

Lecture 27: The ARV Algorithm

Recall our initial assumption that $|B(i, 1/4)| < \frac{n}{4}$ for all $i \in V$. This assumption was justified by showing that if there was some $i \in V$ with

$|\{j \in V : d(i, j) \leq 1/4\}| \geq n/4$ then we could efficiently find $S \subseteq V$ with $S(S) = O(1) \cdot S(G)$.

Our next claim was the following. (This is slightly different from what we did in class.)

Claim. There exists $\theta \in V$ s.t. $|B(\theta, 8)| \geq \frac{3n}{4}$.

Furthermore, let $U = B(\theta, 8) - B(\theta, 1/4)$.

Then $|U| \geq n/2$ and for all $i \in U$, there exist at least $n/4$ vertices $j \in U$ s.t. $d(i, j) > \frac{1}{4}$.

Proof. Suppose no such θ exists. Then for all $i \in V$, we have more than $n/4$ vertices at distance > 8 . So we have $\sum_{i,j \in V} d(i, j) = \underbrace{\sum_{i \in V} (\sum_{j \in V} d(i, j))}_{\text{This is } 2n^2} > n \cdot 8 \frac{n}{4} = 2n^2$

a contradiction. Thus there exists some $\theta \in V$ s.t. $|B(\theta, 8)| \geq 3n/4$. Since $|B(\theta, 1/4)| \leq n/4$ by our initial assumption, for $U = B(\theta, 8) - B(\theta, 1/4)$

$$|U| \geq |B(\theta, 8)| - |B(\theta, 1/4)| \geq \frac{3n}{4} - \frac{n}{4} = \frac{n}{2}$$

Finally pick any $i \in U$. Since $|B(i, 1/4)| < n/4$ while $|U| \geq n/2$, there are $\geq \frac{n}{4}$ $j \in U$ s.t. $d(i, j) > 1/4$. ■

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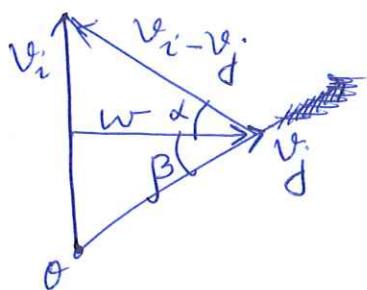
Recall the sets L and R computed in the algorithm.

Theorem. There is a constant $c > 0$ such that $\Pr[|L| \geq cn \text{ and } |R| \geq cn] \geq c$.

Proof. Let us pick $i, j \in U$ such that $d(i, j) > \frac{1}{4}$. We have $\frac{1}{4} \leq \|v_i\|^2 \leq 8$ (recall that we moved the origin to o) and so $\frac{1}{4} \leq \|v_j\|^2 \leq 8$.

Without loss of generality, assume $\|v_i\| \geq \|v_j\|$.

$$\text{Let } w = v_j - \frac{(v_j, v_i) \cdot v_i}{\|v_i\|^2}.$$



We have the following inequalities:

$$\|v_i\|^2 \leq \|v_i - v_j\|^2 + \|v_j\|^2$$

$$\|v_i - v_j\|^2 \leq \|v_i\|^2 + \|v_j\|^2.$$

The first inequality says that the angle between v_j and $v_i - v_j$ is not obtuse and the second says that the angle between v_i and v_j is not obtuse.

Let α and β be as indicated in the picture above. Then $\alpha + \beta \leq \pi/2$.

- if $\alpha \leq \pi/4$ then $\|w\| = \|v_i - v_j\| \cos \alpha$

$$\geq \frac{1}{2} \cdot \frac{1}{\sqrt{2}} = \frac{1}{2\sqrt{2}}$$

(since $d(i, j) = \|v_i - v_j\|^2 > \frac{1}{4}$).

- if $\beta \leq \pi/4$ then $\|w\| = \|v_j\| \cos \beta$

$$\geq \frac{1}{2} \cdot \frac{1}{\sqrt{2}} = \frac{1}{2\sqrt{2}}$$

(since $\|v_j\|^2 \geq \frac{1}{4}$)

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$$\text{Thus } \|w\| \geq \frac{1}{2\sqrt{2}}.$$

Recall that we need to bound $\Pr[i \in L \text{ and } j \in R]$.

