



# Multiparty Communication Complexity of $AC^0$

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## Abstract

We prove  $n^{\Omega(1)}$  lower bounds on the multiparty communication complexity of  $AC^0$  functions in the number-on-forehead (NOF) model for up to  $\Theta(\log n)$  players. These are the first lower bounds for any  $AC^0$  function for  $\omega(\log \log n)$  players. In particular we show that there are families of depth 3 read-once  $AC^0$  formulas having  $k$ -player randomized multiparty NOF communication complexity  $n^{\Omega(1)}/2^{O(k)}$ . We show similar lower bounds for depth 4 read-once  $AC^0$  formulas that have nondeterministic communication complexity  $O(\log^2 n)$ , yielding exponential separations between  $k$ -party nondeterministic and randomized communication complexity for  $AC^0$  functions.

As a consequence of the latter bound, we obtain an  $n^{\Omega(1/k)}/2^{O(k)}$  lower bound on the  $k$ -party NOF communication complexity of set disjointness. This is non-trivial for up to  $\Theta(\sqrt{\log n})$  players which is significantly larger than the up to  $\Theta(\log \log n)$  players allowed in the best previous lower bounds for multiparty set disjointness given by Lee and Shraibman [LS08] and Chattopadhyay and Ada [CA08] (though our complexity bounds themselves are not as strong as those in [LS08, CA08] for  $o(\log \log n)$  players).

We derive these results by extending the  $k$ -party generalization in [CA08, LS08] of the pattern matrix method of Sherstov [She07, She08]. Using this technique, we derive a new sufficient criterion for strong communication complexity lower bounds based on functions having many diverse subfunctions that do not have good low-degree polynomial approximations. This criterion guarantees that such functions have orthogonalizing distributions that are “max-smooth” as opposed to the “min-smooth” orthogonalizing distributions used by Razborov and Sherstov [RS08] to analyze the sign-rank of  $AC^0$ .

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# 1 Introduction

Recently, Sherstov introduced the so-called pattern matrix method to derive discrepancy bounds [She07, She08] yielding a new strong method for obtaining lower bounds for 2-party quantum communication complexity. His method was then generalized for  $k \geq 2$  players [Cha07, CA08, LS08] to yield the first lower bounds for the general multiparty number-on-forehead communication complexity of set disjointness for more than 2 players, improving a long line of research on the problem. The communication lower bound for  $k$  players is  $\Omega(n^{\frac{1}{2k}}/2^{2^{\Theta(k)}})$  which yields a non-trivial separation between randomized and nondeterministic  $k$ -party models for  $k \leq \epsilon \log \log n$  for some constant  $\epsilon > 0$ . This separation between randomized and nondeterministic communication complexity was extended by David and Pitassi and David, Pitassi, and Viola to  $\Omega(\log n)$  players for significantly more complex functions than disjointness that based on pseudorandom generators [DPV]. Their construction uses a more complex criterion than the simple masking version of the pattern matrix method used in [CA08]. Set disjointness is an  $\text{AC}^0$  function and David, Pitassi, and Viola asked the question of whether one could prove a separation for  $\Omega(\log n)$  players using an  $\text{AC}^0$  function or even whether one could prove any non-trivial lower bound for  $\omega(\log \log n)$  players for any  $\text{AC}^0$  function since their functions are also only in  $\text{AC}^0$  for  $k = O(\log \log n)$ .

We resolve this question positively by showing there is a read-once function in  $\text{AC}_3^0$  that has  $n^{\Omega(1)}$  communication complexity for  $k = \Omega(\log n)$  players. Moreover there is a read-once function in  $\text{AC}_4^0$  that for  $\Omega(\log n)$  players has nondeterministic communication complexity  $O(\log^2 n)$  and randomized communication complexity  $n^{\Omega(1)}$ ; i.e.,  $(\text{NP}_{k(n)}^{\text{cc}} - \text{BPP}_{k(n)}^{\text{cc}}) \cap \text{AC}_4^0 \neq \emptyset$  for  $k(n) \leq \delta \log n$  for some explicit constant  $\delta > 0$ . Our method significantly improves the power of the pattern matrix method for proving strong communication complexity lower bounds.

As a consequence of the lower bound for the function we use to separate  $\text{NP}_{k(n)}^{\text{cc}}$  from  $\text{BPP}_{k(n)}^{\text{cc}}$ , we obtain  $n^{\Omega(1/k)}/2^{O(k)}$  lower bounds on the  $k$ -party NOF communication complexity of set disjointness which is non-trivial for up to  $\Theta(\sqrt{\log n})$  players. The best previous lower bounds of Lee and Shraibman [LS08] and Chattopadhyay and Ada [CA08] for set disjointness describe above do not apply for  $\omega(\log \log n)$  players.

The high-level idea of the  $k$ -party version of the pattern matrix method as described in [CA08] is as follows. Suppose that we want to prove  $k$ -party lower bounds for a function  $\mathcal{F}$ . The general idea is to show that  $\mathcal{F}$  can express some  $\mathcal{F}_k^f$  (specified below) which is a function that under many projection patterns is the same as a function  $f$  of large approximate degree. If  $f$  has large approximate degree, then Sherstov showed that there exists another function  $g$  and a distribution  $\mu$  on inputs such that with respect to  $\mu$ ,  $g$  is both highly correlated with  $f$  and orthogonal to all low-degree polynomials. It follows that  $\mathcal{F}_k^f$  is also highly correlated with  $\mathcal{F}_k^g$  and, using the generalized discrepancy method for communication complexity lower bounds it suffices to prove a discrepancy lower bound for the latter function. Thanks to the orthogonality of  $g$  to all low degree polynomials this is possible using an iterated application of the Cauchy-Schwartz inequality as in Babai, Nisan, and Szegedy [BNS92]. For example, the bound for set disjointness  $\text{DISJ}_{k,n}(x) = \bigvee_{i=1}^n \bigwedge_{j=1}^k x_{ij}$ , which more properly should be called set intersection, corresponds to the case that  $f = \text{OR}$  which has approximate degree  $\Omega(\sqrt{n})$ .

In the two party case, Razborov and Sherstov [RS08] extended Sherstov's method to yield sign-rank lower bounds for the  $\text{AC}_3^0$  function  $\mathcal{F}_2^{MP}$  where  $MP(x) = \bigwedge_{i=1}^m \bigvee_{j=1}^{4m^2} x_{ij}$  is the so-called Minsky-Papert function which has approximate degree  $\Omega(m)$ . The key to their argument is to show that there is an orthogonalizing distribution  $\mu$  for  $MP$  that is "min-smooth" in that it

assigns probability at least  $8^{-m}2^{-n-1}$  to any input vector on which  $MP$  is true.

We prove our results by showing that any function  $f$  for which there is a diverse collection of partial assignments  $\rho$  such that each of the subfunctions  $f|_\rho$  of  $f$  requires large approximate degree, there is an orthogonalizing distribution  $\mu$  for  $f$  that is “max-smooth” in that the probability of subsets defined by partial assignments cannot be too much larger than under the uniform distribution. The diversity of the partial assignments is determined by a parameter  $\alpha$  so we call the degree bound the  $(\epsilon, \alpha)$ -approximate degree. This property is somewhat delicate but applies directly to PARITY and, much more importantly, to  $\text{TRIBES}_{p,q}(x) = \bigvee_{i=1}^q \bigwedge_{j=1}^p x_{ij}$  for certain choices of  $p$  and  $q$ . Since  $\text{TRIBES}_{p,q}$  is a subfunction of many other functions we can use it to obtain lower bounds for many functions in  $\text{AC}^0$ . (The property unfortunately does not apply to OR but we are able to derive our lower bounds for  $\text{DISJ}_{k,n}$  via reduction.) Our lower bound method also shows that the simple masking version of the pattern matrix method is sufficient to obtain strong lower bounds.

**Results** Let  $T$  be the set of all Boolean functions that map the all 0’s input to false and each input with precisely one 1 to true. For any integers  $m, s, k > 0$ , any Boolean function  $f$  on  $m$  bits, and any  $s$ -bit function  $t \in T$ , we define the following function on  $msk$  bits:

$$\mathcal{H}_k^{f,t}(x_1, \dots, x_k) := f(t(\bigwedge_{i=1}^k x_{11i}, \dots, \bigwedge_{i=1}^k x_{1si}), \dots, t(\bigwedge_{i=1}^k x_{m1i}, \dots, \bigwedge_{i=1}^k x_{msi})),$$

for any  $x_1, \dots, x_k \in \{0, 1\}^{ms}$ . Let  $n = ms$ . We associate each such  $\mathcal{H}_k^{f,t}$  with the  $k$ -party NOF communication problem in which player  $i$  can see all  $x_j$  except for  $x_i$  and they want to compute  $\mathcal{H}_k^{f,t}$ .

