27. Multiparty communication complexity of disjointness

Lecturer: Jaikumar Radhakrishnan

Scribe: Swagato Sanyal

In this lecture we shall derive a lower bound on multiparty communication complexity of the Disjointness problem, in the "Number on forehead" model introduced in the last lecture. Recall that earlier we proved lower bound of the complexity of the same problem in "Number in hand" model.

27.1 Multiparty Set Disjointness Problem

We have a universe [n] and we have k subsets of the universe b_1, \ldots, b_k , each given as a bit vector of size n. In such a bit vector, the *i*-th entry is 1 *iff* the element *i* is a member of that set. There are k players, and b_i is the input to *i*-th player. According to our model, each player *i* has access to every b_j such that $j \neq i$. The goal is to decide whether the intersection of b_1, \ldots, b_k is non-empty.

$$\mathsf{DISJ}_{\mathsf{n},\mathsf{k}}(b_1,\ldots,b_k) = \begin{cases} 1 & (\text{if } \cap_{i=1}^k b_i = \varphi) \\ 0 & (\text{otherwise}) \end{cases}$$

In this lecture we shall prove the following theorem by Sherstov:

Theorem 27.1. $R_{1/3}(\text{DISJ}_{n,k}) = \Omega\left(\begin{array}{c} n \\ \frac{1}{4^k} \end{array}\right)^{\frac{1}{4}}$.

We will move to the -1, +1 world from the 0, 1 world. Thus -1 and +1 will stand for 1 and 0 respectively. For proving the above theorem we shall use the following claim without proof. AND_n is the *n* variable logical AND function.

Claim 27.2. There exists a constant $\delta_0 > 0$ such that any polynomial that 1/3-approximates AND_n has degree at least $\delta_0 \sqrt{n}$.

For any function g, let $deg_{\delta}(g)$ denote the degree of a polynomial with least degree which δ -approximates g. The broad idea of the proof is showing that if there is a cheap communication protocol for multiparty disjointness, then there is a low degree polynomial that closely approximates AND_n .

For some r, define a function $F : \{0, 1\}^{nrk} \longrightarrow \{0, 1\}$ as follows. Think of the input as n matrices, each of dimension $r \times k$. Treat each matrix as an input to $\mathsf{DISJ}_{\mathsf{r},\mathsf{k}}$. Apply the function $\mathsf{DISJ}_{\mathsf{r},\mathsf{k}}$ to each matrix to get one bit for each of them. Finally define F to be the logical AND of those bits. Note that F is actually disjointness with k players and a larger universe. Let $deg_{\delta}(\mathsf{AND}_{\mathsf{n}}) = d_{\delta}$. We will prove the following theorem:

Theorem 27.3.

$$\forall \varepsilon, \delta \ge 0, 2^{R_{\varepsilon}(F)} \ge (\delta - 2\varepsilon) \left(\begin{array}{c} \frac{d_{\delta} \sqrt{r}}{2^{k} e n} \end{array} \right)^{d_{\delta}}.$$

27.2 Proof of Theorem 27.3

Let Π be a ε -error randomized protocol for F. Let $\mu_{r,k}^{+1}$ and $\mu_{r,k}^{-1}$ be two distributions (to be fixed later) on $\{0,1\}^{rk}$ supported on negative and positive instances of $\mathsf{DISJ}_{r,k}$ respectively. Let $\mu = \frac{\mu_{r,k}^{+1} + \mu_{r,k}^{-1}}{2}$. We obtain a randomized process for computing AND_n from Π as follows: Given n bits z_1, \dots, z_n (by bits we mean +1, -1 values), we draw $\mathsf{DISJ}_{r,k}$ instances X_1, \dots, X_n from probability distributions $\mu_{r,k}^{z_1}, \dots, \mu_{r,k}^{z_n}$ respectively. Clearly $\Pi(X_1, \dots, X_n) =$ $AND(X_1, \dots, X_n)$ with probability at least $1 - \varepsilon$. Let $P(z_1, \dots, z_n)$ be the expected value computed by the randomized process on inputs z_1, \dots, z_n (the expectation is over the random choices that the process makes). Let $\sum_{S \subseteq [n]} \gamma_S \prod_{i \in S} z_i$ be the Fourier expansion of

