# CS675: Convex and Combinatorial Optimization Fall 2014 

Combinatorial Problems as Convex Programs

Instructor: Shaddin Dughmi

## The Max Cut Problem

Given an undirected graph $G=(V, E)$, find a partition of $V$ into ( $S, V \backslash S$ ) maximizing number of edges with exactly one end in $S$.

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\begin{array}{ll}
\operatorname{maximize} & \sum_{(i, j) \in E} \frac{1-x_{i} x_{j}}{2} \\
\text { subject to } & x_{i} \in\{-1,1\}, \quad \text { for } i \in V .
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Instead of requiring $x_{i}$ to be on the 1 dimensional sphere, we relax and permit it to be in the $n$-dimensional sphere.

## Vector Program relaxation

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\begin{array}{lll}
\operatorname{maximize} & \sum_{(i, j) \in E} \frac{1-\vec{v}_{i} \cdot \vec{v}_{j}}{2} \\
\text { subject to } & \left\|\vec{v}_{i}\right\|_{2}=1, & \text { for } i \in V . \\
& \vec{v}_{i} \in \mathbb{R}^{n}, & \text { for } i \in V .
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## SDP Relaxation

- Recall: An $n \times n$ matrix $Y$ is PSD iff $Y=V^{T} V$ for $n \times n$ matrix $V$
- When diagonal entires of $Y$ are $1, V$ has unit length columns
- Equivalently: PSD matrices encode pairwise dot products of columns of $V$
- Recall: $Y$ and $V$ can be recovered from each other efficiently


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## Randomized Algorithm for Max Cut

(1) Solve the SDP to get $Y \succeq 0$
(2) Decompose $Y$ to $V V^{T}$
(3) Pick a random vector $r$ on the unit sphere
(4) Place all nodes $i$ with $v_{i} \cdot r \geq 0$ on one side of the cut, and all others on the other side

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## Lemma

The SDP cuts each edge with probability at least $0.878 \frac{1-Y_{i j}}{2}$
Consequently, by linearity of expectation, expected number of edges cut is at least 0.878 OPT.

## Lemma

The SDP cuts each edge with probability at least $0.878 \frac{1-Y_{i j}}{2}$
We use the following fact
Fact
For all angles $\theta \in[0, \pi]$,

$$
\frac{\theta}{\pi} \geq 0.878 \cdot \frac{1}{2}(1-\cos (\theta))
$$

to prove the Lemma on the board.

