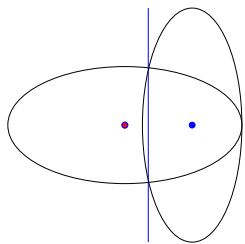


# Today

Semidefinite Programming  
...for Approximating MaxCut.

## Semidefinite Programming: polynomial time.

Ellipsoid algorithm.



Only get close!  
 $O(\log \frac{1}{\epsilon})$  dependence on closeness.

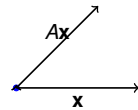
- Enclosing Ellipse.
- Center point not feasible.
- New Ellipsoid.
- $\leq (1 - 1/\text{poly}(n))$  volume.
- Center point feasible?
- Linear Programming:
- find violated constraint.
- Semidefinite Programming:
- find  $x$  where  $x^T A x \leq 0$ .
- Compute smallest eigenvalue.

## Positive Semidefinite Matrices

$n \times n$  matrix  $A$  positive semidefinite  $\rightarrow$  for all  $x \in \mathbb{R}^n$ ,  $x^T A x \geq 0$ .

Hmm...

$Ax$  is same direction as  $x$ ?



Spectral decomposition:

$$A = \sum_i \lambda_i v_i v_i^T.$$

$\lambda_i \geq 0$  – eigenvalue.

$v_i$  – associated eigenvector.

$$A = B^T B \quad B = \sqrt{\lambda_i} \begin{bmatrix} v_1 \\ v_2 \\ \dots \end{bmatrix}$$

Representation  $\rightarrow$  positive semidefinite too:  $(x^T B^T)(Bx) \geq 0$

Possibly many such representations.

## Semidefinite Programming: another view.

$$\begin{aligned} \max \quad & A \cdot C \\ A \cdot X_i & \geq b_i \\ A & \succeq 0 \end{aligned}$$

Recall  $A = B^T B$ .

Programming over vectors:  $v_1, v_2, \dots, v_n$ .

Linear Constraints over  $v_i \cdot v_j$   
quadratic..kind of!

## Semidefinite Programming.

$$\begin{aligned} \max \quad & A \cdot C \\ A \cdot X_i & \geq b_i \\ A & \succeq 0 \end{aligned}$$

(1)

$A \cdot X$  is matrix inner product:  $\sum_{ij} a_{ij} x_{ij}$ .

view  $A$  and  $X$  as  $n^2$  dimensional vector.

Linear Programming?  $A$  must be diagonal.

Constraint for each  $i \neq j$ ,  
 $X_{jk}$  is 1 at entry  $jk$ , 0 elsewhere.  $b_{jk}$  is 0.

Solvable?

Convex: Solution  $A$  and  $A'$ .

$\mu A + (1 - \mu)A'$  is solution.

Linear constraints, objective.

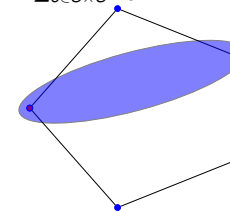
$x^T A x, x^T A' x \geq 0 \implies x^T (\mu A + (1 - \mu)A') x \geq 0$ .

Actually: psd is "cone" constraint.

## Max Cut

Given a graph  $G = (V, E)$ , with  $w : E \rightarrow \mathbb{R}$ , find  $S$  where  
 $\sum_{e \in S \times S} w_e$  is maximized.

(2)



Max Cut Size: 4

## Factor half approximation?

Random: choose a side at random.

Each edge has probability  $\frac{1}{2}$  of being cut.

Expected value of solution is half total edge weight.

Greedy: choose larger choice.

When each node comes, cuts at least half previous edges.

Can we do better?

## Embedding problem.

Assign variables  $x_1, \dots, x_n$  to vertices.

$x_i$  are  $\pm 1$ .

Maximize  $\sum_{ij} w_{ij} \frac{1-x_i x_j}{2}$ .

Cost of cut indicated by  $\pm 1$  vector!

Integer? Quadratic?

Assign vectors  $v_1, v_2, \dots, v_n$  to vertices.

$|v_i| = 1$

Maximize  $\sum_{ij} w_{ij} \frac{1-v_i \cdot v_j}{2}$ .

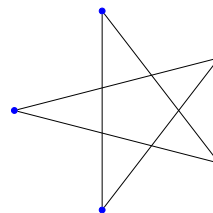
Semidefinite Program. Can solve? (Basically.)

Assign vectors  $v_1, v_2, \dots, v_n$  to vertices.

$|v_i| = 1$

Maximize  $\sum_{ij} w_{ij} \frac{1-v_i \cdot v_j}{2}$ .

Example?

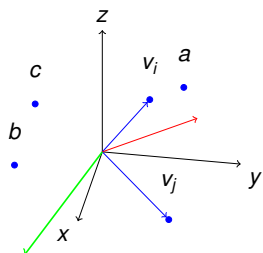


Solution Value:  $5 \frac{(1-\cos(4\pi/5))}{2} \approx 4.52$

Higher than opt.

Round and not lose too much?

## Hyperplane rounding.



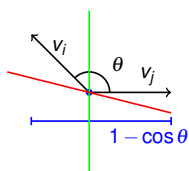
Normal to hyperplane:  
red does not separate!  
green does.

## Hyperplane rounding.

Take a random vector,  $w$

Let  $S = \{w \cdot v \geq 0\}$

**Claim 1:** Expected weight of  $(S, V - S)$  is at least 0.878 SDPOPT.



Ratio is  $\frac{2\theta}{\pi(1-\cos\theta)}$ .  
Always bigger than .878!

SDP value for edge  $(i, j)$ .  
 $\frac{(1-\cos\theta)}{2}$

Prob. of cutting:

$\frac{\theta}{\pi}$

Expected value in rounding!

See you on Thursday.