \|x\|_2 = 1\}$$

$$E = \{(x, y) \mid \langle x, y \rangle \leq \frac{1}{2} - \frac{1}{2}\beta\}$$

See notes posted for check.

Univer

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