

More SPPs:

Today: examples where QP formulation isn't obvious or doesn't exist

Correlation Clustering:

Input: - $G = (V, E)$
 - $w^-: E \rightarrow \mathbb{R}^+$
 - $w^+: E \rightarrow \mathbb{R}^+$

Feasible solution: Partition $\mathcal{S}$ of V

$\delta(\mathcal{S})$ = edges go between parts of $\mathcal{S}$
 $E(\mathcal{S})$ = edges where both endpoints in same part of $\mathcal{S}$

Objective: $\max W(\mathcal{S}) = \max \sum_{e \in E(\mathcal{S})} w^+(e) + \sum_{e \in \delta(\mathcal{S})} w^-(e)$

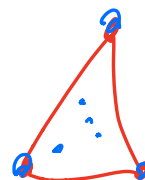
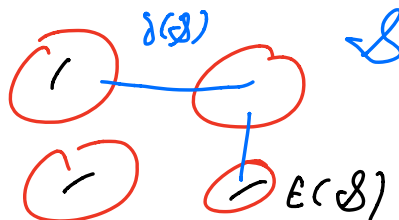
Simple $1/2$ -approximation: return better of $\mathcal{S}_1 = \{V\}$,
 $\mathcal{S}_2 = \{\{i\} : i \in V\}$

value of $\mathcal{S}_1 = W(\mathcal{S}_1) = \sum_{e \in E} w^+(e)$

value of $\mathcal{S}_2 = W(\mathcal{S}_2) = \sum_{e \in E} w^-(e)$

Every partition $\mathcal{S}$ has $W(\mathcal{S}) \leq \sum_{e \in E} w^+(e) + \sum_{e \in E} w^-(e)$

$\Rightarrow \max(W(\mathcal{S}_1), W(\mathcal{S}_2)) \geq \frac{1}{2} \cdot \text{OPT}$



Vector Programming Formulation :

$$\max \sum_{(i,j) \in E} (w^+(i,j) \langle v_i, v_j \rangle + w^-(i,j) (1 - \langle v_i, v_j \rangle))$$

$$\text{s.t. } v_i \in \{e_1, e_2, \dots, e_n\} \quad \forall i \in [n]$$

$\uparrow$
standard basis vectors

Thm: This is an exact formulation

pf: Let $\mathcal{S} = (\mathcal{S}_1, \mathcal{S}_2, \dots, \mathcal{S}_k)$ partition of V

For $j \in [k]$ and $i \in \mathcal{S}_j$, let $v_i = e_j$

$$\Rightarrow \sum_{(i,j) \in E} (w^+(i,j) \langle v_i, v_j \rangle + w^-(i,j) (1 - \langle v_i, v_j \rangle))$$

$$\sim \sum_{(i,j) \in E(\mathcal{S})} w^+(i,j) + \sum_{(i,j) \notin E(\mathcal{S})} w^-(i,j) = W(\mathcal{S})$$

Other direction: Let $\{v_i\}$ solution to vector program.

$$\forall j \in [k], \text{ let } \mathcal{S}_j = \{i \in V : v_i = e_j\}$$

$\Rightarrow \{\mathcal{S}_1, \mathcal{S}_2, \dots, \mathcal{S}_n\}$ form partition of V & is vector program objective

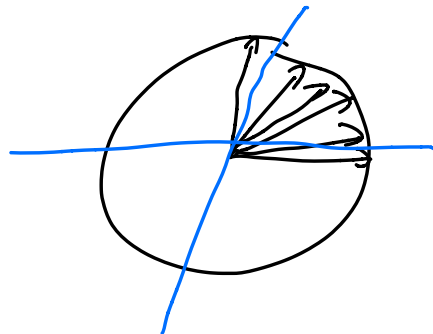
Relax to SDP:

$$\max \sum_{(i,j) \in E} (w^+(i,j) \langle v_i, v_j \rangle + w^-(i,j) (1 - \langle v_i, v_j \rangle))$$

$$\text{s.t. } \langle v_i, v_i \rangle = 1 \quad \forall i \in V \quad (\text{unit vectors})$$

$$\langle v_i, v_j \rangle \geq 0 \quad \forall i, j \in V$$

$$v_i \in \mathbb{R}^n \quad \forall i \in V$$



Randomly: **two** random hyperplanes

- Let r_1, r_2 unit vectors drawn independently, uniformly at random from set of all unit vectors

$$\text{- Let } R_1 = \{i \in V : \langle r_1, v_i \rangle \geq 0, \langle r_2, v_i \rangle \geq 0\}$$

$$R_2 = \{i \in V : \langle r_1, v_i \rangle \geq 0, \langle r_2, v_i \rangle < 0\}$$

$$R_3 = \{i \in V : \langle r_1, v_i \rangle < 0, \langle r_2, v_i \rangle \geq 0\}$$

$$R_4 = \{i \in V : \langle r_1, v_i \rangle < 0, \langle r_2, v_i \rangle < 0\}$$