For instance, setting  $f$  and  $t$  to be OR makes  $\mathcal{H}_k^{f,t}$  the set disjointness function  $\text{DISJ}_{k,n}$  and setting both  $f$  and  $t$  to be PARITY makes  $\mathcal{H}_k^{f,t}$  the Generalized Inner Product (GIP) function.

Given  $f$  as above and  $n = ms$ , we also define a function on  $nk$  bits by:

$$\mathcal{F}_k^f(x, y_1, \dots, y_{k-1}) := f(x | \phi_{\text{AND}}(\bigwedge_{j=1}^{k-1} y_j)),$$

where  $\phi_{\text{AND}}(z)$  returns the set of non-zero indices in  $z$  and  $x|S$  is the bit vector obtained by restricting  $x$  to indices in  $S$ . We also associate with each  $\mathcal{F}_k^f$  the  $k$ -party NOF communication problem on  $x, y_1, \dots, y_{k-1} \in \{0, 1\}^n$  in which the player 0 holds  $x$  and for  $1 \leq i \leq k-1$ , player  $i$  holds  $y_i$ , and they want to compute  $\mathcal{F}_k^f$ .

If we partition the above  $n$ -bit input string  $x$  into  $m$  blocks of size  $s$  and we restrict the inputs  $y_1, \dots, y_{k-1}$  such that the set  $S = \phi_{\text{AND}}(\bigwedge_{j=1}^{k-1} y_j)$  as above selects exactly one bit in each of the  $m$  blocks, then it is easy to see that in this case  $\mathcal{F}_k^f$  is a subfunction of  $\mathcal{H}_k^{f,t}$ . From now on, unless stated otherwise, we will assume that the inputs always satisfy this restriction.

We show that  $(\epsilon, \alpha)$ -approximate degree lower bounds for a function  $f$  allows one to derive lower bounds for  $\mathcal{F}_k^f$ .

**Theorem 1.1.** *Let  $n = ms$  where  $s = \lceil 4e \rceil^{k-1}$ . For  $0 \leq \alpha < 1$  and any Boolean function  $f$  on  $m$  bits with  $(5/6, \alpha)$ -approximate degree  $d$ , the function  $\mathcal{F}_k^f$  defined on  $n \cdot k$  bits requires  $R_{1/3}^k(\mathcal{F}_k^f)$  that is  $\Omega(d/2^k)$  for  $k \leq (1 - \alpha) \log_2 d$ .*

**Corollary 1.2.** *Under the same conditions as above, if  $t$  is any  $s$ -bit function in  $T$  then  $\mathcal{H}_k^{f,t}$  has  $k$ -party randomized NOF communication complexity  $\Omega(d/2^k)$ .*

By analyzing the approximation properties of the  $m = ps$  bit function  $\text{TRIBES}_{p,q}$  for suitable choices of  $p$  and  $q$ , we obtain the first  $\text{AC}^0$  function separating  $\text{NP}_k^{\text{cc}}$  from  $\text{BPP}_k^{\text{cc}}$  for  $k = \omega(\log \log n)$ . The separation is non-trivial for  $k$  up to  $\Theta(\log n)$ .

**Theorem 1.3.** *There is a constant  $a > 0$  such for any  $n$  and integer  $k = k(n) \leq a \log_2 n$  such that  $n \geq ms$  for  $s = \lceil 4e \rceil^{k-1}$  and  $m$  is of the form  $pq$  such that  $\lceil q^{0.2} \rceil < 2^p \leq \frac{1}{6} q^{0.3} \ln 2$  and the following holds. The randomized  $k$ -party NOF communication complexity of  $\mathcal{H}_k^{\text{TRIBES}_{p,q}, \text{OR}_s}$  is  $\Omega(n^{0.3-\epsilon}/c^k)$  for  $c = 2(4e)^{0.3}$  and its nondeterministic  $k$ -party NOF communication complexity is  $O(\log^2 n)$ .*

Since the function  $\mathcal{H}_k^{\text{TRIBES}_{p,q}, \text{OR}_s}$  is given by a read-once depth 4 formula we have the following theorem.

**Corollary 1.4.** *There is a constant  $\delta > 0$  such that for any  $k(n) \leq \delta \log_2 n$  there is a function in  $\text{AC}_4^0 \cap (\text{NP}_k^{\text{cc}} - \text{BPP}_k^{\text{cc}})$ .*

By a reduction from  $\mathcal{H}_k^{\text{TRIBES}_{p,q}, \text{OR}_s}$  to  $\text{DISJ}_{k,n}$  we obtain the following lower bound.

**Theorem 1.5.** *The randomized  $k$ -party NOF communication complexity of  $\text{DISJ}_{k,n}$  is  $n^{\Omega(1/k)}/2^{O(k)}$ .*

Write  $\text{TRIBES}'_{p,q}$  for the dual function to  $\text{TRIBES}_{p,q}$ ,  $\text{TRIBES}'_{p,q}(x) = \bigwedge_{i=1}^q \bigvee_{j=1}^p x_{ij}$ . Observe that  $\mathcal{H}_k^{\text{TRIBES}'_{p,q}, \text{OR}_s}$  is a read-once depth 3  $\text{AC}^0$  function since the two layers of  $\vee$  gates can be combined. Since  $\text{TRIBES}'_{p,q}$  has the same degree approximation properties as  $\text{TRIBES}_{p,q}$ , we obtain a similar lower bound for read-once  $\text{AC}_3^0$  functions.

**Theorem 1.6.** *There is a constant  $a > 0$  such for any  $n$  and integer  $k = k(n) \leq a \log_2 n$  such that following holds. There is a function  $\mathcal{F}$  in read-once  $\text{AC}_3^0$ , namely  $\mathcal{H}_k^{\text{TRIBES}'_{p,q}, \text{OR}_s}$  for  $s = \lceil 4e \rceil^{k-1}$  and  $\lceil q^{0.2} \rceil < 2^p \leq \frac{1}{6} q^{0.3} \ln 2$ , whose randomized  $k$ -party NOF communication complexity is  $\Omega(n^{0.3-\epsilon}/c^k)$  for  $c = 2(4e)^{0.3}$ .*

Our technique yields a new sufficient criterion for functions to have high randomized communication complexity. It can be used to provide strong lower bounds for  $k = O(\log n)$  for many other functions including the extension of the Minsky-Papert function  $MP$  considered by Razborov and Sherstov, and the Generalized Inner Product.

Our paper is organized as follows. In Section 2 we give an overview of the method of [She08, CA08] based on orthogonalizing distributions for functions of large  $\epsilon$ -approximate degree and briefly discuss its limitations. In Sections 3 and 4 we define a new notion which we call the  $(\epsilon, \alpha)$ -approximate degree of a function and show how we can use it to prove Theorem 1.1. In Section 5 we prove that the function  $\text{TRIBES}_{p,q}$  has large  $(\epsilon, \alpha)$ -approximate degree. Then we prove Theorems 1.3, 1.5, and 1.6 in Section 6 and conclude by considering a variety of other functions.

## 2 Preliminaries

### 2.1 Notations and Terminology

We follow the notation used in [DPV]. We will assume that a Boolean function on  $m$  bits is a map  $f : \{0, 1\}^m \rightarrow \{-1, 1\}$ .

**Correlation** Let  $f, g : \{0, 1\}^m \mapsto \mathbb{R}$  be two functions, and let  $\mu$  be a distribution on  $\{0, 1\}^m$ . We define the *correlation* between  $f$  and  $g$  under  $\mu$  to be  $\text{Cor}_\mu(f, g) := \mathbf{E}_{x \sim \mu}[f(x)g(x)]$ . If  $\mathcal{G}$  is a class of functions  $g : \{0, 1\}^m \mapsto \mathbb{R}$ , we define the correlation between  $f$  and  $\mathcal{G}$  under  $\mu$  to be  $\text{Cor}_\mu(f, \mathcal{G}) := \max_{g \in \mathcal{G}} \text{Cor}_\mu(f, g)$ .

