Lec. 23: The gap-Hamming problem (part II)

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#### Summary

In this lecture, we have proved a lower bound of  $\Omega(n)$  for the gap Hamming distance  $(\text{GHD}_n)$  problem. This result was first proved in [CR11] and followed by several other proofs [Vid11, She11]. We have followed the proof given by A.A.Sherstov [She11], which uses the corruption bound from problem set 2 (see Appendix).

**Theorem 23.1** ( [Shel1]).  $R_{\frac{1}{2}}(ORT_n) = \Omega(n)$ .

Corollary 23.2 ( [She11]).  $R_{\frac{1}{2}}(\text{GHD}_n) = \Omega(n).$ 

# 23.1 Proof outline

- 1. To prove  $R_{\frac{1}{3}}(\text{GHD}_n) = \Omega(n)$ , we define another problem called gap orthogonality  $\text{ORT}_n$  and we reduce  $\text{ORT}_n$  to  $\text{GHD}_N$ . It then suffices to prove  $R_{\frac{1}{3}}(\text{ORT}_n) = \Omega(n)$ .
- 2. To prove  $R_{\frac{1}{3}}(\text{ORT}_n) = \Omega(n)$ , by Yao's lemma, it suffices to prove  $D_{\frac{1}{3}}^{\mu}(\text{ORT}_n) = \Omega(n)$  for some  $\mu$ . We choose  $\mu$  to be uniform.
- 3. We use the corruption bound to prove  $R_{\frac{1}{3}}(\text{ORT}_n) = \Omega(n)$ . In order to use this, we need to show that
  - (a)  $\mu(\text{ORT}^{-1}(+1))$  is large (i.e.,  $\Theta(1)$ ), and
  - (b) there exists a small enough  $\varepsilon$  such that any rectangle that is not  $\varepsilon$ -1-corrupted must be small. That is,  $\forall S, T \subseteq \{-1, +1\}^n$ , if  $\mu(\text{ORT}^{-1}(+1) \cap (S \times T)) \leq \varepsilon \mu(\text{ORT}^{-1}(-1) \cap (S \times T))$  then  $\mu(S \times T) \leq \exp(-\Omega(n))$ .

### 23.2 Definitions

The gap orthogonality problem is defined as follows. Let  $x, y \in \{-1, +1\}^n$ . The function  $ORT_n(x, y)$  is defined as

$$ORT_n(x,y) = \begin{cases} -1 & \text{if } |\langle x,y\rangle| \le \frac{\sqrt{n}}{8} \\ +1 & \text{if } |\langle x,y\rangle| \ge \frac{\sqrt{n}}{4} \end{cases}$$

The general gap Hamming distance problem  $\text{GHD}_{n,t,g}$  is defined as follows. Let  $x, y \in \{-1, +1\}^n$ . The function  $\text{GHD}_{n,t,g}(x, y)$  is defined as,

$$\operatorname{GHD}_{n,t,g}(x,y) = \begin{cases} -1 & \text{if } \langle x,y \rangle \leq t-g \\ \\ +1 & \text{if } \langle x,y \rangle \geq t+g \end{cases}$$

Note that both  $ORT_n$  and  $GHD_{n,t,g}$  are partial functions. The function  $GHD_N$  is just  $GHD_{N,0,\sqrt{N}}$ .

## 23.3 Proof Details

First we establish the first stage from the proof outline.

Claim 23.3. ORT<sub>n</sub> reduces to GHD<sub>N</sub>, where N = O(n).

*Proof.* In the last lecture, we saw that  $\text{GHD}_{n,t,g}$  reduces to  $\text{GHD}_N$ , using a padding technique. The idea is to first reduce  $\text{ORT}_n$  to  $\text{GHD}_{n,t,g}$  which in turn reduces to  $\text{GHD}_N$ .

