

LECTURE 17

- Last class, we proposed a semidefinite relaxation:
 - combinatorial, NP hard
 - semidefinite program, tractable

$$\max_{x_i \in \{\pm 1\}} \sum_{ij} a_{ij} x_i x_j \leq \max_{\substack{Z \succeq 0, \\ Z_{ii} = 1 \forall i}} \sum_{ij} a_{ij} Z_{ij} \quad (*)$$

\Downarrow Z_{ij}

- We will prove tightness: $RHS \leq C \cdot LHS$.

- But before that, let's rewrite (*) in terms of:

Def/Prop | The Gram matrix of vectors $v_1, \dots, v_n \in \mathbb{R}^d$ is $[\langle v_i, v_j \rangle]_{i,j=1}^n$.
 It is always PSD. (DIY)
 Conversely, $\forall n \times n$ PSD matrix is a Gram matrix of some vectors $v_1, \dots, v_n \in \mathbb{R}^n$. (DIY: consider the square root)

- In RHS (*), $Z = [\langle v_i, v_j \rangle]$, $Z_{ii} = \langle v_i, v_i \rangle = \|v_i\|_2^2 = 1 \forall i \Rightarrow$

$$RHS(*) = \max_{\substack{v_i \in \mathbb{R}^n \\ \text{unit}}} \sum_{ij} a_{ij} \langle v_i, v_j \rangle \quad (**)$$

SDP \Rightarrow tractable (1000's of variables)

- We will work out tightness of Max Cut:

SEMIDEFINITE RELAXATION OF MAX CUT

[Goemans-Williamson '1994]

- Recall from Lec. 16 that for a graph G with adjacency matrix (a_{ij}) ,

$$\text{Max Cut}(G) = \frac{1}{4} \max_{x_i \in \{\pm 1\}} \sum_{i,j=1}^n a_{ij} (1 - x_i x_j)$$

$x_i = 1$ ~~$x_j = -1$~~

- According to (*), a semidefinite relaxation is

$$\text{SDP}(G) = \frac{1}{4} \max_{\substack{v_i \in \mathbb{R}^n \\ \text{unit}}} \sum_{i,j=1}^n a_{ij} (1 - \langle v_i, v_j \rangle)$$

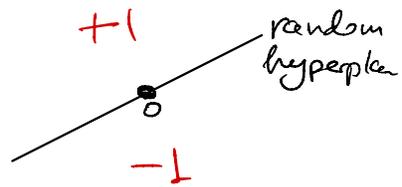
Trivially, $\text{Max Cut}(G) \leq \text{SDP}(G)$ ($v_i = (\pm 1, 0, 0, \dots)$) (*)

- Solve SDP; how do we convert solution (v_i) to (x_i) ?

- "Randomized rounding"

Let $g \sim N(0, I_n)$.
rotation invariant

$v \mapsto \text{sign} \langle v, g \rangle$



- Proposed alg.:
 - (a) Solve $\text{SDP}(G) \Rightarrow$ solution (v_i)
 - (b) Randomized rounding: $x_i := \text{sign} \langle v_i, g \rangle$.
 Output partition: $i: x_i = 1$ ~~$j: x_j = -1$~~

Accuracy:

TKM The expected cut of this partition $\geq 0.878 \cdot \text{Max Cut}(G)$

Proof

$$\text{Expected cut} = \mathbb{E} \frac{1}{4} \sum_{i,j=1}^n a_{ij} (1 - x_i x_j)$$

$$= \frac{1}{4} \sum_{i,j=1}^n a_{ij} (1 - \mathbb{E} x_i x_j)$$

=

$$\mathbb{E} \text{sign} \langle v_i, g \rangle \langle v_j, g \rangle = ?$$

Lemma (Grothendieck's identity) \forall unit vectors $u, v \in \mathbb{R}^n$, $g \sim N(0, I_n)$

$$\mathbb{E} \text{sign} \langle u, g \rangle \text{sign} \langle v, g \rangle = \frac{2}{\pi} \arcsin \langle u, v \rangle.$$

Rotation invariance

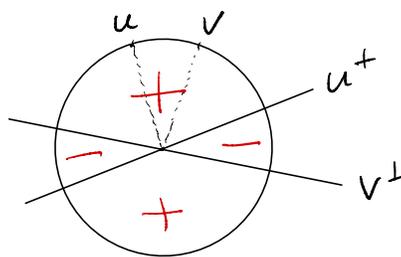
$$\Rightarrow \text{wlog, } u, v \in \mathbb{R}^2$$

$$\Rightarrow g \sim N(0, I_2)$$

$$g \mapsto g / \|g\|_2 =: \theta \sim \text{Unif}(\text{circle}).$$

$$\Rightarrow \mathbb{E} \text{sign} \langle u, \theta \rangle \text{sign} \langle v, \theta \rangle = \text{length}(\text{"+" arcs}) - \text{length}(\text{"-" arcs}).$$

Then trigonometry (DIR).

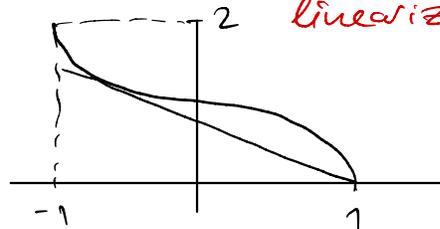


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$$\frac{1}{4} \sum_{i,j=1}^n a_{ij} \left(1 - \frac{2}{\pi} \arcsin \langle v_i, v_j \rangle \right) = \frac{1}{4} \sum_{i,j} a_{ij} \cdot \frac{2}{\pi} \arccos \langle v_i, v_j \rangle$$

linearize?

Fact: $\frac{2}{\pi} \arccos \theta \geq 0.878 (1 - \theta)$



≥

$$0.878 \cdot \frac{1}{4} \sum_{i,j} a_{ij} (1 - \langle v_i, v_j \rangle)$$

$$\text{SDP}(G) \geq \text{Max Cut}(G).$$

(*) 1.2

QED.