**Communication complexity** We denote by  $R_\epsilon^k(f)$  the cost of the best  $k$ -party randomized NOF communication protocol for  $f$  with two-sided error at most  $\epsilon$ , and  $N^k(f)$  the cost of the best  $k$ -party nondeterministic communication protocol for  $f$ . We denote by  $\Pi_k^c$  the class of all deterministic  $k$ -party communication protocols of cost at most  $c$ .

**Fact 2.1.** [KN97] *If there exists a distribution  $\mu$  such that  $\text{Cor}_\mu(f, \Pi_k^c) \leq 1/3$  then  $R_{1/3}^k(f) \geq c$ .*

**Lemma 2.2** ([BNS92]). *Let  $f : \{0, 1\}^{m \times k} \mapsto \mathbb{R}$  and  $U_m$  be the uniform distribution on  $\{0, 1\}^m$ . Then,*

$$\text{Cor}_{U_m}(f, \Pi_k^c)^{2^{k-1}} \leq 2^{c \cdot 2^{k-1}} \cdot \mathbf{E}_{y_1^0, \dots, y_{k-1}^0, y_1^1, \dots, y_{k-1}^1 \in \{0, 1\}^m} \left[ \left| \mathbf{E}_{x \in \{0, 1\}^m} \left[ \prod_{u \in \{0, 1\}^{k-1}} f(x, y_1^u, \dots, y_{k-1}^u) \right] \right| \right].$$

**Approximate degree** The  $\epsilon$ -approximate degree of  $f$ ,  $\text{deg}_\epsilon(f)$ , is the smallest  $d$  for which there exists a multivariate real-valued polynomial  $p$  of degree  $d$  such that  $\|f - p\|_\infty = \max_x |f(x) - p(x)| \leq \epsilon$ . Following [NS94] we have the following property of approximate degree of OR.

**Proposition 2.3.** *Let  $\text{OR}_m : \{0, 1\}^m \rightarrow \{1, -1\}$ . For  $0 \leq \epsilon < 1$ ,  $\text{deg}_\epsilon(\text{OR}_m) \geq \sqrt{(1 - \epsilon)m/2}$ .*

Define an inner product  $\langle \cdot, \cdot \rangle$  on the set of functions  $f : \{0, 1\}^m \rightarrow \mathbb{R}$  by  $\langle f, g \rangle = \mathbf{E}[f \cdot g]$ . For  $S \subseteq [m]$ , let  $\chi_S : \{0, 1\}^m \rightarrow \{-1, 1\}$  be the function  $\chi_S = \prod_{i \in S} (-1)^{x_i}$ . The  $\chi_S$  for  $S \subseteq [m]$  form an orthonormal basis of this space.

**Lemma 2.4** ([She08]). *If  $f : \{0, 1\}^m \mapsto \{-1, 1\}$  is a Boolean function with  $\text{deg}_\epsilon(f) \geq d$  then there exists a function  $g : \{0, 1\}^m \mapsto \{-1, 1\}$  and a distribution  $\mu$  on  $\{0, 1\}^m$  such that:*

1.  $\text{Cor}_\mu(g, f) > \epsilon$ ; and
2. for every  $S \subseteq [m]$  with  $|S| < d$  and every function  $h : \{0, 1\}^{|S|} \mapsto \mathbb{R}$ ,  $\mathbf{E}_{x \sim \mu}[g(x) \cdot h(x|S)] = 0$ .

*Proof.* Let  $\Phi_d$  be the space of polynomials of degree less than  $d$ . By definition,  $\text{deg}_\epsilon(f) \geq d$  if and only if  $\min_{q \in \Phi_d} \|f - q\|_\infty > \epsilon$ . By duality of norms we have  $\min_{q \in \Phi_d} \|f - q\|_\infty = \max_{p \in \Phi_d^\perp, \|p\|_1 = 1} \langle f, p \rangle$ . Writing  $\mu(x) = |p(x)|$  the condition  $\|p\|_1 = 1$  implies that  $\mu$  is a probability distribution and letting  $g(x) = p(x)/\mu(x)$  for  $\mu(x) \neq 0$  and  $g(x) = 1$  if  $\mu(x) = 0$ . Then  $p(x) = \mu(x)g(x)$ . Therefore

$$\epsilon < \langle f, p \rangle = \mathbf{E}[f \cdot p] = \mathbf{E}[f \cdot g \cdot \mu] = \mathbf{E}_{x \sim \mu}[f(x)g(x)] = \text{Cor}_\mu(f, g).$$

Moreover since  $p \in \Phi_d^\perp$ , we have  $0 = \langle \chi_S, p \rangle = \mathbf{E}_{x \sim \mu}[\chi_S(x)g(x)]$ . Now for  $h : \{0, 1\}^{|S|} \rightarrow \mathbb{R}$  for  $|S| \leq d$ ,  $h(x|S)$  can be expressed as a degree  $|S|$  polynomial and by linearity  $\mathbf{E}_{x \sim \mu}[g(x) \cdot h(x|S)] = 0$ .  $\square$

We will extend this lemma in Section 3 using more general LP duality.

## 2.2 The correlation method

We give an overview of the method as described in [CA08], which extends ideas of [She07, She08] from 2-party to  $k$ -party communication complexity, with specific details at those points that we are extending in this paper.

Given a Boolean function  $f$  on  $m$  bits, where  $f$  has large  $5/6$ -approximate degree  $d$  (i.e,  $d$  is polynomial in  $m$ ), we want to lower bound  $R_{1/3}^k(\mathcal{F}_k^f)$ , where  $\mathcal{F}_k^f(x, y_1, \dots, y_{k-1})$  is on  $n \cdot k$  bits for  $n = m \cdot s$ .

From Lemma 2.4, we obtain another Boolean function  $g$  and a distribution  $\mu$  such that:

1.  $\text{Cor}_\mu(g, f) \geq 5/6$ ; and
2. for every  $S \subseteq [m]$  with  $|S| < d$  and every function  $h : \{0, 1\}^{|S|} \mapsto \mathbb{R}$ ,  $\mathbf{E}_{x \sim \mu}[g(x) \cdot h(x|S)] = 0$ .

Divide each player's  $n$ -bit input into  $m$  blocks of size  $s$ . Let  $\ell$  be that  $n/m = s = \ell^{k-1}$ . Hence we can imagine that  $x$  consists of  $m$  arrays, each having  $k-1$  dimensions. For  $1 \leq i \leq k-1$ , each of the  $m$  blocks in  $y_i$  is (a bit vector representing) an index in  $[\ell]$ . Therefore we can view each  $y_i$  as in  $[\ell]^m$ . Thus  $\phi_{AND}(y_1, \dots, y_{k-1})$  selects exactly one bit of  $x$  in each of  $m$  blocks.

Based on  $\mu$ , we define a distribution  $\lambda$  on  $n \cdot k$  bits in a straightforward way as follows:

$$\lambda(x, y_1, \dots, y_{k-1}) := \frac{\mu(x|\phi_{AND}(y_1, \dots, y_{k-1}))}{\ell^{km} 2^{n-m}}$$

for eligible  $y_1, \dots, y_{k-1}$  and 0 otherwise. Here “eligible” means that  $y_1, \dots, y_{k-1}$  satisfy the above requirements. Then it can be verified that  $\text{Cor}_\lambda(\mathcal{F}_k^f, \mathcal{F}_k^g) = \text{Cor}_\mu(f, g) \geq 5/6$ . Consequently,

$$\text{Cor}_\lambda(\mathcal{F}_k^f, \Pi_k^c) \leq \text{Cor}_\lambda(\mathcal{F}_k^g, \Pi_k^c) + 1/6.$$

Therefore we only need to bound  $\text{Cor}_\lambda(\mathcal{F}_k^g, \Pi_k^c)$ . Then by Lemma 2.2,

$$\begin{aligned} \text{Cor}_\lambda(\mathcal{F}_k^g, \Pi_k^c)^{2^{k-1}} &= 2^{m2^{k-1}} \text{Cor}_{U_m}(\mu(x|\phi_{AND}(y_1, \dots, y_{k-1}))g(x|\phi_{AND}(y_1, \dots, y_{k-1})), \Pi_k^c)^{2^{k-1}} \\ &\leq 2^{(c+m) \cdot 2^{k-1}} \cdot \mathbf{E}_{y_1^0, \dots, y_{k-1}^0, y_1^1, \dots, y_{k-1}^1} H(y_1^0, \dots, y_{k-1}^0, y_1^1, \dots, y_{k-1}^1), \end{aligned}$$

where

$$H(y_1^0, \dots, y_{k-1}^0, y_1^1, \dots, y_{k-1}^1) := \left| \mathbf{E}_x \left[ \prod_{u \in \{0,1\}^{k-1}} \mu(x|\phi_{AND}(y_1^{u_1}, \dots, y_{k-1}^{u_{k-1}})) g(x|\phi_{AND}(y_1^{u_1}, \dots, y_{k-1}^{u_{k-1}})) \right] \right|.$$

For  $1 \leq i \leq k-1$ , let  $r_i \in \{0, \dots, m\}$  be the number of blocks for which  $y_i^0$  and  $y_i^1$  give the same index. Let  $r = \sum r_i$ . We rely on the following three propositions to continue the proof. Proposition 2.5 and Proposition 2.7 are the same as in [CA08], so we do not give their proofs. We will prove an extension of Proposition 2.6 in Section 3.

