

## Seminar on Algorithms and Geometry

### Lecture 3

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## 1 Sparsest Cut

### 1.1 Approximation Algorithm for Sparsest Cut

**Stage 0:** Solve the relaxation of the sparsest cut problem. As it a linear program, it can be solved in polynomial time.

**Stage 1:** Embed the metric  $d$  found by the LP into  $l_1$  with distortion  $O(\log n)$ . This also can be done in polynomial time by Bourgain's theorem. So exists  $f : V \rightarrow l_1$  such that for every  $u, v \in V$

$$d(u, v) \leq \|f(u) - f(v)\| \leq O(\log n)d(u, v).$$

Therefore,  $\sum_{(u,v) \in E} \|f(u) - f(v)\|_1 \leq O(\log n)LP$  and  $\sum_i^k \|f(s_i) - f(t_i)\|_1 = 1$  where LP is the solution found by the linear programming relaxation.

**Lemma 1** Every  $n$ -point metric  $\tilde{d}$  that embeds isometrically into  $l_1$  can be written as a positive combination of cut metrics  $\tau_i$ . I.e., there exists  $\alpha_i > 0$  such that  $\tilde{d}(x, y) = \sum \alpha_i \tau_i(x, y)$  for every  $x, y \in V$ . Furthermore, such  $\alpha_i$  can be found in polynomial time and the number of  $\alpha_i > 0$  is at most  $\binom{n}{2}$ .

We now use Lemma 1 for the second stage of the approximation algorithm.

**Stage 2:** Write the distance  $\tilde{d}$  from the embedding as  $\tilde{d} = \sum \alpha_i \tau_i(x, y)$  for  $\tau_i$  cut metrics.

We now show that at least one of the cut metrics  $\tau_i$  yields the desired approximation.

**Claim 2** There exists  $j^*$  such that the objective  $OBJ|_{d=\tau_{j^*}} \leq OBJ|_{d=\tilde{d}}$ , i.e.,  $\frac{\sum \tau_{j^*}(u, v)}{\tau_{j^*}(s_i, t_i)} \leq \frac{\sum \tilde{d}(u, v)}{\tilde{d}(s_i, t_i)}$ .

**Proof** The proof is a generalization of the following.  $\forall a_1, \dots, a_m, b_1, \dots, b_m > 0 \min_l \frac{a_l}{b_l} \leq \frac{\sum a_l}{\sum b_l}$ . Assume, towards contradiction, that  $\min_l \frac{a_l}{b_l} > \frac{\sum a_l}{\sum b_l}$ . We get that  $b_1 \frac{\sum a_l}{\sum b_l} < a_1$ . Similarly  $b_i \frac{\sum a_l}{\sum b_l} < a_i$  for every  $1 \leq i \leq m$ . Therefore,  $\sum_l a_l = \sum_i b_i \frac{\sum a_l}{\sum b_l} < \sum_l a_l$ , a contradiction. ■

Finally,  $\tau_{j^*}$  gives us  $S_G^* \subseteq V$  whose value =  $OBJ|_{d=\tau_{j^*}} \leq OBJ|_{d=d^*} \leq O(\log n)LP \leq O(\log n)OPT$ .

Refinement: We can improve the approximation ratio to  $O(\log k)$  by having one side of distortion guarantee only for demand pairs, i.e.,  $d(u, v) \leq \|f(u) - f(v)\|$  only for the  $k$  demand pairs and  $\|f(u) - f(v)\| \leq Dd(u, v)$  for all pairs.

**Theorem 3 (Aumann-Rabani, Linial-London-Rabinovich95)** *Sparsest cut can be approximated in poly-time within factor  $O(\log k)$ .*

## 2 Minimum Bisection

The input of the *Minimum Bisection* problem is a graph  $G = (V, E)$  such that  $|V| = n$ . The goal is to find a cut  $(S, \bar{S})$  such that  $|S| = |\bar{S}| = n/2$  so as to minimize  $e(S, \bar{S})$ . This problem is known to be NP-hard.

Recall that the sparsest-cut problem with uniform demands is the search of a cut  $S$  that minimizes  $\frac{e(S, \bar{S})}{|S||\bar{S}|}$ . Note that  $\frac{e(S, \bar{S})}{|S||\bar{S}|} \cong \frac{e(S, \bar{S})}{\min\{|S|, |\bar{S}|\} \cdot n}$ , up to a factor of 2.

We now show a poly-time algorithm that finds a  $\frac{2}{3}$ -balanced cut  $S$  of cost  $e(S, \bar{S}) \leq O(\log n) \cdot b_G^*$  where  $b_G^*$  is the optimal cost of the minimum bisection problem on the graph  $G$ .

**Algorithm 2/3 – balanced – cut( $G = (V, E)$ )**

1. Set  $G_{alg} \leftarrow (V, E)$ , denote by  $V_{alg}$  the set of vertices of the graph  $G_{alg}$
2. While  $|V_{alg}| \leq \frac{2n}{3}$ 
  - use  $O(\log n)$  approximation algorithm for the sparsest-cut with uniform demands problem on  $G_{alg}$  to find a cut  $(S, \bar{S})$ , where  $|S| \leq |\bar{S}|$ .
  - remove  $S$  from  $G_{alg}$ .
3. return  $V_{alg}$ , the vertices of  $G_{alg}$ .

**Claim 4** *The set of vertices returned by the algorithm  $V_{alg}$  satisfies  $\frac{n}{3} \leq |V_{alg}| \leq \frac{2n}{3}$ .*

**Proof** At the beginning of the last iteration  $|V_{alg}| \geq \frac{2n}{3}$  and we remove at most half the vertices from  $V_{alg}$  (since we remove the smaller side of the cut). ■

Denote by  $b_G^*$  the optimal cost of the minimum bisection problem and by  $S_G^*$  the optimal cut that achieves the cost  $b_G^*$  where  $|S_G^*| = n/2$ .