Let  $x, y \in \{-1, +1\}^n$  which satisfies the promise (i.e.,  $ORT_n(x, y)$  is defined). Then  $ORT_n(x, y)$  can be solved using 2 calls to  $GHD_{n,t,q}(x, y)$ , as follows.

$$\operatorname{ORT}_{n}(x,y) = \begin{cases} -1 & \text{if } \operatorname{GHD}_{n,\frac{3\sqrt{n}}{16},\frac{\sqrt{n}}{16}}(x,y) = -1 \ \& \ \operatorname{GHD}_{n,-\frac{3\sqrt{n}}{16},\frac{\sqrt{n}}{16}}(x,y) = +1 \\ +1 & \text{if } \operatorname{GHD}_{n,-\frac{3\sqrt{n}}{16},\frac{\sqrt{n}}{16}}(x,y) = -1 \ \& \ \operatorname{GHD}_{n,\frac{3\sqrt{n}}{16},\frac{\sqrt{n}}{16}}(x,y) = +1 \end{cases}$$

We know that  $\text{GHD}_{m,t,g}$  can be reduced to  $\text{GHD}_N$  using padding. Examining the parameters m, t, g in the calls here relative to n, we see that  $N \in O(n)$  suffices.

Thus, we have shown that  $ORT_n$  reduces to  $GHD_N$ .

Now consider the Stage 3(a) from the proof outline.

## Claim 23.4. $\mu(ORT^{-1}(-1)) = \Theta(1).$

*Proof.* Let  $x, y \in \{-1, +1\}^n$ . We know that  $\langle x, y \rangle = n - 2\Delta(x, y)$  (where  $\Delta(.,.)$  is the Hamming distance). Note that if  $\Delta(x, y) \in [\frac{n}{2} - \frac{\sqrt{n}}{8}, \frac{n}{2} + \frac{\sqrt{n}}{8}]$  then  $|\langle x, y \rangle| \leq \frac{\sqrt{n}}{8}$ . For each fixed x, we count number of y's such that  $|\langle x, y \rangle| \leq \frac{\sqrt{n}}{8}$ . Using the fact that there is an absolute constant c > 0 such that for k close to n/2,  $\binom{n}{k} \geq \frac{2^n}{c\sqrt{n}}$ , we see that

Number of 
$$y's$$
 with  $\left[|\langle x,y\rangle| \le \frac{\sqrt{n}}{8}\right] = \sum_{k=\frac{n}{2}-\frac{\sqrt{n}}{8}}^{\frac{n}{2}+\frac{\sqrt{n}}{8}} \binom{n}{k} \ge \frac{\sqrt{n}}{4} \cdot \frac{2^n}{c\sqrt{n}} = \frac{2^n}{4c}$ 

Thus, the total number of  $x, y \in \{-1, +1\}^n$  such that  $|\langle x, y \rangle| \leq \frac{\sqrt{n}}{8}$  is at least  $\frac{2^{2n}}{4c}$ . Since  $\mu$  is the uniform distribution, we have

$$\mu(\text{ORT}^{-1}(-1)) \ge \frac{2^{2n}}{4c} \cdot \frac{1}{4^n} = \frac{1}{4c}$$

The rest of the lecture is devoted to proving Stage 3(b) of the proof outline; that is, showing that any rectangle that is not 1-corrupted must in fact be small. We proceed via the following steps.

**Step 0.** For parameters  $\varepsilon, \alpha$  to be chosen later, let  $\rho = 2/2^{\alpha n}$ . Assume to the contrary that some rectangle  $R = S \times T$  is not 1-corrupted and is large.

Large:  $\mu(R) \ge \rho$ . Since  $\mu(R) \le |S|/2^n$ , we can then conclude that  $|S| \ge \rho 2^n =$  $2 \cdot 2^{(1-\alpha)n}$ . Similarly, we can conclude that  $|T| \ge 2 \cdot 2^{(1-\alpha)n}$ .