- Return $\mathcal{A} = \{R_1, R_2, R_3, R_4\}$

$$X_{ij} = \begin{cases} 1 & \text{if } i, j \text{ in same cluster} \\ 0 & \text{otherwise} \end{cases}$$

last time:

$$P[\text{random hyperplane separates } i, j] = \frac{\theta_{ij}}{\pi}$$

$$\Rightarrow E[X_{ij}] = \left(1 - \frac{\theta_{ij}}{\pi}\right)^2$$

$$\Rightarrow E[w(S)] = E\left[\sum_{\{i,j\} \in E(S)} w^+(i,j) + \sum_{\{i,j\} \in \delta(S)} w^-(i,j)\right]$$

$$= E\left[\sum_{\{i,j\} \in E} \left(w^+(i,j) X_{ij} + w^-(i,j) (1 - X_{ij})\right)\right]$$

$$= \sum_{\{i,j\} \in E} \left(w^+(i,j) E[X_{ij}] + w^-(i,j) (1 - E[X_{ij}])\right)$$

$$= \sum_{\{i,j\} \in E} \left(w^+(i,j) \left(1 - \frac{\theta_{ij}}{\pi}\right)^2 + w^-(i,j) \left(1 - \left(1 - \frac{\theta_{ij}}{\pi}\right)^2\right)\right)$$

$$\text{Trig facts: } \left. \begin{aligned} \left(1 - \frac{\theta_{ij}}{\pi}\right)^2 &\geq \frac{3}{4} \cos \theta_{ij} \\ 1 - \left(1 - \frac{\theta_{ij}}{\pi}\right)^2 &\geq \frac{3}{4} (1 - \cos \theta_{ij}) \end{aligned} \right\} \text{ if } \theta_{ij} \leq \frac{\pi}{2}$$

$$\text{SPP constraint } \langle v_i, v_j \rangle \geq 0 \Rightarrow \theta_{ij} \leq \frac{\pi}{2}$$

$$\geq \sum_{\{i,j\} \in E} \left(w^+(i,j) \cdot \frac{3}{4} \cos \theta_{ij} + w^-(i,j) \cdot \frac{3}{4} (1 - \cos \theta_{ij})\right)$$

$$= \frac{3}{4} \sum_{\{i,j\} \in E} \left(w^+(i,j) \langle v_i, v_j \rangle + w^-(i,j) (1 - \langle v_i, v_j \rangle) \right)$$

$$= \frac{3}{4} \cdot \text{OPT(SDP)}$$

$\Rightarrow \frac{3}{4}$ -approximation

Max-2SAT:

Input: - Variables $x_1, x_2, \dots, x_n$

- CNF clauses $C_1, C_2, \dots, C_m$, each with 2 literals

$(x_1 \vee \bar{x}_2), (\bar{x}_2 \vee x_5), (x_3 \vee x_1), \dots$

Feasible solution: Assignment $\{x_1, \dots, x_n\} \rightarrow \{T, F\}$

Objective: max # satisfied (true) clauses

Note: 2SAT not NP-hard, but Max-2SAT is!

want to write strict quadratic program, where

$y_i = -1$ corresponds to $x_i = T$

$y_i = 1$ corresponds to $x_i = F$

Problem: Symmetry!

$$C = X_i \vee \bar{X}_j \Rightarrow y_i = -1, y_j = 1 \text{ should be ok, but}$$
$$y_i = 1, y_j = -1 \text{ should not.}$$

But strictness means can only look at products!

Solution: add "dummy variable" $y_T \in \{-1, 1\}$

$$y_i = y_T \text{ corresponds to } x_i = T$$

$$y_i = -y_T \text{ corresponds to } x_i = F$$

Consider clause $X_i \vee X_j$, values y_i, y_j, y_T :

$$\frac{3 + y_i y_T + y_j y_T - y_i y_j}{4} = \begin{cases} 0 & \text{if } y_i = y_j = -y_T \\ 1 & \text{otherwise} \end{cases}$$

Clause $X_i \vee \bar{X}_j$, values y_i, y_j, y_T : same thing, negate y_j

$$\frac{3 + y_i y_T - y_j y_T + y_i y_j}{4} = \begin{cases} 0 & \text{if } y_i = -y_T, y_j = y_T \\ 1 & \text{otherwise} \end{cases}$$

clause $\bar{x}_i \vee x_j$, values y_i, y_j, y_T :

$$\frac{3 - y_i y_T + y_j y_T + y_i y_j}{4} = \begin{cases} 0 & \text{if } y_i = y_T, y_j = -y_T \\ 1 & \text{otherwise} \end{cases}$$

clause $\bar{x}_i \vee \bar{x}_j$, values y_i, y_j, y_T :

$$\frac{3 - y_i y_T - y_j y_T - y_i y_j}{4} = \begin{cases} 0 & \text{if } y_i = y_j = y_T \\ 1 & \text{otherwise} \end{cases}$$

