

## Lecture 14

*In which we begin to discuss the Arora-Rao-Vazirani rounding procedure.*

Recall that, in a graph  $G = (V, E)$  with adjacency matrix  $A$ , then ARV relaxation of the sparsest cut problem is the following semidefinite program.

$$\begin{aligned}
 & \text{minimize} && \frac{1}{2|E|} \sum_{u,v} A_{u,v} \|\mathbf{x}_u - \mathbf{x}_v\|^2 \\
 & \text{subject to} && \sum_{u,v} \|\mathbf{x}_u - \mathbf{x}_v\|^2 = |V|^2 \\
 & && \|\mathbf{x}_u - \mathbf{x}_v\|^2 \leq \|\mathbf{x}_u - \mathbf{x}_w\|^2 + \|\mathbf{x}_w - \mathbf{x}_v\|^2 \quad \forall u, v, w \in V \\
 & && \mathbf{x}_u \in \mathbb{R}^d \quad \forall u \in V
 \end{aligned}$$

If we denote by  $ARV(G)$  the optimum of the relaxation, then we claimed that

$$ARV(G) \leq \phi(G) \leq O(\sqrt{\log |V|}) \cdot ARV(G)$$

where the first inequality follows from the fact that  $ARV(G)$  is a relaxation of  $\phi(G)$ , and the second inequality is the result whose proof we begin to discuss today.

## 1 Rounding the Arora-Rao-Vazirani Relaxation

Given the equivalence between the sparsest cut problem and the “ $\ell_1$  relaxation” of sparsest cut, it will be enough to prove the following result.

**Theorem 1 (Rounding of ARV)** *Let  $G$  be a graph,  $A$  its adjacency matrix, and  $\{\mathbf{x}_v\}_{v \in V}$  be a feasible solution to the ARV relaxation.*

*Then there is a mapping  $f : V \rightarrow \mathbb{R}$  such that*

$$\frac{\sum_{u,v} A_{u,v} |f(u) - f(v)|}{\sum_{u,v} |f(u) - f(v)|} \leq O(\sqrt{\log |V|}) \cdot \frac{\sum_{u,v} A_{u,v} \|\mathbf{x}_u - \mathbf{x}_v\|^2}{\sum_{u,v} \|\mathbf{x}_u - \mathbf{x}_v\|^2}$$

As in the rounding of the Leighton-Rao relaxation via Bourgain's theorem, we will identify a set  $S \subseteq V$ , and define

$$f_S(v) := \min_{s \in S} \|\mathbf{x}_s - \mathbf{x}_v\|^2 \quad (1)$$

Recall that, as we saw in the proof of Bourgain's embedding theorem, no matter how we choose the set  $S$  we have

$$|f_S(u) - f_S(v)| \leq \|\mathbf{x}_u - \mathbf{x}_v\|^2 \quad (2)$$

where we are not using any facts about  $\|\cdot - \cdot\|^2$  other than the fact that, for solutions of the ARV relaxation, it is a distance function that obeys the triangle inequality.

This means that, in order to prove the theorem, we just have to find a set  $S \subseteq V$  such that

$$\sum_{u,v} |f_S(u) - f_S(v)| \geq \frac{1}{O(\sqrt{\log |V|})} \cdot \sum_{u,v} \|\mathbf{x}_u - \mathbf{x}_v\|^2 \quad (3)$$

and this is a considerable simplification because the above expression is completely independent of the graph! The remaining problem is purely one about geometry.

Recall that if we have a set of vectors  $\{\mathbf{x}_v\}_{v \in V}$  such that the distance function  $d(u, v) := \|\mathbf{x}_u - \mathbf{x}_v\|^2$  satisfies the triangle inequality, then we say that  $d(\cdot, \cdot)$  is a (semi-)metric of *negative type*.

After these preliminaries observations, our goal is to prove the following theorem.

**Theorem 2 (Rounding of ARV – Revisited)** *If  $d(\cdot, \cdot)$  is a semimetric of negative type over a set  $V$ , then there is a set  $S$  such that if we define*

$$f_S(v) := \min_{s \in S} \{d(s, v)\}$$

*we have*

$$\sum_{u,v} |f_S(u) - f_S(v)| \geq \frac{1}{O(\sqrt{\log |V|})} \cdot \sum_{u,v} d(u, v)$$

*Furthermore, the set  $S$  can be found in randomized polynomial time with high probability given a set of vector  $\{\mathbf{x}_v\}_{v \in V}$  such that  $d(u, v) = \|\mathbf{x}_u - \mathbf{x}_v\|^2$ .*

Since the statement is scale-invariant, we can restrict ourselves, with no loss of generality, to the case  $\sum_{u,v} d(u, v) = |V|^2$ .

**Remark 3** Let us discuss some intuition before continuing with the proof.

As our experience proving Bourgain's embedding theorem shows us, it is rather difficult to pick sets such that  $|f_S(u) - f_S(v)|$  is not much smaller than  $d(u, v)$ . Here we have a somewhat simpler case to solve because we are not trying to preserve all distances, but only the average pairwise distance. A simple observation is that if we find a set  $S$  which contains  $\Omega(|V|)$  elements and such that  $\Omega(|V|)$  elements of  $V$  are at distance  $\Omega(\delta)$  from  $S$ , then we immediately get  $\sum_{u,v} |f_S(u) - f_S(v)| \geq \Omega(\delta|V|^2)$ , because there will be  $\Omega(|V|^2)$  pairs  $u, v$  such that  $f_S(u) = 0$  and  $f_S(v) \geq \delta$ . In particular, if we could find such a set with  $\delta = 1/O(\sqrt{\log |V|})$  then we would be done. Unfortunately this is too much to ask for in general, because we always have  $|f_S(u) - f_S(v)| \leq d(u, v)$ , which means that if we want  $\sum_{u,v} |f_S(u) - f_S(v)|$  to have  $\Omega(V^2)$  noticeably large terms we must also have that  $d(u, v)$  is noticeably large for  $\Omega(|V|^2)$  pairs of points, which is not always true.