**Claim 5** *The cost of  $V_{alg}$  is at most  $O(\log n)b_G^*$ .*

**Proof** Let  $S_\ell$  be the set removed in iteration  $\ell$ . Let  $S_\ell^*$  be the set that best minimizes  $\frac{e(S'_\ell, \bar{S}'_\ell)}{|S'_\ell|}$  in iteration  $\ell$ . Where  $\bar{S}'_\ell$  is the complement of the cut  $S'_\ell$  in the graph of iteration  $\ell$ . The set  $V_{alg}$  in iteration  $\ell$ , denoted by  $V_{alg}^\ell$ , contains at least  $\frac{2n}{3}$  nodes, therefore  $S_G^* \cap V_{alg}^\ell > \frac{n}{2} - \frac{n}{3} = \frac{n}{6}$  and also  $\bar{S}_G^* \cap V_{alg}^\ell > \frac{n}{6}$ . We get that  $\frac{e(S_\ell^*, \bar{S}_\ell^*)}{|S_\ell^*|} \leq \frac{b_G^*}{n/6}$  and as we use a  $\log n$ -approximation algorithm we now get  $\frac{e(S_\ell, \bar{S}_\ell)}{|S_\ell|} \leq O(\log n) \frac{e(S_\ell^*, \bar{S}_\ell^*)}{|S_\ell^*|} \leq O(\log n) \frac{b_G^*}{n/6}$ . Hence  $e(V_{alg}, \bar{V}_{alg}) \leq \sum_\ell e(S_\ell, \bar{S}_\ell) \leq O(\log n) \frac{b_G^*}{n/6} \cdot \sum_\ell |S_\ell| \leq O(b_G^* \log n)$ .

**Theorem 6 (Leighton-Rao 88)** *There is a poly-time algorithm that finds a  $\frac{2}{3}$ -balanced cut  $S$  of cost  $e(S, \bar{S}) \leq O(\log n) \cdot b_G^*$  where  $b_G^*$  is the optimal cost of the minimum bisection problem on the graph  $G$ .*

### 3 Distortion Lower Bounds

We now show a specific  $n$ -point space such that embedding this space to  $l_2$  requires distortion of at least  $\sqrt{\log n}$ .

**Lemma 7 (Short diagonals)** *Let  $x_1, x_2, x_3, x_4$  be points in  $l_2$ . Then  $\|x_1 - x_3\|^2 + \|x_2 - x_4\|^2 \leq \|x_1 - x_2\|^2 + \|x_2 - x_3\|^2 + \|x_3 - x_4\|^2 + \|x_4 - x_1\|^2$ .*

**Proof** Observe that it suffices to prove it for  $x_1, x_2, x_3, x_4 \in R$ . For points  $x_i$  in some  $R^d$ , simply apply the inequality on each coordinate and then add these inequalities together. So consider  $x_1, x_2, x_3, x_4 \in R$ ,  $\|x_1 - x_2\|^2 + \|x_2 - x_3\|^2 + \|x_3 - x_4\|^2 + \|x_4 - x_1\|^2 - \|x_1 - x_3\|^2 - \|x_2 - x_4\|^2 = |x_1 - x_2 + x_3 - x_4|^2 \geq 0$ . ■

**Theorem 8 (Enflo69)** *Let  $G = (V, E)$  be the discrete cube  $\{0, 1\}^m$  and shortest-path distance  $d_G(x, y) = \#(\text{bits } i \text{ such that } x_i \neq y_i)$ . Then embedding  $d_G$  into  $l_2$  requires distortion  $\geq \sqrt{m} = \sqrt{\log |V|}$ .*

Remark: The above is optimal. The identity mapping:  $x \rightarrow x$  has distortion  $\sqrt{m}$ .

**Proof** Consider  $V = \{0, 1\}^m$ . For  $x \in V$ , let  $\bar{x} \in \{0, 1\}^m$  be the complement of  $x$ . We will show that for every  $f : V \rightarrow l_2$ :

$$E_{x \in V}[\|f(x) - f(\bar{x})\|^2] \leq m \cdot E_{(x,y) \in E}[\|f(x) - f(y)\|^2]. \quad (1)$$

This would be enough to prove the lemma. By 1,

$$E_{x \in V}[\|f(x) - f(\bar{x})\|^2] \leq m \cdot E_{(x,y) \in E}[\|f(x) - f(y)\|^2] \leq m \cdot E_{(x,y) \in E}[d_G(x, y)^2] \leq m \cdot 1 = m.$$

So there exists a point  $x$  such that  $\|f(x) - f(\bar{x})\| \leq \sqrt{m} = \frac{d_G(x, \bar{x})}{\sqrt{m}}$ .

We now prove equation 1. We prove it by induction. For  $m = 2$ , use the short diagonal Lemma (divided by 2). Assume the claim holds for  $m' < m$  and consider  $m' = m$ . Let  $x$  be a point in  $\{0, 1\}^{m-1}$ . Apply the short diagonals lemma to  $x0, x1, \bar{x}0, \bar{x}1$ . We get  $\|f(x0) - f(\bar{x}1)\|^2 + \|f(x1) - f(\bar{x}0)\|^2 \leq \|f(x0) - f(x1)\|^2 + \|f(x1) - f(\bar{x}1)\|^2 + \|f(\bar{x}1) - f(\bar{x}0)\|^2 + \|f(\bar{x}0) - f(x0)\|^2$ . Finally, by summing on all  $x$ 's and using the induction hypothesis we get the desired inequality. ■