 $P(z_1, \dots, z_n)$. For each $S \subseteq [n]$, γ_S is $E_{z_1 \sim U\{0,1\}}, \dots, z_n \sim U\{0,1\}}[P(z_1, \dots, z_n)\Pi_{i \in S} z_i]$. Let $\Pi(X)$ be the expected output of Π on X. Check that,

$$\gamma_S = E_{X_1 \sim \mu, \cdots, X_n \sim \mu} [\Pi(X_1, \cdots, X_n) \prod_{i \in S} \mathsf{DISJ}_{\mathsf{r},\mathsf{k}}(X_i)]$$
(27.2.1)

In the last lecture we saw that in NOF model, each leaf of the protocol tree of any deterministic protocol corresponds to intersection of k cylinders. The *i*-th cylinder is a set of inputs, where the membership does not depend on the *i*-th input. For every cylinder intersection \mathcal{K} , let $\mathcal{K}()$ denote the indicator function of membership in \mathcal{K} . Let \mathcal{C} be the set of all cylinder intersections of a deterministic protocol Ψ . Then Ψ can be written as:

$$\Psi(X) = \sum_{\mathcal{K} \in \mathcal{C}} a_{\mathcal{K}} \mathcal{K}(X)$$

where X is the input to the protocol, and $a_{\mathcal{K}}$ is the output (i.e. a +1/-1 value) of Ψ when X belongs to the cyliner intersection \mathcal{K} . Since a randomized protocol's output is a probability distribution over outputs of some deterministic protocols, we can write

$$\Pi(X) = \sum_{\mathcal{K} \in \mathcal{C}} a_{\mathcal{K}} \mathcal{K}(X)$$

where $\Pi(X)$ is the expected output of Π on X, and $a_{\mathcal{K}}$ is the expected output of Π when X is in \mathcal{K} . Both expectations are over internal randomizations of Π . $a_{\mathcal{K}}$ is a real between -1 and +1, and $\sum_{\mathcal{K}} |a_{\mathcal{K}}| \leq 2^c$, where c is the communication complexity of Π . Thus from 27.2.1 we have

$$\gamma_{S} = \sum_{\mathcal{K} \in \mathcal{C}} a_{\mathcal{X}} E_{X_{1} \sim \mu, \cdots, X_{n} \sim \mu} [\mathcal{X}(X_{1}, \cdots, X_{n}) \prod_{i \in S} \mathsf{DISJ}_{\mathsf{r},\mathsf{k}}(X_{i})]$$

Now we will analyze how well does the polynomial $\sum_{S \subseteq [n], |S| < d_{\delta}} \gamma_S . \prod_{i \in S} z_i$ approximate $P(z_1, \dots, z_n)$.

Let the error on input z_1, \dots, z_n resulted because of dropping all monomials with degree greater than d_{δ} be $\varepsilon'(z_1, \dots, z_n)$. Then,

$$\varepsilon'(z_1,\cdots,z_n) = \sum_{\mathcal{X}\in\mathcal{C}} \sum_{S\subseteq[n],|S|\geq d_{\delta}} a_{\mathcal{X}} E_{X_1\sim\mu,\cdots,X_n\sim\mu}[\mathcal{X}(X_1,\cdots,X_n)\prod_{i\in S} \mathsf{DISJ}_{\mathsf{r},\mathsf{k}}(X_i)]\prod_{i\in S} z_i$$

$$\leq \sum_{\mathcal{X}\in\mathcal{C}} \sum_{S\subseteq[n],|S|\geq d_{\delta}} a_{\mathcal{X}} E_{X_{1}\sim\mu,\cdots,X_{n}\sim\mu} [\mathcal{X}(X_{1},\cdots,X_{n})\prod_{i\in S} \mathsf{DISJ}_{\mathsf{r},\mathsf{k}}(X_{i})]$$