Not 1-corrupted:  $\mu(\text{ORT}^{-1}(+1) \cap (S \times T)) \leq \varepsilon \mu(\text{ORT}^{-1}(-1) \cap (S \times T)) \leq \varepsilon \mu(S \times T).$ 

- **Step 1.** Using the assumption that R has a "very high" density of -1s (because it is not 1-corrupted), find  $A \subseteq S$  with a "fairly high" density of -1s (to be formally defined below) in each row, such that  $|A| \geq \frac{|S|}{2}$ .
- **Step 2.** Using the assumption that S is large, and hence that A is large, show that there exists a set  $A' \subseteq A$  of  $k = \frac{n}{10}$  "near-orthogonal" vectors  $x_1, \ldots, x_k$ .
- **Step 3.** Show that for any set W of m near-orthogonal vectors  $x_1, ..., x_m$ , a random y is, with high probability, far from orthogonal to at least one  $x_i$  (and so the corresponding  $ORT(x_i, y)$  is +1).
- **Step 4.** Since  $A' \subseteq A$ ,  $A' \times T$  has a fairly high density of -1s. Find a  $B \subseteq T$  with a moderately high density of -1s within  $A' \times B$  such that |B| > |T|/3. Using the assumption that T is large, conclude that B is quite large. From Steps 2,3, conclude that B cannot be quite large. This gives a contradiction.

#### **Proof of Step 1:**

Define the set A as follows:

$$A = \{x \in S \mid \# +1 \text{'s in } \{x\} \times T \le 2\varepsilon |T|\}$$

An averaging argument shows that  $|A| \ge \frac{|S|}{2} \ge 2^{(1-\alpha)n}$ . (If  $|A| < \frac{|S|}{2}$ , then there are at least  $(\frac{|S|}{2} + 1)$  rows in S such that each row has more than  $2\varepsilon |T|$  entries as +1. Thus, we have more than  $\varepsilon |T||S|$  entries as +1 in the rectangle  $S \times T$ , contradicting the assumption that R is not 1-corrupted.)

#### **Proof of Step 2:**

Say that a set of vectors  $x_1, ..., x_k$  in  $\{+1, -1\}^n$  is near-orthogonal if for each  $i \in [k-1]$ , the vector  $x_{i+1}$  is almost orthogonal to (has a very small projection on) the subspace spanned by  $x_1, \ldots, x_i$ . Specifically, for each i,  $||\operatorname{proj}_{\operatorname{span}\{x_1, \ldots, x_i\}} x_{i+1}|| \leq \frac{\sqrt{n}}{3}$ . We prove the following. Let  $A \subseteq \{-1, +1\}^n$ . For a sufficiently small constant  $\alpha > 0$ ,

which will be specified at the end of this Step, if  $|A| > 2^{(1-\alpha)n}$  then A contains a set A' of  $k = \lfloor \frac{n}{10} \rfloor$  near-orthogonal vectors  $x_1, ..., x_k$ .

Pick  $x_1 \in A$  aribitrarily. Using Talagrand's inequality (see Appendix), a randomly

picked x is unlikely to have a large projection on the space spanned by  $x_1$ .

$$\begin{split} \Pr_x \left[ ||\operatorname{proj}_{x_1} x|| > \frac{\sqrt{n}}{3} \right] &\leq \Pr_x \left[ \ ||\operatorname{proj}_{x_1} x|| - 1| > \frac{\sqrt{n}}{3} - 1 - c + c \right] \\ &< \exp(-\frac{(\frac{\sqrt{n}}{3} - c - 1)^2}{c}) \\ &\leq 2^{-\beta_1 n} \quad \text{for some constant } \beta_1 \text{ that depends only on } c. \end{split}$$

Also, since |A| is large, a randomly picked x is likely to be in A with good probability;  $\Pr_{x \in r\{-1,+1\}^n}[x \in A] > 2^{-\alpha n}$ . Putting these two together, a randomly picked x is likely to be both in A and have a small projection on the space spanned by  $x_1$ . That is,

$$\begin{aligned} \Pr_x \left[ ||\operatorname{proj}_{x_1} x|| &\leq \frac{\sqrt{n}}{3} \text{ AND } x \in A \setminus \{x_1\} \right] &= 1 - \Pr_x \left[ ||\operatorname{proj}_{x_1} x|| > \frac{\sqrt{n}}{3} \text{ OR } x \notin A \setminus \{x_1\} \right] \\ &\geq 1 - [2^{-\beta_1 n} + 1 - 2^{-\alpha n}] \quad (\text{union bound}) \\ &= 2^{-\alpha n} - 2^{-\beta_1 n} \end{aligned}$$

Let  $\alpha < \beta_1$ . Then the RHS is strictly positive. Therefore, there exists  $x_2 \in A$  that is near orthogonal to  $x_1$ . Let's fix it.