**Proposition 2.5.** *If  $r < d$ , then  $H(y_1^0, \dots, y_{k-1}^0, y_1^1, \dots, y_{k-1}^1) = 0$ .*

**Proposition 2.6.**  $H(y_1^0, \dots, y_{k-1}^0, y_1^1, \dots, y_{k-1}^1) \leq \frac{2^{(2^{k-1}-1)r}}{2^{2^{k-1}m}}$ .

**Proposition 2.7.** *For  $d \leq j \leq (k-1)m$ ,  $\Pr[r = j] \leq \left(\frac{e(k-1)m}{j(\ell-1)}\right)^j (1 - \frac{1}{\ell})^{(k-1)m}$ .*

In [CA08, LS08], to prove the lower bound for  $\text{DISJ}_{k,n}$ , the function  $f$  is set to  $\text{OR}_m$  and  $t$  is set to  $\text{OR}_s$ . By Proposition 2.3,  $d = \text{deg}_{5/6}(\text{OR}_m) \geq \sqrt{m/12}$ . Plugging the bound in Proposition 2.7 together with the bounds from Proposition 2.5 for  $r < d$  and from Proposition 2.6 when  $r \geq d$  into the above correlation inequality it is not hard to show that

$$\text{Cor}_\lambda(\mathcal{F}_k^g, \Pi_k^c) \leq \frac{2^c}{2^{d/2^k}},$$

for  $\ell > \frac{2^{2^k} kem}{d}$ . Hence for  $k = O(\log \log n)$  and  $c$  a small enough polynomial in  $n$ , we have a polynomial lower bound for  $R_{1/3}^k(\text{DISJ}_{k,n}) \geq c$ .

The key limitation of the above technique is the required lower bound on  $\ell$  which follows from the weakness of the upper bound in Proposition 2.6. That weakness is implied by how little can be assumed about the orthogonalizing distribution  $\mu$  given by Lemma 2.4. In particular, the arguments in [She08, CA08, LS08] all allow that  $\mu$  may assign all of its probability mass to small subsets of points defined by partial assignments. Indeed, when the function  $f$  is  $\text{OR}_m$ , this is the case. However, we will show that for other very simple functions  $f$  one can choose the orthogonalizing distribution  $\mu$  so that it does not assign too much weight on such small sets of points; that is,  $\mu$  is “max-smooth”. To guarantee this property of  $\mu$  we need to strengthen Lemma 2.4 by assuming more of  $f$  than just large approximate degree.

### 3 Beyond approximate degree: a new sufficient criterion for strong communication complexity bounds

A  $\rho \in \{0, 1, *\}^m$  is called a *restriction*. For any restriction  $\rho$ , let  $\text{unset}(\rho) \subseteq [m]$  be the set of star positions in  $\rho$ , let  $|\rho| = m - |\text{unset}(\rho)|$ , and let  $C_\rho$  be the set of all  $x \in \{0, 1\}^m$  such that for any  $1 \leq i \leq m$ , either  $\rho_i = *$  or  $\rho_i = x_i$ . Hence  $|C_\rho| = 2^{m-|\rho|}$ . Given a restriction  $\rho \in \{0, 1, *\}^m$  and a function  $f$  on  $\{0, 1\}^m$ , we define  $f|_\rho$  on  $\{0, 1\}^{m-|\rho|}$  in the natural way.

The approximate degree of a function  $f$  says how hard it is to approximate  $f$ . In this paper, we need a stronger notion which requires that many widely distributed restrictions of  $f$  also require large approximate degree.

**Definition** Given  $0 < \epsilon, \alpha \leq 1$  and  $d > 0$ , let  $\Pi = \Pi_{d,\epsilon}(f) \subseteq \{0, 1, *\}^m$  be a set of restrictions such that for any  $\pi \in \Pi$ ,  $\text{deg}_\epsilon(f|_\pi) \geq d$ . We say that  $f$  has  $(\epsilon, \alpha)$ -approximate degree at least  $d$ , denoted as  $\text{deg}_{\epsilon,\alpha}(f) \geq d$ , if restrictions in  $\Pi$  are spread out “evenly”. Formally, there is a distribution  $\nu$  on  $\Pi$  such that for any  $\rho \in \{0, 1, *\}^m$  with  $|\rho| \geq d^\alpha$ , then

$$\Pr_{\pi \sim \nu} [C_\rho \cap C_\pi \neq \emptyset] \leq 2^{|\rho|^\alpha - |\rho|}.$$

The set  $\Pi$  and the distribution  $\nu$  are the *witnesses* for the  $(\epsilon, \alpha)$ -approximate degree of  $f$ . Note that  $\text{deg}_\epsilon(f) = \text{deg}_{\epsilon,1}(f)$ .

We will use this definition to prove the following theorem.

**Theorem 3.1** (restatement of Theorem 1.1). *Let  $n = ms$  where  $s = \lceil 4e \rceil^{k-1}$ . For  $0 \leq \alpha < 1$  and any Boolean function  $f$  on  $m$  bits with  $(5/6, \alpha)$ -approximate degree  $d$ , the function  $\mathcal{F}_k^f$  defined on  $n \cdot k$  bits requires  $R_{1/3}^k(\mathcal{F}_k^f)$  that is  $\Omega(d/2^k)$  for  $k \leq (1 - \alpha) \log_2 d$ .*

To prove the theorem, we first need the following consequence of large  $(\epsilon, \alpha)$ -approximate degree. We postpone its proof to Section 4.

**Lemma 3.2** (extension of Lemma 2.4). *Given  $0 < \epsilon, \alpha \leq 1$ . If  $f : \{0, 1\}^m \mapsto \{-1, 1\}$  is a Boolean function with  $(\epsilon, \alpha)$ -approximate degree  $d$ , there exist a function  $g : \{0, 1\}^m \mapsto \{-1, 1\}$  and a distribution  $\mu$  on  $\{0, 1\}^m$  such that:*

1.  $\text{Cor}_\mu(g, f) \geq \epsilon$ ;
2. for every  $T \subseteq [m]$  with  $|T| < d$  and every function  $h : \{0, 1\}^{|T|} \mapsto \mathbb{R}$ ,  $\mathbf{E}_{x \sim \mu}[g(x) \cdot h(x|T)] = 0$ ; and
3. for any restriction  $\rho$  with  $|\rho| \geq d^\alpha$ ,  $\mu(C_\rho) \leq 2^{|\rho|^\alpha - |\rho|} / \epsilon$ .

Note that, although the upper bound on  $\mu(C_\rho)$  may seem quite weak, it will be sufficient to obtain an exponential improvement in the dependence of communication complexity lower bounds on  $k$ . Moreover, we note in Section 4 that for any function  $f$  computed by an  $\text{AC}^0$  circuit the assumption and the upper bound are essentially the best possible for  $d = \log^{\omega(1)} n$ .

We now use Lemma 3.2 to prove an improvement of Proposition 2.6. This is the key to our improved bounds.