There is, however, the following argument, which goes back to Leighton and Rao: either there are  $\Omega(|V|)$  points concentrated in a ball whose radius is a quarter (say) of the average pairwise distance, and then we can use that ball to get an  $\ell_1$  mapping with only *constant* error; or there are  $\Omega(|V|)$  points in a ball of radius twice the average pairwise distance, such that the pairwise distances of the points in the ball account for a constant fraction of all pairwise distances. In particular, the sum of pairwise distances includes  $\Omega(|V|^2)$  terms which are  $\Omega(1)$ .

After we do this reduction and some scaling, we are left with the task of proving the following theorem: suppose we are given an  $n$ -point negative type metric in which the points are contained in a ball of radius 1 and are such that the sum of pairwise distances is  $\Omega(n^2)$ ; then there is a subset  $S$  of size  $\Omega(n)$  such that there are  $\Omega(n)$  points whose distance from the set is  $1/O(\sqrt{\log n})$ . This theorem is the main result of the Arora-Rao-Vazirani paper. (Strictly speaking, this form of the theorem was proved later by Lee – Arora, Rao and Vazirani had a slightly weaker formulation.)

We begin by considering the case in which a constant fraction of the points are concentrated in a small ball.

**Definition 4 (Ball)** For a point  $z \in V$  and a radius  $r > 0$ , the ball of radius  $r$  and center  $z$  is the set

$$B(z, r) := \{v : d(z, v) \leq r\}$$

**Lemma 5** For every vertex  $z$ , if we define  $S := B(z, 1/4)$ , then

$$\sum_{u,v} |f_S(u) - f_S(v)| \geq \frac{|S|}{2|V|} \sum_{u,v} d(u, v)$$

PROOF: Our first calculation is to show that the typical value of  $f_S(u)$  is rather large. We note that for every two vertices  $u$  and  $v$ , if we call  $a$  a closest vertex in  $S$  to  $u$ , and  $b$  a closest vertex to  $v$  in  $S$ , we have

$$\begin{aligned} d(u, v) &\leq d(u, a) + d(a, z) + d(z, b) + d(b, v) \\ &\leq f_S(u) + f_S(v) + \frac{1}{2} \end{aligned}$$

and so

$$|V|^2 = \sum_{u, v} d(u, v) \leq 2|V| \cdot \sum_v f_S(v) + \frac{|V|^2}{2}$$

that is,

$$\sum_v f_S(v) \geq \frac{|V|}{2}$$

Now we can get a lower bound to the sum of  $\ell_1$  distances given by the embedding  $f_S(\cdot)$ .

$$\begin{aligned} &\sum_{u, v} |f_S(u) - f_S(v)| \\ &\geq \sum_{u \in S, v \in V} |f_S(v)| \\ &= |S| \sum_v f_S(v) \\ &\geq \frac{1}{2} |S| \cdot |V| \end{aligned}$$

□

This means that if there is a vertex  $z$  such that  $|B(z, 1/4)| = \Omega(|V|)$ , or even  $|B(z, 1/4)| = \Omega(|V|/\sqrt{\log |V|})$ , then we are done.

Otherwise, we will find a set of  $\Omega(|V|)$  vertices such that their average pairwise distances are within a constant factor of their maximum pairwise distances, and then we will work on finding an embedding for such a set of points. (The condition that the average distance is a constant fraction of the maximal distance will be very helpful in subsequent calculations.)

**Lemma 6** *Suppose that for every vertex  $z$  we have  $|B(z, 1/4)| \leq |V|/4$ . Then there is a vertex  $w$  such that, if we set  $S = B(w, 2)$ , we have*

- $|S| \geq \frac{1}{2} \cdot |V|$
- $\sum_{u,v \in S} d(u,v) \geq \frac{1}{8} |S|^2$

PROOF: Let  $w$  be a vertex that maximizes  $|B(w, 2)|$ ; then  $|B(w, 2)| \geq |V|/2$ , because if we had  $|B(u, 2)| < |V|/2$  for every vertex  $u$ , then we would have

$$\sum_{u,v} d(u,v) > \sum_u 2 \cdot (|V - B(u, 2)|) > |V|^2$$

Regarding the sum of pairwise distances of elements of  $S$ , we have

$$\sum_{u,v \in S} d(u,v) > \sum_{u \in S} \frac{1}{4} (|S - B(u, 1/4)|) \geq |S| \cdot \frac{1}{4} \cdot \frac{|S|}{2}$$

□

The proof of the main theorem now reduces to proving the following geometric fact.

**Theorem 7** *Let  $d$  be a negative-type metric over a set  $V$  such that the points are contained in a unit ball and have constant average distance, that is,*

- *there is a vertex  $z$  such that  $d(v, z) \leq 1$  for every  $v \in V$*
- $\sum_{u,v \in V} d(u,v) \geq c \cdot |V|^2$

*Then there are sets  $S, T \subseteq V$  such that*

- $|S|, |T| \geq \Omega(|V|)$ ;
- *for every  $u \in S$  and every  $v \in S$ ,  $d(u,v) \geq 1/O(\sqrt{\log |V|})$*

*where the multiplicative factors hidden in the  $O(\cdot)$  and  $\Omega(\cdot)$  notations depend only on  $c$ .*