We continue adding vectors this way. If  $x_1, \ldots, x_j$  span space V, then  $\dim(V) \leq j$ .

$$\begin{aligned} \Pr_x \left[ ||\operatorname{proj}_{\operatorname{span}(x_1,\dots,x_j)} x|| &> \frac{\sqrt{n}}{3} \right] &\leq \Pr_x \left[ \ ||\operatorname{proj}_{\operatorname{span}(x_1,\dots,x_j)} x|| - \sqrt{j}| &> \frac{\sqrt{n}}{3} - \sqrt{j} - c + c \right] \\ &< \exp(-\frac{(\frac{\sqrt{n}}{3} - c - \sqrt{j})^2}{c}) \\ &\leq \exp(-\beta_j n) \end{aligned}$$

Here  $\beta_j$  is a constant depending only on c and j. By a similar argument as above, a random x will, with non-zero probability, be a new element of A and have a small projection with respect to the already chosen vectors. So we can find  $x_{j+1}$ .

Choose  $\alpha < \beta_j$  for  $1 \le j \le k = \frac{n}{10}$ . In fact, let  $\alpha$  be slightly smaller than the smallest  $\beta_j$ , to account for the fact that each  $x_{j+1}$  must be chosen not just from A but from  $A \setminus \{x_1, \ldots, x_j\}$ . Then we can choose  $k = \lfloor \frac{n}{10} \rfloor$  near-orthogonal vectors in A.

#### **Proof of Step 3:**

Let  $m \leq n/10$ , and fix any set W of vectors  $x_1, ..., x_m \in \{-1, +1\}^n$  that are near-orthogonal. We want to show that with high probability, a random y is far-from-orthogonal to at least one  $x_i$ . Considering the complement event, we want to show that

$$\Pr_{y}\left[\forall i, |\langle y, x_i \rangle| \leq \frac{\sqrt{n}}{4}\right] \in \exp(-\Omega(m))$$

Consider the  $m \times n$  matrix M whose  $i^{th}$  row is  $x_i$  (We omit the notation for transpose; clear from the context). We want to show that  $\Pr_y[||My||_{\infty} \leq \frac{\sqrt{n}}{4}] \leq \exp(-\Omega(m))$ . Since  $||My||_{\infty} \leq ||My|| \leq \sqrt{m} ||My||_{\infty}$ , it suffices to instead prove that

$$\Pr_{y}\left[||My||^{2} \ge \frac{mn}{16}\right] \ge 1 - \exp(-\Omega(m)).$$

Consider a singular value decomposition SVD of M as  $M = UDV^t$  where U, V are unitary matrices of order m and n respectively, and D is a "rectangular diagonal"  $m \times n$ matrix with diagonal entries  $\sigma_1 \geq ... \geq \sigma_m$ . D is uniquely defined by M. Let  $u_i, v_i$ denote the columns of U, V respectively. Then for any vector y,  $||My||^2 = (My)^t (My) = (UDV^t y)^t (UDV^t y) = (DV^t y)^t (DV^t y) = \sum_{i=1}^m \sigma_i^2 \langle v_i, y \rangle^2$ .

To show that  $||My||^2$  is large for many y, we prove the following:

Claim 23.5. Many  $\sigma_i$  are large. Specifically,  $\sigma_1 \geq \sigma_2 \geq ... \geq \sigma_{\lceil \frac{m}{4} \rceil} \geq 0.51\sqrt{n}$ .

First, let us see why proving this claim is enough. Let  $V = \{v_i : \sigma_i \ge 0.51\sqrt{n}\}$ . By the claim, dim span $(V) \ge \frac{m}{4}$ . For every vector y,

$$||My||^{2} = \sum_{i=1}^{m} \sigma_{i}^{2} \langle y, v_{i} \rangle^{2} \ge (0.51\sqrt{n})^{2} \sum_{v_{i} \in V} \langle y, v_{i} \rangle^{2} \ge 0.26n ||\operatorname{proj}_{\operatorname{span}(V)} y||^{2}$$