Strict Quadratic Programming Formulation:

$$\begin{aligned} \max \quad & \sum_{\text{clause } x_i \vee x_j} \frac{3 + y_i y_T + y_j y_T - y_i y_j}{4} + \sum_{\text{clause } x_i \vee \bar{x}_j} \frac{3 + y_i y_T - y_j y_T + y_i y_j}{4} \\ & + \sum_{\text{clause } \bar{x}_i \vee x_j} \frac{3 - y_i y_T + y_j y_T + y_i y_j}{4} + \sum_{\text{clause } \bar{x}_i \vee \bar{x}_j} \frac{3 - y_i y_T - y_j y_T - y_i y_j}{4} \end{aligned}$$

$$\text{s.t.} \quad y_i^2 = 1 \quad \forall i \in [n]$$

$$y_T^2 = 1$$

Relax to SDP:

$$\max \sum_{\substack{\text{clause } \\ x_i \vee x_j}} \frac{3 + \langle v_i, v_T \rangle + \langle v_j, v_T \rangle - \langle v_i, v_j \rangle}{4} + \sum_{\substack{\text{clause } \\ x_i \vee \bar{x}_j}} \frac{3 + \langle v_i, v_T \rangle - \langle v_j, v_T \rangle + \langle v_i, v_j \rangle}{4} \\ + \sum_{\substack{\text{clause } \\ \bar{x}_i \vee x_j}} \frac{3 - \langle v_i, v_T \rangle + \langle v_j, v_T \rangle + \langle v_i, v_j \rangle}{4} + \sum_{\substack{\text{clause } \\ \bar{x}_i \vee \bar{x}_j}} \frac{3 - \langle v_i, v_T \rangle - \langle v_j, v_T \rangle - \langle v_i, v_j \rangle}{4}$$

$$\text{s.t. } \langle v_i, v_i \rangle = 1 \quad \forall i \in [n]$$

$$v_i \in \mathbb{R}^n \quad \forall i \in [n]$$

$$\langle v_T, v_T \rangle = 1$$

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

Relaxation $\Rightarrow \text{OPT(SDP)} \geq \text{OPT}$

Rounding: random hyperplane!

- Choose unit vector $r \in \mathbb{R}^n$ u.a.r.

$$\text{- Set } x_i = \begin{cases} T & \text{if } \text{sign}(\langle r, v_i \rangle) = \text{sign}(\langle r, v_T \rangle) \\ F & \text{if } \text{sign}(\langle r, v_i \rangle) \neq \text{sign}(\langle r, v_T \rangle) \end{cases}$$

(consider clause $x_i \vee x_j$ (other types similar))

$$\frac{3 + \langle v_i, v_T \rangle + \langle v_j, v_T \rangle - \langle v_i, v_j \rangle}{4} = \frac{1}{4} \left((1 + \langle v_i, v_T \rangle) + (1 + \langle v_j, v_T \rangle) + (1 - \langle v_i, v_j \rangle) \right)$$

$\Rightarrow$ all terms $(1 \pm \langle v_k, v_l \rangle)$ for some k, l (possibly $= T$)

Analyze each term

Recall $\alpha_{GW} = \inf_{0 \leq \theta \leq \pi} \frac{2\theta}{\pi(1-\cos\theta)}$

consider $1 - \langle v_k, v_l \rangle$

$\Rightarrow$ contribution to SDP is $1 - \langle v_k, v_l \rangle = 1 - \cos \theta_{kl}$

$\Pr[v_k, v_l \text{ separated by hyperplane}] = \frac{\theta_{kl}}{\pi}$

$\Rightarrow E[1 - y_k y_l] = 2 \cdot \Pr[v_k, v_l \text{ separated}] = 2 \frac{\theta_{kl}}{\pi} \geq \alpha_{GW} (1 - \cos \theta_{kl})$

$\Rightarrow$ in expectation, rounded solution gets $\geq \alpha_{GW} \cdot \text{SDP}$

consider $1 + \langle v_k, v_l \rangle$

$\Rightarrow$ contribution to SDP = $1 + \cos \theta_{kl}$

contribution to rounded solution: $E[1 + y_k y_l] = 2 \left(1 - \frac{\theta_{kl}}{\pi}\right)$

$\Rightarrow$ ratio b/w contributions = $\frac{2 \left(1 - \frac{\theta_{kl}}{\pi}\right)}{1 + \cos \theta_{kl}} = \frac{2(\pi - \theta_{kl})}{\pi(1 + \cos \theta_{kl})}$

Let $\theta' = \pi - \theta_{kl} \Rightarrow \frac{2\theta'}{\pi(1 - \cos \theta')}$ $\geq \alpha_{GW}$

$\cos(\pi - \theta) = -\cos \theta$

$\Rightarrow$ rounded sol. $\geq \alpha_{GW} \cdot \text{SDP}$