# Maximum Cut

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Abstract

TODO

# 1 Introduction

## 2 Mathematical Formulations

### 3 Relaxations for Max Cut

#### 3.1 Eigenvalue Relaxation

**Proposition 1.** The optimal value of  $z^*$  in the maxcut problem defined on graph G = (V, E) satisfies

$$z^* \le z^{ev} := \frac{n}{4} \lambda_{max}(L(G))$$

*Proof.* From the previous section we saw that

$$z^* = \frac{1}{4} \max_{\mathbf{z} \in \{-1,1\}} \sum_{i \in V} \sum_{j \in V} w_{ij} (1 - z_i z_j)$$
 (1)

Since  $z_i \in \{-1, 1\}$  it is clear that  $z_i z_i = 1$ . Then, we can express (1) as

$$\frac{1}{4} \max_{\mathbf{z} \in \{-1,1\}} \sum_{i \in V} \sum_{j \in V} w_{ij} (z_i z_i - z_i z_j) \tag{2}$$

Rearranging these terms we see that

$$\frac{1}{4} \max_{\mathbf{z} \in \{-1,1\}} \sum_{i \in V} z_i \left( \sum_{j \in V} w_{ij} \right) z_i - \sum_{i \in V} z_i \sum_{j \in V} w_{ij} z_j \tag{3}$$

Recall that  $D(G)_{ii} = \sum_{j \in V} w_{ij}$  and  $D(G)_{ij} = 0$  for  $i, j \in V$ . The first term in (3) can be expressed as  $\mathbf{z}^{\top}D(G)\mathbf{z}$ . Likewise, the weighted adjacency matrix  $A(G)_{ii} = 0$  and  $A(G)_{ij} = w_{ij}$  for  $i, j \in V$ . The second term can be expressed as  $\mathbf{z}^{\top}A(G)\mathbf{z}$  and now

$$\frac{1}{4} \max_{\mathbf{z} \in \{-1,1\}} \mathbf{z}^{\top} D(G) \mathbf{z} - \mathbf{z}^{\top} A(G) \mathbf{z}$$

$$\tag{4}$$

Since the matrix product is distributive we have

$$\frac{1}{4} \max_{\mathbf{z} \in \{-1,1\}} \mathbf{z}^{\top} \left( D(G) - A(G) \right) \mathbf{z} \tag{5}$$

By definition of the Laplacian of G we have that the optimal solution of maxcut can be computed as

$$\frac{1}{4} \max \{ \mathbf{z}^{\mathsf{T}} L(G) \mathbf{z} \mid \mathbf{z} \in \{-1, 1\} \}$$
 (6)

Taking the continuous relaxation of (6), namely  $z \in [-1, 1]$ , is equivalent to maximizing over the  $\|\mathbf{z}\|_{\infty} = 1$ . We can relax this further by maximizing over a ball of radius  $\sqrt{n}$ . In other words, our region is now defined where  $\|\mathbf{z}\| \leq \sqrt{n}$ .

$$z^* \le \frac{1}{4} \max\{\mathbf{z}^\top L(G)\mathbf{z} \mid \mathbf{z} \le \sqrt{n}\}$$
 (7)

We can define our problem over the unit ball with a simple transformation. Let  $\|\mathbf{z}\| = \sqrt{n}\mathbf{y}$  and we now have

$$z^* \le \frac{n}{4} \max \{ \mathbf{y}^\top L(G) \mathbf{y} \mid ||\mathbf{y}|| \le 1 \}$$
 (8)

Since for any symmetric matrix  $\mathbf{A}$ ,  $\max\{\mathbf{x}^{\top}\mathbf{A}\mathbf{x} \mid \|\mathbf{x}\| \leq 1\} = \lambda_{\max(\mathbf{A})}$  and L(G) is symmetric, then we have

$$z^* \le z^{ev} := \frac{n}{4} \lambda_{max}(L(G)) \tag{9}$$