**Lemma 3.3.** *If  $f : \{0, 1\}^m \rightarrow \{1, -1\}$  has  $(\epsilon, \alpha)$ -approximate degree  $d$ , if  $g$  and  $\mu$  are given by the application of Lemma 3.2 to  $f$ , and if  $r \geq d$ , then*

$$H(y_1^0, \dots, y_{k-1}^0, y_1^1, \dots, y_{k-1}^1) \leq \frac{2^{(2^{k-1}-1)r^\alpha}}{2^{2^{k-1}m} \epsilon^{2^{k-1}-1}}.$$

*Proof.* The proof of this lemma is similar to that of [CA08] except that we apply the upper bound from the third condition of Lemma 3.2. Let  $Y_{0^{k-1}}$  represent the set of  $m$  variables indexed jointly by  $y_1^0, \dots, y_{k-1}^0$ . There is precisely one variable chosen from each of the  $m$  blocks. Then in increasing order for each nonzero  $u \in \{0, 1\}^{k-1}$ , we let  $Y_u$  represent the set of variables indexed jointly by  $y_1^{u_1}, \dots, y_{k-1}^{u_{k-1}}$  that are not in  $Y_{0^{k-1}} \cup \bigcup_{u' < u} Y_{u'}$ . By definition we then have for each nonzero  $u$ ,  $|Y_u| \geq m - r$ . Let  $Z = \bigcup_{u \in \{0, 1\}^{k-1}} Y_u$ .

Since  $g$  is 1/-1 valued,

$$\begin{aligned} H(y_1^0, \dots, y_{k-1}^0, y_1^1, \dots, y_{k-1}^1) &= \left| \mathbf{E}_x \left[ \prod_{u \in \{0, 1\}^{k-1}} \mu(x | \phi_{\text{AND}}(y_1^{u_1}, \dots, y_{k-1}^{u_{k-1}})) g(x | \phi_{\text{AND}}(y_1^{u_1}, \dots, y_{k-1}^{u_{k-1}})) \right] \right| \\ &\leq \mathbf{E}_Z \prod_{u \in \{0, 1\}^{k-1}} \mu(x | \phi_{\text{AND}}(y_1^{u_1}, \dots, y_{k-1}^{u_{k-1}})) \\ &= \mathbf{E}_{Y_0^{k-1}} \mu(x | \phi_{\text{AND}}(y_1^0, \dots, y_{k-1}^0)) \\ &\times \max_{Y_{0^{k-1}}} \mathbf{E}_{Y_{0\dots 01}} \mu(x | \phi_{\text{AND}}(y_1^0, \dots, y_{k-1}^1)) \\ &\times \max_{Y_{0\dots 01} \cup Y_{0\dots 10}} \mathbf{E}_{Y_{0\dots 10}} \mu(x | \phi_{\text{AND}}(y_1^0, \dots, y_{k-1}^0)) \\ &\times \dots \end{aligned} \tag{1}$$

and so on repeatedly for all  $2^{k-1}$  of the  $Y_u$ . The term at line (1) equals  $2^{-m}$  because  $\mu$  is a distribution. Now we bound each of the remaining terms. For each non-zero  $u \in \{0, 1\}^{k-1}$ , the corresponding term with  $u$  is

$$T_u = \max_{\bigcup_{u' < u} Y_{u'}} \mathbf{E}_{Y_u} \mu(x | \phi_{\text{AND}}(y_1^{u_1}, \dots, y_{k-1}^{u_{k-1}})).$$



Let  $Y_u = m - i \geq m - r$ . If  $i < r^\alpha$ , then we can upper bound  $T_u$  as

$$T_u \leq \frac{1}{2^{m-i}} < 2^{r^\alpha - m}.$$

Otherwise,  $i \geq r^\alpha \geq d^\alpha$ . Since  $\mu$  is as defined, we can then bound  $T_u$  by

$$T_u \leq \frac{2^{i^\alpha - i} / \epsilon}{2^{m-i}} \leq \frac{2^{r^\alpha - m}}{\epsilon}.$$

Thus in both cases,  $T_u \leq \frac{2^{r^\alpha - m}}{\epsilon}$ . Hence the lemma follows.  $\square$

Now we are ready to prove the main theorem of this section.

*Proof of Theorem 3.1.* Apply Lemma 3.2 with  $\epsilon = 5/6$  to obtain  $g$  and  $\mu$ . Then follow the approach as outlined in Section 2. What remains is to show that  $\text{Cor}_\lambda(\mathcal{F}_k^g, \Pi_k^c) \leq 1/6$ . Now we have, by Proposition 2.5, Lemma 3.3, and Proposition 2.7,

$$\begin{aligned} \text{Cor}_\lambda(\mathcal{F}_k^g, \Pi_k^c)^{2^{k-1}} &\leq 2^{(c+m) \cdot 2^{k-1}} \cdot \mathbf{E}_{y_1^0, \dots, y_{k-1}^0, y_1^1, \dots, y_{k-1}^1} H(y_1^0, \dots, y_{k-1}^0, y_1^1, \dots, y_{k-1}^1) \\ &\leq 2^{c2^{k-1}} \sum_{j=d}^{(k-1)m} 2^{(2^{k-1}-1)j^\alpha} \left(\frac{6}{5}\right)^{2^{k-1}-1} \left(\frac{e(k-1)m}{j(\ell-1)}\right)^j \left(1 - \frac{1}{\ell}\right)^{(k-1)m}. \end{aligned} \quad (2)$$

Since  $k \leq (1 - \alpha) \log_2 d$ , we have  $(2^{k-1} - 1)j^\alpha < d^{1-\alpha}j^\alpha \leq j$  for  $j \geq d$  so (2) is

$$\begin{aligned} &\leq \left(\frac{6}{5}2^c\right)^{2^{k-1}} \sum_{j=d}^{(k-1)m} \left(\frac{2e(k-1)m}{j(\ell-1)}\right)^j \left(1 - \frac{1}{\ell}\right)^{(k-1)m} \\ &\leq \frac{\left(\frac{6}{5}2^c\right)^{2^{k-1}}}{2^d}, \end{aligned} \quad (3)$$

for  $\ell = \lceil 4e \rceil$ . Hence

$$\text{Cor}_\lambda(\mathcal{F}_k^g, \Pi_k^c) \leq \frac{\frac{6}{5}2^c}{2^{d/2^{k-1}}} \leq 1/6,$$

as long as  $c \leq \log_2\left(\frac{5}{36}2^{d/2^{k-1}}\right)$ . Hence  $R_{1/3}^k$  is  $\Omega(d/2^k)$  for  $k \leq (1 - \alpha) \log_2 d$ .  $\square$

## 4 Proof of Lemma 3.2

*Proof.* As in the proof for Lemma 2.4, we write the requirements down as a linear program and study its dual. The lemma is implied by proving that the following linear program  $\mathcal{P}$  has optimal value 1:

Minimize  $\eta$  subject to

$$y_S : \quad \sum_{x \in \{0,1\}^m} h(x) \chi_S(x) = 0 \quad |S| < d \quad (4)$$

$$\beta : \quad \sum_{x \in \{0,1\}^m} h(x) f(x) \geq \epsilon \quad (5)$$

$$v_x : \quad \mu(x) - h(x) \geq 0 \quad x \in \{0,1\}^m \quad (6)$$

$$w_x : \quad \mu(x) + h(x) \geq 0 \quad x \in \{0,1\}^m \quad (7)$$

$$a_\rho : \quad \eta - 2^{|\rho| - |\rho|^\alpha} \sum_{x \in C_\rho} \mu(x) \geq 0 \quad \rho \in \{0,1,*\}^m, |\rho| \geq d^\alpha \quad (8)$$

$$\gamma : \quad \sum_{x \in \{0,1\}^m} \mu(x) = 1 \quad (9)$$

Suppose that we have optimum  $\eta = 1$ . In this LP formulation, inequality  $\gamma$  ensures that the function  $\mu$  is a probability distribution, and inequalities  $v_x$  and  $w_x$  ensure that  $\mu(x) \geq |h(x)|$  so  $\|h\|_1 \leq 1$ . If  $\|h\|_1 = 1$ , then we must have  $\mu(x) = |h(x)|$  and we can write  $h(x) = \mu(x)g(x)$  as in the proof of Lemma 2.4 and then the inequalities  $y_S$  will ensure that  $\text{Cor}_\mu(g, \chi_S) = 0$  for  $|S| < d$  and inequality  $\beta$  will ensure that  $\text{Cor}_\mu(f, g) \geq \epsilon$  as required. Finally, each inequality  $a_\rho$  ensures that  $\mu(C_\rho) \leq 2^{-|\rho| + |\rho|^\alpha} = 2^{-|\rho| + |\rho|^\alpha}$  which is actually a little stronger than our claim.

