MATH 8500 Algorithmic Graph Theory, Spring 2017, OSU Lecture 5: *l*-Way Cut Continued Instructor: Anastasios Sidiropoulos Scribe: Tim Carpenter

1 *l*-Way Cut Continued

Recall from previous lectures the following problem statement

 ℓ -Way Cut: Input: G = (V, E), n = |V|Goal: Find a partition S_1, \ldots, S_ℓ of V maximizing |E(S)|, where

 $E(S) = \{\{u, v\} \in E : u \in S_i \text{ and } v \in S_j \text{ for } i \neq j\}$

and the following definition

Definition 1.1. We say that a partition V_1, \ldots, V_k of V is " ϵ -sufficient" if $|\Delta(S,T)| \le \epsilon \cdot n^2, \forall S, T \subseteq V, S \cap T = \emptyset,$ where • $\Delta(S,T) = e(S,T) - \sum_{i=1}^k \sum_{j=1}^k d_{i,j} |S_i| |T_i|$

•
$$e(S,T) = |E(S,T)|$$

•
$$d_{i,j} = d(V_i, V_j) = e(V_i, V_j) / |V_i| |V_j|.$$

Consider the example of Figure 1. Here we have the following:

- e(S,T) = 3
- $d_{1,1} = d_{2,2} = 0$

•
$$d_{1,2} = d_{2,1} = 4^2/4^2 = 1$$

Thus, in this case we have that

$$\sum_{i=1}^{k} \sum_{j=1}^{k} d_{i,j} |S_i| |T_i| = d_{1,2} |S_1| |T_2| + d_{2,1} |S_2| |T_s| = 2 \cdot 1 + 1 \cdot 1 = 3$$

and therefore Figure 1 is 0-sufficient.



Figure 1: An example of a 0-sufficient partitioning.

Theorem 1.2. There is a randomized polynomial time algorithm which given an *n*-vertex graph G, with probability at least 3/4, computes a partition S_{ϵ} such that

$$|E(S_{\epsilon})| \ge |E(S^*)| - \epsilon n^2,$$

where S^* is an optimal partition.

Proof. Compute an ϵ -sufficient partition V_1, \ldots, V_k of G using [Alon et al.]. Let $S = S_1, \ldots, S_\ell$ be an ℓ -way cut of G. Let $S_{i,r} = S_r \cap V_i$, and $T_{i,r} = V_i \setminus S_{i,r}$. We have that

$$2|E(S)| = \sum_{r=1}^{\ell} e(S_r, V \setminus S_r).$$

By the definition of ϵ -sufficient, we have that

$$2|E(S)| = \left(\sum_{r=1}^{\ell} \sum_{i=1}^{k} \sum_{j=1}^{k} d_{i,j}|S_{i,r}||T_{j,r}|\right) + \Theta$$

where $\Theta \leq \ell \epsilon n^2$. Let $v_j = |V_j|, \ \rho = \lfloor \frac{\epsilon n}{k} \rfloor, \ \bar{v}_j = \lfloor \frac{v_j}{\rho} \rfloor, \ n_{i,r} = |S_{i,r}|, \ \bar{n}_{i,r} = \lfloor \frac{n_{i,r}}{\rho} \rfloor$. We have

$$|n_{i,r}(v_j - n_{j,r}) - \rho^2 \bar{n}_{i,r}(\bar{v}_j - \bar{n}_{j,r})| \le \rho(v_i + v_j).$$

Thus

$$\left|\sum_{r=1}^{\ell}\sum_{i=1}^{k}\sum_{j=1}^{k}d_{i,j}n_{i,r}(v_j - n_{j,r}) - \rho^2\sum_{r=1}^{\ell}\sum_{i=1}^{k}\sum_{j=1}^{k}d_{i,j}\bar{n}_{i,r}(\bar{v}_j + \bar{n}_{j,r})\right| \le 2\epsilon\ell n^2.$$

Therefore,

$$\left| 2|E(S)| - \rho^2 \sum_{r=1}^{\ell} \sum_{i=1}^{k} \sum_{j=1}^{k} d_{i,j} \bar{n}_{i,r} (\bar{v}_j + \bar{n}_{j,r}) \right| \le 3\epsilon \ell n^2.$$



Figure 2: Illustration of $S_{i,r}$ and $T_{i,r}$. The outer circle corresponds to V, and the inner circle to S. The vertical lines mark the partitions V_1, \ldots, V_k .

Each $\bar{n}_{i,r}$ has at most $\frac{1}{\epsilon}$ different possible values. There are $k\ell$ variables $n_{i,r}$. Therefore, there are $(\frac{1}{\epsilon})^{k\ell}$ possibilities. It is sufficient to iterate over all these possibilities, and select the instance maximizing

$$\sum_{r=1}^{\ell} \sum_{i=1}^{k} \sum_{j=1}^{k} d_{i,j} \bar{n}_{i,r} (\bar{v}_j + \bar{n}_{j,r}).$$

Recall that we have the following lemma from a previous lecture:

Lemma 1.3. If
$$|E| \ge \gamma n^2$$
 then $|E(S^*)| \ge (1 - \frac{1}{\ell})|E| \ge \gamma (1 - \frac{1}{\ell})n^2$.

Using this lemma and the theorem above, we have the following corollary:

Corollary 1.4. For all $\epsilon > 0$ there exists a polynomial time algorithm that computes a $(1 + \epsilon)$ -approximation ℓ -way cut on dense graphs.

Thus, we have a **PTAS** (Polynomial Time Approximation Scheme) for ℓ -Way Cut on dense graphs.