

Math 582G

Today: Applications of SOS & moments in combinatorial optimization

## Constrained Polynomial Optimization

Given  $f, g_1, \dots, g_r \in \mathbb{R}[x_1, \dots, x_n]_{\leq d}$ , find

$$f^* = \max_{x \in \mathbb{R}^n} f(x) \text{ s.t. } g_1(x) = 0, \dots, g_r(x) = 0$$

$$= \min_{c \in \mathbb{R}, x \in \mathbb{R}^n} c \text{ s.t. } c - f(x) \geq 0 \text{ on } \mathbb{V}_{\mathbb{R}}(g_1, \dots, g_r)$$

$$\leq f_{\text{SOS}, d}^* = \min c \text{ s.t. } c - f(x) = \sigma + \sum_{i=1}^r h_i g_i$$

for some  $\sigma \in \text{SOS}_{n, 2d}$

Dual version:

$$f_{\text{SOS}, d}^* = \max_{L \in \mathbb{R}[x]_{\leq d}^*} L(f(x)) \text{ s.t. } L \in \text{SOS}_{n, 2d}^*$$
$$L(h_i g_i) = 0 \quad \forall h_i \in \mathbb{R}[x]$$

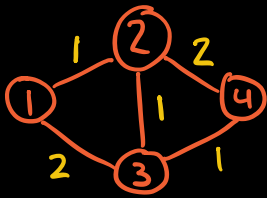
with  $\deg(h_i g_i) \leq d$

Interesting applications in comb. opt. when  $\mathbb{V}_{\mathbb{R}}(g_1, \dots, g_r)$  is finite!

MAXCUT: Let  $G = ([n], E)$  be a graph on  $n$  vertices with edges  $E \subseteq \binom{[n]}{2}$  and weights  $w: E \rightarrow \mathbb{R}_{\geq 0}$ .

$$\text{MAXCUT}(G) = \max_{S \subseteq [n]} w(S, [n] \setminus S) = \max_{S \subseteq [n]} \sum_{\substack{i \in S \\ j \in [n] \setminus S}} w(i, j)$$

Ex:



$$S = \{1, 2\} \quad w(12, 34) = 5$$

$$S = \{1, 4\} \quad w(14, 23) = 6$$

Known to be NP Hard in general.

→ hope for good approximations in polynomial time

As a poly. opt. problem:

$$\text{Claim: } \text{MAXCUT}(G) = \max \sum_{i, j \in E} w_{ij} \left( \frac{1 - x_i x_j}{2} \right) \quad \text{s.t. } x_i^2 = 1 \text{ for } i \in [n]$$

(Proof) Take  $S = \{i \in [n] \text{ s.t. } x_i = 1\}$ . Then

$$\frac{w_{ij}(1 - x_i x_j)}{2} = \begin{cases} 0 & \text{if } i, j \in S \text{ or } i, j \notin S \\ w_{ij} & \text{if } i \in S, j \notin S \text{ or } j \in S, i \notin S. \end{cases}$$

SOS approximation:

$$\min c \quad \text{s.t.} \quad c - \sum_{i, j \in E} w_{ij} \frac{(1 - x_i x_j)}{2} = \sigma + \sum_{i=1}^n h_i (1 - x_i^2) \\ \sigma \in \text{SOS}_{n, 2}$$

## Dual moment approx

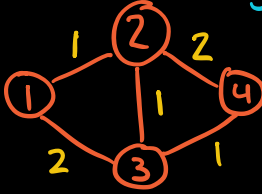
$$\max_{L \in \mathbb{R}[x_1, \dots, x_n]_{\leq 2}^*} L\left(\sum_{ij \in E} w_{ij} \frac{(1-x_i x_j)}{2}\right) \quad \text{s.t.} \quad \begin{array}{l} L \geq 0 \text{ on } \text{SDS}_{n,2} \\ L(1-x_i^2) = 0 \quad \forall i \end{array}$$

Write  $y_{ij} = L(x_i x_j)$ . Then  $L(1-x_i^2) = 0 \Leftrightarrow y_{ii} = 1$

Translation of dual:

(Goemans-Williamson approximation)

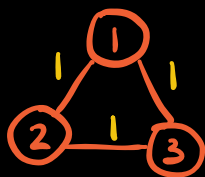
$$\text{GW}(G) = \max \sum_{ij} w_{ij} \frac{(1-y_{ij})}{2} \quad \text{s.t.} \quad Y = \begin{pmatrix} 1 & y_{12} & \dots & y_{1n} \\ y_{12} & 1 & & \\ \vdots & & \ddots & \\ y_{1n} & \dots & & 1 \end{pmatrix} \succeq 0$$