The only issue is that an optimal solution might have  $\|h\|_1 < 1$ . However in this case inequality  $\beta$  ensures that  $\|h\|_1 \geq \epsilon$ . Therefore, for any solution of the above LP with function  $h$ , we can define another function  $h'(x) = h(x)/\|h\|_1$  with  $\|h'\|_1 = 1$  and a new probability distribution  $\mu'$  by  $\mu'(x) = |h'(x)| \leq \mu(x)/\|h\|_1 \leq \mu(x)/\epsilon$ . This new  $h'$  and  $\mu'$  still satisfy all the inequalities as before except possibly inequality  $a_\rho$  but in this case if we increase  $\eta$  by a  $1/\|h\|_1$  factor it will also be satisfied. Therefore, the  $\mu'(C_\rho) \leq 2^{-|\rho| + |\rho|^\alpha}/\epsilon$ .

Here is the dual LP:

Maximize  $\beta \cdot \epsilon + \gamma$  subject to

$$\eta : \quad \sum_{\rho \in \{0,1,*\}^m, |\rho| \geq d^\alpha} a_\rho = 1 \quad (10)$$

$$\mu(x) : \quad v_x + w_x + \gamma - \sum_{C_\rho \ni x, |\rho| \geq d^\alpha} 2^{|\rho| - |\rho|^\alpha} a_\rho = 0 \quad x \in \{0,1\}^m \quad (11)$$

$$g(x) : \quad \beta f(x) + \sum_{|S| < d} y_S \chi_S(x) + w_x - v_x = 0 \quad x \in \{0,1\}^m \quad (12)$$

$$\beta, v_x, w_x, a_\rho \geq 0 \quad x \in \{0,1\}^m \quad (13)$$

Since  $y_S$  are arbitrary we can replace  $\sum_{|S| < d} y_S \chi_S(x)$  by  $p_d(x)$  where  $p_d$  is an arbitrary polynomial of degree  $< d$  to obtain the modified dual:

Maximize  $\beta \cdot \epsilon + \gamma$  subject to

$$\eta : \quad \sum_{\rho \in \{0,1,*\}^m, |\rho| \geq d^\alpha} a_\rho = 1 \quad (14)$$

$$\mu(x) : \quad v_x + w_x + \gamma - \sum_{C_\rho \ni x, |\rho| \geq d^\alpha} 2^{|\rho| - |\rho|^\alpha} a_\rho = 0 \quad x \in \{0,1\}^m \quad (15)$$

$$g(x) : \quad \beta f(x) + p_d(x) + w_x - v_x = 0 \quad x \in \{0,1\}^m \quad (16)$$

$$\beta, v_x, w_x, a_\rho \geq 0 \quad x \in \{0,1\}^m \quad (17)$$

Equations (15) and (16) for  $x \in \{0,1\}^m$  together are equivalent to:

$$2w_x + \beta f(x) + p_d(x) + \gamma - \sum_{C_\rho \ni x, |\rho| \geq d^\alpha} 2^{|\rho| - |\rho|^\alpha} a_\rho = 0$$

and

$$2v_x - \beta f(x) - p_d(x) + \gamma - \sum_{C_\rho \ni x, |\rho| \geq d^\alpha} 2^{|\rho| - |\rho|^\alpha} a_\rho = 0.$$

Since these are the only constraints on  $v_x$  and  $w_x$  respectively other than negativity these can be satisfied by any solution to

$$\beta f(x) + p_d(x) + \gamma \leq \sum_{C_\rho \ni x, |\rho| \geq d^\alpha} 2^{|\rho| - |\rho|^\alpha} a_\rho$$

and

$$-\beta f(x) - p_d(x) + \gamma \leq \sum_{C_\rho \ni x, |\rho| \geq d^\alpha} 2^{|\rho| - |\rho|^\alpha} a_\rho,$$

which together are equivalent to

$$|\beta f(x) + p_d(x)| + \gamma \leq \sum_{C_\rho \ni x, |\rho| \geq d^\alpha} 2^{|\rho| - |\rho|^\alpha} a_\rho.$$

Since  $p_d(x)$  is an arbitrary polynomial function of degree less than  $d$  we can write  $p_d = -\beta p'_d$  where  $p'_d$  is another arbitrary polynomial function of degree less than  $d$  and we can replace the terms  $|\beta f(x) + p_d(x)|$  by  $\beta |f(x) - p'_d(x)|$ .

Therefore the dual program  $\mathcal{D}$  is equivalent to maximizing  $\beta \cdot \epsilon + \gamma$  subject to

$$\beta |f(x) - p'_d(x)| + \gamma \leq \sum_{C_\rho \ni x, |\rho| \geq d^\alpha} 2^{|\rho| - |\rho|^\alpha} a_\rho$$

for all  $x \in \{0,1\}^m$ ,  $a_\rho$  is probability distribution on the set of all restrictions of size at least  $d^\alpha$ , and  $p'_d$  is a real-valued function of degree  $< d$ .

Now, let  $B$  be the set of points at which  $|f(x) - p'_d(x)| \geq \epsilon$ . For any  $x \in B$ , the value of the objective function of  $\mathcal{D}$ , which is  $\beta \cdot \epsilon + \gamma$ , is not more than

$$\beta |f(x) - p'_d(x)| + \gamma \leq \sum_{C_\rho \ni x, |\rho| \geq d^\alpha} 2^{|\rho| - |\rho|^\alpha} a_\rho. \quad (18)$$

Let  $R(x)$  denote the right-hand side of inequality (18). It suffices to prove that  $R(x) \leq 1$  for some  $x \in B$ . This is, in turn, equivalent to proving that

$$\min_{x \in B} R(x) \leq 1,$$

for any distribution  $a_\rho$ . Suppose, by contradiction, that there exists a distribution  $a_\rho$  such that  $R(x) > 1$  for any  $x \in B$ . Let  $\Pi$ , the set of restrictions, and  $\nu$ , a distribution on  $\Pi$ , be the witnesses for the  $(\epsilon, \alpha)$ -approximate degree of  $f$ . Picking  $\pi \in \Pi$  randomly according to  $\nu$ , we define the random variable

$$I_\pi := \sum_{\rho: |\rho| \geq d^\alpha, C_\rho \cap C_\pi \neq \emptyset} 2^{|\rho| - |\rho|^\alpha} a_\rho.$$

Then,

$$\mathbf{E}_{\pi \sim \nu}(I_\pi) = \sum_{\rho: |\rho| \geq d^\alpha} \Pr[C_\rho \cap C_\pi \neq \emptyset] \cdot 2^{|\rho| - |\rho|^\alpha} a_\rho \leq \sum_{\rho: |\rho| \geq d^\alpha} 2^{|\rho|^\alpha - |\rho|} \cdot 2^{|\rho| - |\rho|^\alpha} a_\rho \leq 1.$$

Therefore there exists  $\pi \in \Pi$  for which  $I_\pi \leq 1$ . If there exists  $x \in B$  such that  $x \in C_\pi$ , then since

$$R(x) = \sum_{C_\rho \ni x, |\rho| \geq d^\alpha} 2^{|\rho| - |\rho|^\alpha} a_\rho > 1,$$

we would have  $I_\pi > 1$ . Thus  $C_\pi \cap B = \emptyset$ . So for any  $x \in C_\pi$ , we have  $|f(x) - p'_d(x)| \leq \epsilon$ . But since the degree of  $p'_d$  is less than  $d$  this contradicts the fact that  $\deg_\epsilon(f|_\pi) \geq d$ . Thus the lemma follows.  $\square$

We note that the bounds in Lemma 3.2 are essentially the best possible for  $\text{AC}^0$  functions: By results of Linial, Mansour, and Nisan [LMN89], for any  $\text{AC}^0$  function  $f$  there is a function  $p_d$  of  $\log^{O(1)} n$  degree, such that  $\|f - p_d\|_2^2 \leq 2^{n-n^\delta}$  for some constant  $\delta > 0$ . Let  $B_n$  be the set of  $x$  such that  $|f(x) - p_d(x)| \geq \epsilon$ . Then  $|B_n| \epsilon^2 \leq \sum_{x \in B_n} |f(x) - p_d(x)|^2 \leq \|f - p_d\|_2^2 \leq 2^{n-n^\delta}$  so  $|B_n| \leq 2^{n-n^\delta} / \epsilon^2$ . Also, if we tried to replace the upper bound on  $\mu(C_\rho)$  by some  $c(|\rho|)$  where  $c(n)$  is  $\omega(1/|B_n|)$  then we could choose  $a_x = 1/|B_n|$  for  $x \in B_n$  and  $a_\rho = 0$  for all other  $\rho$  and for these values  $\beta$  would be unbounded.