Claim. If $r^T v_i \in [-2, -1]$ and $r^T w \geq 3$ then $i \in L$ and $j \in R$.

Proof. We know that $r^T v_i \leq -1$ puts $v_i \in L$.

What we need to show now is that $r^T v_j \geq 1$.

That will put $j \in R$.

Since $v_j = (v_i, v_j) \frac{v_i}{\|v_i\|^2} + w$, taking a dot product with r on both sides,

$$\begin{aligned} (r, v_j) &= (r, v_i) \frac{(v_i, v_j)}{\|v_i\|^2} + (r, w) \\ &\geq -2 \cdot \underbrace{1}_{(since \|v_j\| \leq \|v_i\|)} + 3 = 1. \end{aligned}$$

So we have $\Pr[i \in L \text{ and } j \in R] \geq \Pr[-2 \leq r^T v_i \leq -1 \text{ and } r^T w \geq 3]$

This equals (since v_i & w are orthogonal) $r^T w \geq 3$.

$$\Pr\left[\frac{-2}{\|v_i\|} \leq \frac{r^T v_i}{\|v_i\|} \leq \frac{-1}{\|v_i\|}\right] \cdot \Pr\left[\frac{r^T w}{\|w\|} \geq \frac{3}{\|w\|}\right].$$

Here we used the fact that $r^T x$ and $r^T y$ are independently distributed for vectors x and y orthogonal to each other.

For any unit vector x , $r^T x \sim N(0, 1)$. The prob. that a random variable $\sim N(0, 1)$ takes a value in an $\sqrt{2}(1)$ -sized interval is at least a constant.

Thus $\Pr\left[\frac{-2}{\|v_i\|} \leq \frac{r^T v_i}{\|v_i\|} \leq \frac{-1}{\|v_i\|}\right] = \Omega(1)$ and similarly,
 $\Pr\left[\frac{r^T w}{\|w\|} \geq \frac{3}{\|w\|}\right] = \Omega(1).$

Recall that $|U| \geq n/2$. Also each $i \in U$ has $\geq n/4$ vertices $j \in U$ s.t. $d(i, j) \geq 1/4$. Thus $E[L_U \times R_U]$ is $\Omega(n^2)$, where $L_U = L \cap U$ and $R_U = R \cap U$. This leads to the theorem statement. \blacksquare

A high level overview of the structure theorem. (by Barak & Steurer, 2016)

We know that $\frac{r^T v}{\|v\|} \sim N(0, 1)$. This implies that

$$\Pr[r^T v \geq \alpha] \leq e^{-\alpha^2/\|v\|^2}$$

$$\text{Thus } \Pr[r^T(v_i - v_j) \geq C\sqrt{\ln n}] \leq e^{-C^2 \ln n / \|v_i - v_j\|^2}$$

Since $\|v_i - v_j\|^2 \leq \|v_i\|^2 + \|v_j\|^2 \leq 16$, this probability is at most $e^{-C^2 \ln n / 16} = \frac{1}{n^{C^2/16}}$. This constraint holds for all $i, j \in U$.

So for sufficiently large C , we have $r^T(v_i - v_j) \leq C\sqrt{\ln n}$ for all $i, j \in U$ whp. Then one can show that

$$E\left[\max_{i, j \in U} r^T(v_i - v_j)\right] \leq C\sqrt{\ln n}.$$

Projection lemma. $\frac{1}{\Delta} \cdot \left(\frac{E[|M|]}{n}\right)^3 \leq E\left[\max_{i, j \in U} r^T(v_i - v_j)\right]$

From the projection lemma, for the right choice of constants and $\Delta = \Omega\left(\frac{1}{\sqrt{\log n}}\right)$, we get $\left(\frac{E[|M|]}{n}\right)^3 \leq \left(\frac{C}{2}\right)^6$, which proves the structure theorem.

So the projection lemma implies the structure theorem.