Ex:   $\text{GW}(G) = 6$  achieved by  $Y^* = \begin{pmatrix} 1 & -1 & -1 & 1 \\ -1 & 1 & 1 & -1 \\ -1 & 1 & 1 & -1 \\ 1 & -1 & -1 & 1 \end{pmatrix}$

Dual SOS certificate:

$$\begin{aligned} & 6 - \frac{2}{2}(2 - x_1 x_3 - x_2 x_4) - \frac{1}{2}(3 - x_1 x_2 - x_2 x_3 - x_3 x_4) \\ &= \frac{1}{4} \left[ X^T \begin{pmatrix} 3 & 1 & 2 & 0 \\ 1 & 2 & 1 & 2 \\ 2 & 1 & 2 & 1 \\ 0 & 2 & 1 & 3 \end{pmatrix} X + 3(1 - x_1^2 + 1 - x_4^2) + 2(1 - x_2^2 + 1 - x_3^2) \right] \end{aligned}$$

Ex (non-ex)



$$\text{MAXCUT}(G) = 2$$

$$\text{GW}(G) = \frac{9}{4}$$

achieved by  $Y^* = \begin{pmatrix} 1 & -1/2 & -1/2 \\ -1/2 & 1 & -1/2 \\ -1/2 & -1/2 & 1 \end{pmatrix} \rightarrow \sum_{ij} \frac{(1+1/2)}{2} = 3 \cdot \frac{3}{4} = \frac{9}{4}$

$$2 - \frac{1}{2}(3 - x_1x_2 - x_1x_3 - x_2x_3) \geq 0 \text{ on } \{\pm 1\}^3$$

but  $\neq \sigma + \sum h_i(1-x_i^2)$  with  $\sigma \in \text{SOS}_{3,1}$

$$\text{Thm: } .878 \text{ GW}(G) \leq \text{MAXCUT}(G) \leq \text{GW}(G)$$

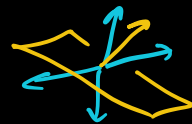
(Proof) Suppose  $\gamma^*$  achieves  $\text{GW}(G)$ .

Write  $\gamma^* = VV^T$  where  $V = \begin{pmatrix} -v_1 \\ \vdots \\ -v_n \end{pmatrix} \in \mathbb{R}^{n \times r}$ .

Note  
 $x_{ij} = \langle v_i, v_j \rangle$

Choose  $u \in \mathbb{R}^r$ ,  $\|u\|_2 = 1$  uniformly at random.

Let  $S = \{i \in [n] : \langle v_i, u \rangle > 0\}$ .



$$\mathbb{E}_u (\omega(S, [n] \setminus S)) = \sum_{i,j \in E} w_{ij} \text{Prob}(i,j \text{ separated})$$



$$= \sum_{i,j \in E} w_{ij} \frac{\text{angle}(v_i, v_j)}{\pi}$$

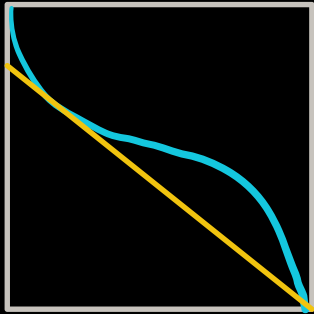
$$= \sum_{i,j \in E} w_{ij} \frac{\arccos(\langle v_i, v_j \rangle)}{\pi}$$

$$= \sum_{i,j \in E} w_{ij} \frac{\arccos(\gamma_{ij})}{\pi}$$

$$\geq \gamma \cdot \sum_{i,j \in E} w_{ij} \frac{(1 - \gamma_{ij})}{2} = \gamma \text{ GW}(G)$$

Here  $\gamma = .878$  is largest value s.t.

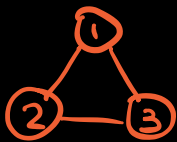
$$\frac{\arccos(t)}{\pi} \geq \gamma \cdot \frac{1-t}{2} \text{ for all } t \in [-1, 1]$$



$$\frac{\arccos(t)}{\pi}$$

$$.878 \left( \frac{1-t}{2} \right)$$

Ex:



$$Y^* = \begin{pmatrix} 1 & -1/2 & -1/2 \\ -1/2 & 1 & -1/2 \\ -1/2 & -1/2 & 1 \end{pmatrix} = VV^T \text{ for } V = \begin{pmatrix} 1 & 0 \\ -1/2 & \sqrt{3}/2 \\ -1/2 & -\sqrt{3}/2 \end{pmatrix}$$