## 5 TRIBES has large $(\epsilon, \alpha)$ -approximate degree

It is not obvious that any function, let alone a function in  $\text{AC}^0$ , has large  $(\epsilon, \alpha)$ -approximate degree for  $\alpha < 1$ . Recall that the function  $\text{TRIBES}_{p,q}$  on  $m = pq$  bits is defined by

$$\text{TRIBES}_{p,q}(x) = \bigvee_{i=1}^q \bigwedge_{j=1}^p x_{i,j}.$$

Usually the function  $\text{TRIBES}$  is defined so that  $2^p$  is linear or nearly-linear in  $q$ . We will show that, with a different relationship in which  $q \gg 2^p$  but  $p$  is still  $\Theta(\log q)$ , the  $(\epsilon, \alpha)$ -approximate degree of  $\text{TRIBES}_{p,q}$  is large.

**Lemma 5.1.** *Let  $r, q, p$  be positive integers with  $q > r > p \geq 2$  and let  $1 > \alpha > \beta > 0$  be such that  $q^\beta \geq rp$ ,  $2^p - 1 \geq q^{1-\beta}$ ,  $q^\alpha \geq \frac{6}{\ln 2} 2^p r$ , and  $r^{\alpha(\alpha-\beta)} \geq 12(3p/\ln 2)^2$ . Then  $\text{TRIBES}_{p,q}$  has  $(5/6, \alpha)$ -approximate degree at least  $\sqrt{r/12}$ .*

*Proof.* We define a distribution  $\nu$  on restrictions  $R_m^{pr}$  that leave  $pr$  out of the  $m$  variables unset as follows: pick uniformly at random a subset of  $q - r$  of the  $q$  terms of  $\text{TRIBES}_{p,q}$ ; then for each of these terms, assign values to the variables in the term uniformly at random from  $\{\{0, 1\}^p - \mathbf{1}^p\}$ . It is clear that for any  $\pi$  with  $\nu(\pi) > 0$ ,  $\text{OR}_r$  is a subfunction of  $\text{TRIBES}_{p,q}|_\pi$  so  $\text{deg}_{5/6}(\text{TRIBES}_{p,q}|_\pi) \geq \text{deg}_{5/6}(\text{OR}_r) \geq \sqrt{r/12}$ .

Let  $\rho$  be any restriction of size  $i = |\rho| \geq (r/12)^{\alpha/2}$ . By definition, we need to prove that

$$\Pr_{\pi \sim \nu} [C_\rho \cap C_\pi \neq \emptyset] \leq 2^{i\alpha - i}.$$

Now

$$\Pr_{\pi \sim \nu} [C_\rho \cap C_\pi \neq \emptyset] = \frac{1}{\binom{q}{q-r}} \sum_{S \subseteq [q], |S|=q-r} \prod_{j \in S} p_j,$$

where  $p_j$  is the probability that  $\pi$  and  $\rho$  agree on the variables in the  $j$ -th term in  $\text{TRIBES}_{p,q}$ . Write  $i = i_1 + \dots + i_q$ , where  $i_j$  is the number of assignments  $\rho$  makes to variables in the  $j$ -th term of  $\text{TRIBES}_{p,q}$ . Then

$$p_j \leq \frac{2^{p-i_j}}{2^p - 1} = 2^{-i_j} \left(1 + \frac{1}{2^p - 1}\right).$$

Let  $i_S = \sum_{j \in S} i_j$  be the number of assignments  $\rho$  makes to variables in terms in  $S$  and  $k_S = |\{j \in S : i_j > 0\}|$  be the number of terms in  $S$  in which  $\rho$  assigns least one value. Hence,

$$\Pr_{\pi \sim \nu} [C_\rho \cap C_\pi \neq \emptyset] < \frac{1}{\binom{q}{q-r}} \sum_{S \subseteq [q], |S|=q-r} 2^{-i_S} \left(1 + \frac{1}{2^p - 1}\right)^{k_S}. \quad (19)$$

Let  $k = |\{j : i_j > 0\}|$  be the total number of terms in which  $\rho$  assigns at least one value. There are 2 cases: (I)  $k \geq q/2$ , and (II)  $k < q/2$ .

Now consider case (I). Thus  $i \geq q/2$ . In Equation 19, we have  $k_S \leq q$  for every  $S$ . Thus,

$$\Pr_{\pi \sim \nu} [C_\rho \cap C_\pi \neq \emptyset] \leq \frac{1}{\binom{q}{q-r}} \sum_{S \subseteq [m], |S|=q-r} 2^{-i_S} \left(1 + \frac{1}{2^p - 1}\right)^q.$$

It is easy to see that  $i_S \geq i - pr$  for every such  $S$ . Hence we get

$$\frac{1}{\binom{q}{q-r}} \sum_{S \subseteq [q], |S|=q-r} 2^{-i_S} \leq 2^{pr-i} \leq 2^{(2i)^\beta - i},$$

since  $pr \leq q^\beta \leq (2i)^\beta$  in this case. Thus,

$$\Pr_{\pi \sim \nu} [C_\rho \cap C_\pi \neq \emptyset] \leq 2^{(2i)^\beta - i} \left(1 + \frac{1}{2^p - 1}\right)^q \leq 2^{(2i)^\beta - i} e^{q^\beta} \leq 2^{2^\beta(1+1/\ln 2)i^\beta - i},$$

since  $q^{1-\beta} \leq 2^p - 1$  and  $i \geq q/2$ . We upper bound the term  $2^\beta(1+1/\ln 2) i^\beta$  by  $i^\alpha$  as follows: Since  $i \geq (r/12)^{\alpha/2}$ ,

$$i^{\alpha-\beta} \geq (r/12)^{\alpha(\alpha-\beta)/2} \geq (r^{\alpha(\alpha-\beta)}/12)^{1/2} \geq 3p/\ln 2 \quad (20)$$

by our assumption in the statement of the lemma. Since  $p \geq 2$ , we have  $i^{\alpha-\beta} > 6 > 2^\beta(1+1/\ln 2)$  which is all that we need to derive that  $\Pr_{\pi \sim \nu} [C_\rho \cap C_\pi \neq \emptyset] < 2^{i\alpha - i}$  in case I.

Next, we consider case (II). We must have  $k \leq p^{1-\beta}(2^p - 1) i^\beta$ , because otherwise

$$i \geq k > p^{1-\beta}(2^p - 1)i^\beta \geq p^{1-\beta}q^{1-\beta}i^\beta,$$

which implies  $i^{1-\beta} > (pq)^{1-\beta}$  and hence  $i > pq = m$  which is impossible. Therefore

$$\left(1 + \frac{1}{2^p - 1}\right)^{k_S} \leq e^{\frac{k_S}{2^p - 1}} \leq e^{\frac{k}{2^p - 1}} \leq e^{p^{1-\beta}i^\beta}.$$

So,

$$\Pr_{\pi \sim \nu}[C_\rho \cap C_\pi \neq \emptyset] < e^{p^{1-\beta}i^\beta} \mathcal{S} \quad \text{where} \quad \mathcal{S} = \frac{1}{\binom{q}{q-r}} \sum_{S \subset [q], |S|=q-r} 2^{-i_S} = E_{S \sim U}[2^{-i_S}].$$

and  $U$  is the uniform distribution on subsets of  $[q]$  of size  $q - r$ .

Now we continue by upper bounding  $\mathcal{S}$ . For the moment let us assume that  $i$  is divisible by  $p$ . If we view the terms as the bins, and the assigned positions by  $\rho$  as balls placed in corresponding bins, then we observe that  $\mathcal{S}$  can only increase if we move one ball from a bin  $A$  of  $x > 0$  balls to another bin  $B$  of  $y \geq x$  balls. This is because only those  $i_S$  with  $S$  containing exactly one of these two bins are affected by this move. Then, we can write the contribution of these  $S$ 's in  $\mathcal{S}$  before the move as

$$\mathcal{S}' = \sum_{S \subset [q], |S|=q-r, S \cap \{A,B\}=1} 2^{-i_S} = \sum_{S' \subset [q] - \{A,B\}, |S'|=q-r-1} 2^{-i_{S'}} (2^{-x} + 2^{-y}),$$

and after the move as

$$\mathcal{S}'' = \sum_{S' \subset [q] - \{A,B\}, |S'|=q-r-1} 2^{-i_{S'}} (2^{-x+1} + 2^{-y-1}).$$

Since  $y \geq x$ ,  $\mathcal{S}'' > \mathcal{S}'$ .