Using Talagrand's inequality, we can now conclude that

$$\Pr_{y}\left[||My||^{2} \ge \frac{mn}{16}\right] \ge \Pr_{y}\left[||\operatorname{proj}_{\operatorname{span}(V)}y||^{2} \ge \frac{m}{16 \times 0.26}\right] \ge 1 - \exp(-\Omega(m))$$

*Proof.* (Of Claim 23.5) We now prove the claim. First, we get a set of orthogonal vectors  $x'_1, ..., x'_m$  from  $x_1, ..., x_m$ : For each *i*, define  $x'_i = x_i - \text{proj}_{\text{span}(x_1,...,x_{i-1})} x_i$ . (These are the vectors that would be returned by the Gram-Schmidt procedure.) Then we can show (see below) that  $n \ge ||x'_i||^2 \ge \frac{8n}{9}$ .

$$\begin{aligned} ||x_i||^2 &= ||x_i'||^2 + ||\operatorname{proj}_{\operatorname{span}(x_1,\dots,x_{i-1})} x_i||^2 & (\text{since these are orthogonal components of } x_i) \\ n &= ||x_i'||^2 + ||\operatorname{proj}_{\operatorname{span}(x_1,\dots,x_{i-1})} x_i||^2 & (\because x_i \in \{-1,+1\}^n \ ) \\ n &\leq ||x_i'||^2 + \frac{n}{9} & (\because ||\operatorname{proj}_{\operatorname{span}(x_1,\dots,x_{i-1})} x_i|| \leq \frac{\sqrt{n}}{3}, \text{ by near-orthogonality } ) \\ n &\geq ||x_i'||^2 & (\because ||\operatorname{proj}_{\operatorname{span}(x_1,\dots,x_{i-1})} x_i|| \geq 0 \end{aligned}$$

Let M' denote the  $m \times n$  matrix whose  $i^{th}$  row is  $x'_i$ .

Now consider the Frobenius norm of M, defined as  $||M||_F = \sqrt{\sum_{i,j} M_{ij}^2}$ . We will use the following proposition, to be proved later.

**Proposition 23.6.** For all N,  $\sigma_{r+1}(M) \ge \frac{1}{\operatorname{rank}(M)-r} \left( \frac{\langle M, N \rangle}{\sigma_1(N)} - ||M||_F \sqrt{r} \right)$ 

Using the above proposition with N being the Gram-Schmidt matrix M', we get

$$\langle x_i, x_i' \rangle = ||x_i'||^2 \ge \frac{8n}{9}; \qquad \langle M, M' \rangle \ge \frac{8mn}{9}$$
$$\sigma_1(M') \le \max_i ||x_i'|| = \sqrt{n}; \qquad ||M||_F = \sqrt{mn}$$
Hence
$$\sigma_{r+1}(M) \ge \frac{1}{m-r} \left(\frac{\frac{8}{9}mn}{\sqrt{n}} - \sqrt{mn}\sqrt{r}\right)$$

Hence for  $r = \lceil \frac{m}{4} \rceil$ ,  $\sigma_{r+1}(M) \ge 0.51\sqrt{n}$ .

All that remains in this Step is to prove Proposition 23.6).

#### *Proof.* (Of Proposition 23.6)

The largest r singular values satisfy

$$\sigma_1 + \dots + \sigma_r \leq \sqrt{r} \sqrt{(\sigma_1^2 + \dots + \sigma_r^2)} \text{ by Cauchy-Schwartz} \\ \leq \sqrt{r} ||M||_F$$

The remaining singular values satisfy

$$\sigma_{r+1} + \dots + \sigma_m \leq (\operatorname{rank}(M) - r)\sigma_{r+1}$$
Also, by SVD,  $\sum_{i=1}^m \sigma_i \geq \frac{\langle M, N \rangle}{\sigma_1(N)}$   $(\because \langle M, N \rangle = \sum \sigma_i u_i^T N v_i \leq \sum \sigma_i \sigma_1(N))$ 
Hence  $||M||_F \sqrt{r} + (\operatorname{rank}(M) - r)\sigma_{r+1} \geq \sigma_1 + \dots + \sigma_r \geq \frac{\langle M, N \rangle}{\sigma_1(N)}$ 
 $\therefore \sigma_{r+1}(M) \geq \frac{1}{\operatorname{rank}(M) - r} \left(\frac{\langle M, N \rangle}{\sigma_1(N)} - ||M||_F \sqrt{r}\right)$ 

### **Proof of Step 4:**

Recall that  $A' = \{x_1, \ldots, x_k\}$  is contained in A. Hence by choice of A, each row of  $A' \times T$  has at most  $2\varepsilon |T|$  entries that are +1.

