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Lecture 14

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## 1 Demands Multicommodity Flow

Problem definition: We are given a weighted graph G = (V, E), where each edge e has a capacity of c(e). We have k different goods; each good i travels the graph from  $s_i$  to  $t_i$ , and has demand dem(i).

Our goal: to find and maximize f so that each good i has a flow of f \* dem(i) units.

For any cut S, we define  $c(S) = \sum_{e \in (S,\overline{S})} c(e)$ , and dem(S) is the summation of dem(i) for every i whose flow the cut S disrupts (i.e. flow is not possible from  $s_i$  to  $t_i$  after the cut). By these definitions, we immediately find:

$$f \le \frac{c(S)}{dem(S)}$$

We will refer to  $\frac{c(S)}{dem(S)}$  as the sparsity of cut S.

Last lesson, we showed that the optimal (fractional) answer can be viewed as a metric over the vertices V. In the LP problem,  $d_e$  is a variable representing whether or not e is in the cut; after relaxation  $d_e \in [0, 1]$ . For each two vertices  $u, v \in V$ , d(u, v) is the sum of all  $d_e$ s on the shortest path from u to v.

**Definition 1** A function  $y: 2^V \to \mathbb{R}$  is said to be a  $\beta$ -approximate-cut-packing if for every  $e \in E$ , where  $d_e$  is e's distance in the metric,

$$\frac{d_e}{\beta} \le \sum_{S|e \in S} y(S) \le d_e$$

**Theorem 2** Let (V,d) be the metric determined by LP, as seen last lesson. Let y be a  $\beta$ -approximate-cut-packing for (V,d). Of all the cuts S where  $y(S) \neq 0$ , let S' be the cut with minimal sparsity  $(\frac{c(S)}{dem(S)})$ . Then:

$$\beta OPT \le \frac{c(S')}{dem(S')}$$

**Lemma 3** Let  $\sigma: V \to \mathbb{R}^m$  be a mapping. Then there exists a cut-packing y so the weight each edge e 'feels'  $(\sum_{S|e \in S} y(S))$  is e's length in the metric  $l_1$  under  $\sigma$ . In other words, for e = (u, v):

$$\sum_{S|e \in S} y(S) = ||\sigma(u) - \sigma(v)||_{l_1}$$

**Theorem 4** There exists an O(logn) approximation to the SPARSEST-CUT problem.

**Proof:** We will find a  $\beta$ -approximation-cut-packing ( $\beta$ -approx-CP) for  $\beta = O(logn)$ ; this suffices to prove the theorem.

We will embed our metric (V, d) in  $l_1$ , using mapping  $\sigma$ . By Bourgain, we have a distortion of O(logn), which we will denote  $\beta$ . So for every e, we find for e = (u, v)

$$\frac{d_e}{\beta} \le \left| \left| \sigma(u) - \sigma(v) \right| \right|_{l_1} \le d_e$$

Using the lemma, there exists a cut-packing y so that  $\sum_{S|e\in S} y(S) = ||\sigma(u) - \sigma(v)||_{l_1}$ , and so we find

$$\frac{d_e}{\beta} \le \sum_{S|e \in S} y(S) \le d_e$$

which is precisely a  $\beta$ -approx-CP, as required.

## **Proof:** We prove the lemma for m = 1.

We have a mapping  $\sigma$  from our *n* vertices to  $u_1 \leq u_2 \leq ... \leq u_n \in \mathbb{R}$  (in order, without loss of generality). We define:

 $\begin{aligned} y(\{v_1\}) &= u_2 - u_1 \\ y(\{v_1, v_2\}) &= u_3 - u_2 \\ \dots \end{aligned}$ 

 $y(\{v_1, ..., v_k\}) = u_{k+1} - u_k$ 

For any set  $K \subseteq V$  not of the form  $\{v_1, v_2, ..., v_k\}$ , we'll define y(K) = 0. For any cut S where  $(u_1, u_n) \in S$ , we find:

$$\sum_{S|e \in S} y(S) = \sum_{S|v_1 \in S, v_n \notin S} y(S) = \sum_{1 \le k < n} y(\{v_1, v_2, \dots, v_k\}) = u_n - u_{n-1} + u_{n-1} - u_{n-2} + \dots + u_2 - u_1 = u_n - u_1 + u_{n-1} - u_{n-2} + \dots + u_n - u_n$$

And so we have found that  $(u_1, u_n)$  feels  $u_n - u_1 = ||v_n - v_1||_{l_1}$  Similarly, for any  $(u_i, u_j) \in S$ ,  $i \leq j$ , we find

$$\sum_{S|e \in S} y(S) = u_j - u_i = ||v_j - v_i||_{l_1}$$

since the first cut of the  $\{v_1, ..., v_k\}$  form that contains  $v_i$  is  $\{v_1, ..., v_i\}$ , and the last that does *not* contain  $v_j$  is  $\{v_1, ..., v_{j-1}\}$ .

Thus, we have proved the lemma for m = 1.

It is simple to generalize to m dimensions - simply project onto each dimension separately, and repeat the process to find the distance along that dimension.  $l_1$  distance is found by summation of the distances found along each dimension.