$$\leq \sum_{\mathcal{X}\in\mathcal{C}} |a_{\mathcal{X}}| \sum_{S\subseteq[n],|S|\geq d_{\delta}} |E_{X_{1}\sim\mu,\cdots,X_{n}\sim\mu} [\mathcal{X}(X_{1},\cdots,X_{n})\prod_{i\in S} \mathsf{DISJ}_{\mathsf{r},\mathsf{k}}(X_{i})]|$$

To bound this sum, we use a lemma which we will prove in the next lecture. Let $\alpha(s) = \left(\frac{2^{k-1}}{\sqrt{r}}\right)^s$.

Lemma 27.4. There exist distributions $\mu_{r,k}^{+1}$ and $\mu_{r,k}^{-1}$, such that for all $S \subseteq [n]$ with |S| = s, $|E_{X_1 \sim \mu, \cdots, X_n \sim \mu}[\mathcal{X}(X_1, \cdots, X_n) \prod_{i \in S} \mathsf{DISJ}_{\mathsf{r},\mathsf{k}}(X_i)]| \leq \alpha(s).$

For the choice of $\mu_{r,k}^{+1}$ and $\mu_{r,k}^{-1}$ given by the lemma, this gives us

$$\varepsilon'(z_1, \cdots, z_n) \le \sum_{\mathcal{X}} |a_{\mathcal{X}}| \sum_{s=d}^n \alpha(s) \cdot \binom{n}{s}$$
$$\le \sum_{\mathcal{X}} |a_{\mathcal{X}}| \sum_{s=d}^n \left(\frac{2^{k-1} \cdot en}{\sqrt{rs}}\right)^s$$
$$\le \sum_{\mathcal{X}} |a_{\mathcal{X}}| \sum_{s=d}^n \left(\frac{2^{k-1} \cdot en}{\sqrt{rd}}\right)^s$$

If $\frac{2^{k-1} \cdot en}{\sqrt{rd}} < 1$ the sum is geometrically decreasing, and hence bounded above by $2^c \left(\frac{2^k \cdot en}{\sqrt{rd}}\right)^d$. If $\frac{2^{k-1} \cdot en}{\sqrt{rd}} \ge 1$, the quantity $2^c \left(\frac{2^k \cdot en}{\sqrt{rd}}\right)^d$ is at least 1 and hence bounds the error from above. We know that the randomized process approximates AND within error ε . So the expected output of the randomized process (denoted by P()) is within an additive error of 2ε of AND_n. Thus the above polynomial approximates AND within error $2\varepsilon + 2^c \left(\frac{2^k \cdot en}{\sqrt{rd}}\right)^d$. But since the polynomial is of degree less than d_{δ} , it should not approximate AND better than δ . This gives us

$$2\varepsilon + 2^{c} \left(\frac{2^{k} \cdot en}{\sqrt{rd}}\right)^{d} \ge \delta$$
$$\Rightarrow 2^{c} > (\delta - 2\varepsilon) \left(\frac{d_{\delta}\sqrt{r}}{2^{k} en}\right)^{d_{\delta}}$$

This proves Theorem 27.3.

27.3 Proof of Theorem 27.1

Putting $\delta = 1/3$ gives us $d_{\delta} \ge \delta_0 \sqrt{n}$ (claim 27.2). Take $r = 4^k n$. From Theorem 27.1 $2^{R_{1/9}(F)} = 2^{\Omega(\sqrt{n})}$.

Note that F is disjointness on inputs of size $N = 4^k n^2$. Thus,

$$2^{R_{1/9}(F)} = 2^{\left(\frac{N}{4^k}\right)^{1/4}}$$

This proves Theorem 27.1 as we can improve the error from 1/3 to 1/9 by increasing communication by a constant factor.