Define the set B as follows:

$$B = \{y \in T \mid \# +1\text{'s in } A' \times \{y\} \le 3\varepsilon |A'|\}$$

An averaging argument similar to that in Step 1 shows that  $|B| \ge \frac{|T|}{3} \ge (\frac{2}{3})2^{(1-\alpha)n}$ .

We now give an upper bound on the size of B. If  $y \in B$ , then we can pick a set  $A'' \subseteq A'$ , of size exactly  $(1-3\varepsilon)|A'| = (1-3\varepsilon)k$ , such that  $A'' \times \{y\}$  has only -1s. Let W be a subset of A' of size exactly  $(1-3\varepsilon)k$ , and define the set  $B_W \subseteq T$  as follows:

$$B_W = \{ y \in T \mid W \times \{y\} \text{ has only } -1s \}$$
  
Then  $B \subseteq \bigcup_{W \subseteq A'; |W| = (1-3\varepsilon)k} B_W$   
and hence  $|B| \leq \sum_{W \subseteq A'; |W| = (1-3\varepsilon)k} |B_W|$ 

By Step 3, for any such W,  $\Pr[y \in B_W] \in \exp(-\Omega(|W|)) = \exp(-\Omega((1-3\varepsilon)k))$ . Hence  $|B_W| \leq 2^n \exp(-\Omega((1-3\varepsilon)k))$ . The number of choices for W is  $\binom{k}{(1-3\varepsilon)k} = \binom{k}{3\varepsilon k}$ . Hence

$$|B| \le \binom{k}{3\varepsilon k} 2^n \exp(-\Omega((1-3\varepsilon)k)) \le 2^{-H(3\varepsilon)k} 2^n \exp(-\Omega((1-3\varepsilon)k))$$

By choosing suitable  $\alpha, \varepsilon$ , we can see that the bounds on B

$$\left(\frac{2}{3}\right)2^{(1-\alpha)n} \le |B| \le 2^{-H(3\varepsilon)k}2^n \exp(-\Omega((1-3\varepsilon)k))$$

are not simultaneously possible.

# References

- [CR11] AMIT CHAKRABARTI and ODED REGEV. An optimal lower bound on the communication complexity of gap-Hamming-distance. In Proc. 43rd ACM Symp. on Theory of Computing (STOC), pages 51-60. 2011. arXiv:1009.3460, eccc:TR10-140, doi:10.1145/1993636. 1993644.
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## 23.4 Appendix

### 23.4.1 Corruption Bound

**Lemma 23.7.** Let  $f : X \times Y \to \{0, 1\}$  be a Boolean function and  $\mu$  be a probability distribution on  $X \times Y$  such that for every rectangle  $R = S \times T \subseteq X \times Y$  with  $\mu(R) > \rho$ , we have  $\mu(R \cap f^{-1}(1)) > \varepsilon \cdot \mu(R \cap f^{-1}(0))$ . Then, for every  $\delta > 0, 2^{R_{\delta}(f)} \ge \frac{1}{\rho} \cdot (\mu(f^{-1}(0) - \frac{\delta}{\varepsilon}))$ .

### 23.4.2 Talagrand's Inequality

Let  $V \subseteq \mathbb{R}^n$  be a linear subspace of dimension d. Talagrand's inequality states that for a randomly chosen  $x \in_r \{-1, +1\}^n$ , with high probability, the projection of x on V is of length close to  $\sqrt{d}$ . Formally:

There exists a c > 0 such that  $\forall t > 0$ ,

$$\Pr_{x \in {}_{r}\{-1,+1\}^{n}} \left[ \mid || \operatorname{proj}_{V} x|| - \sqrt{\dim V} \mid > t + c \right] < 4 \exp\left(-\frac{t^{2}}{c}\right)$$