Hence w.l.o.g. and with the assumption that  $p$  divides  $i$ , we can assume that the balls are distributed such that every bin is either full, i.e containing  $p$  balls, or empty. Hence  $k = i/p$  and for any  $1 \leq j \leq q$ , either  $i_j = 0$  or  $i_j = p$ .

**Claim 5.2.** *If  $i$  is divisible by  $p$  then  $\mathcal{S} \leq 2^{-i} e^{2^{p+1}rk/q}$ .*

We first see how the claim suffices to prove the lemma. If  $i$  is not divisible by  $p$  then we note that  $\mathcal{S}$  is a decreasing function of  $i$  and apply the claim for the first  $i' = p \lfloor i/p \rfloor > i - p$  positions set by  $\rho$  to obtain an upper bound of  $\mathcal{S} < 2^{p-i} e^{2^{p+1}ri/(pq)}$  that applies for all choices of  $i$ . The overall bound we obtain in this case is then

$$\begin{aligned} \Pr_{\pi \sim \nu}[C_\rho \cap C_\pi \neq \emptyset] &< e^{p^{1-\beta}i^\beta} 2^p e^{2^{p+1}ri/(pq)} 2^{-i} \\ &= 2^{i^\beta p^{1-\beta} / \ln 2 + p + 2^{p+1}ri/(pq \ln 2)} 2^{-i}. \end{aligned}$$

We now consider the exponent  $i^\beta p^{1-\beta} / \ln 2 + p + 2^{p+1}ri/(pq \ln 2)$  and show that it is at most  $i^\alpha$ . For the first term observe that by (20),  $i^{\alpha-\beta} \geq 3p / \ln 2$  so  $i^\beta p^{1-\beta} / \ln 2 \leq i^\alpha / 3$ . For the second term again by (20) we have  $p \leq i^{\alpha-\beta} / 3 \leq i^\alpha / 3$ . For the last term, since  $q^\alpha \geq \frac{6}{\ln 2} 2^p r$ , we have

$$\frac{2^{p+1}ri}{pq \ln 2} \leq \frac{q^\alpha i}{3pq} \leq i(pq)^{\alpha-1} / 3 \leq i^\alpha / 3,$$

since  $i \leq pq$ . Therefore in case II we have  $\Pr_{\pi \sim \nu}[C_\rho \cap C_\pi \neq \emptyset] < 2^{i^\alpha - i}$  as required. It only remains to prove the claim.

**Proof of Claim:** Let  $T = \{i_j \mid i_j = p\}$  be the subset of  $k$  terms assigned by  $\rho$ . Therefore  $i_S = |S \cap T|p$  where  $S$  is a random set of size  $q - r$  and  $T$  is a fixed set of size  $k$  and both are in  $[q]$ . We have two subcases: (IIa) when  $k \leq r$  and (IIb) when  $q/2 \geq k > r$ .

If  $k \leq r$  then we analyze  $\mathcal{S}$  based on the number  $j$  of elements of  $S$  contained in  $T$ . There are  $\binom{k}{j}$  choices of elements of  $T$  to choose from and  $q - r - j$  elements to select from the  $q - k$  elements of  $\bar{T}$ . Therefore

$$\mathcal{S} = \frac{\sum_{j=0}^k \binom{r}{j} \binom{q-k}{q-r-j} 2^{-jp}}{\binom{q}{q-r}}.$$

Now since

$$\frac{\binom{q-k}{q-r-j}}{\binom{q}{q-r}} = \frac{(q-k)!(q-r)!r!}{q!(q-r-j)!(r-(k-j))!} < \frac{(q-r)^j r^{k-j}}{(q-k)^k} = \left(\frac{r}{q-k}\right)^k \left(\frac{q-r}{r}\right)^j,$$

we can upper bound  $\mathcal{S}$  by

$$\begin{aligned} \left(\frac{r}{q-k}\right)^k \sum_{j=0}^k \binom{k}{j} 2^{-pj} \left(\frac{q-r}{r}\right)^j &= \left(\frac{r}{q-k}\right)^k \left(1 + \frac{q-r}{2^p r}\right)^k \\ &= 2^{-pk} \left(\frac{r}{q-k}\right)^k \left(\frac{2^p r + (q-r)}{r}\right)^k \\ &= 2^{-i} \left(\frac{q + (2^p - 1)r}{q-k}\right)^k \\ &= 2^{-i} \left(1 + \frac{(2^p - 1)r + k}{q-k}\right)^k \\ &\leq 2^{-i} \left(1 + \frac{2^p r}{q-k}\right)^k \\ &\leq 2^{-i} e^{2^p r k / (q-k)} \\ &\leq 2^{-i} e^{2^{p+1} r k / q}. \end{aligned}$$

since  $k \leq q/2$ .

In the case that  $r \leq k \leq q/2$  we observe that by symmetry we can equivalently view the expectation  $\mathcal{S}$  as the result of an experiment in which the set  $S$  of size  $q - r$  is chosen first and the set  $T$  of size  $k$  is chosen uniformly at random. We analyze this case based on the number  $j$  of elements of  $\bar{S}$  contained in  $T$ . There are  $\binom{r}{j}$  choices of elements of  $\bar{S}$  to choose from and  $k - j$  elements to select from the  $q - r \geq q/2 \geq k$  elements of  $S$ . Therefore

$$\mathcal{S} = \frac{\sum_{j=0}^r \binom{r}{j} \binom{q-r}{k-j} 2^{-(k-j)p}}{\binom{q}{k}}.$$

Using the fact that

$$\frac{\binom{q-r}{k-j}}{\binom{q}{k}} = \frac{(q-r)!(q-k)!k!}{q!(k-j)!(q-r-k+j)!} < \frac{(q-k)^{r-j} k^j}{(q-r)^r} = \left(\frac{q-k}{q-r}\right)^r \left(\frac{k}{q-k}\right)^j,$$

we upper bound  $\mathcal{S}$  by

$$\begin{aligned}
2^{-pk} \left(\frac{q-k}{q-r}\right)^r \sum_{j=0}^r \binom{r}{j} \left(\frac{2^p k}{q-k}\right)^j &= 2^{-pk} \left(\frac{q-k}{q-r}\right)^r \left(1 + \frac{2^p k}{q-k}\right)^r \\
&= 2^{-i} \left(\frac{q-k}{q-r}\right)^r \left(\frac{q + (2^p - 1)k}{q-k}\right)^r \\
&= 2^{-i} \left(\frac{q + (2^p - 1)k}{q-r}\right)^r \\
&= 2^{-i} \left(1 + \frac{(2^p - 1)k + r}{q-r}\right)^r \\
&\leq 2^{-i} \left(1 + \frac{2^p k}{q-r}\right)^r \\
&\leq 2^{-i} e^{2^p r k / (q-r)} \\
&\leq 2^{-i} e^{2^{p+1} r k / q}
\end{aligned}$$

since  $r \leq q/2$ . □

**Corollary 5.3.** *Given any  $1 > \epsilon > 0$ . Let  $q, p$  be positive integers with  $q > p \geq 2$  such that  $\lceil q^{1-\beta} \rceil < 2^p \leq \frac{1}{6} q^{\alpha+\epsilon-1} \ln 2$  for some fixed constants  $1 > \alpha > \beta > 1 - \epsilon$ . Then for large enough  $q$ ,  $\text{TRIBES}_{p,q}$  has  $(5/6, \alpha)$ -approximate degree at least  $\sqrt{q^{1-\epsilon}/12}$ .*

*Proof.* We apply Lemma 5.1 with  $r := \lfloor q^{1-\epsilon} \rfloor$ . All conditions in the statement of the lemma would then be satisfied for  $q$  large enough. In particular, for  $q$  large enough,

$$q^\beta / r \geq q^{\beta+\epsilon-1} > \log q > p,$$

and

$$r^{\alpha(\alpha-\beta)} = q^{(1-\epsilon)\alpha(\alpha-\beta)} > 12(3 \log q / \ln 2)^2 > 12(3p / \ln 2)^2.$$

□

**Corollary 5.4.** *Let  $q, p$  be positive integers with  $q > p \geq 2$  such that  $\lceil q^{0.2} \rceil < 2^p \leq \frac{1}{6} q^{0.3} \ln 2$ . Then for large enough  $q$ ,  $\text{TRIBES}_{p,q}$  has  $(5/6, 0.9)$ -approximate degree at least  $\sqrt{q^{0.6}/12}$ .*

*Proof.* Follows from the last corollary with  $\epsilon = 0.4$ ,  $\alpha = 0.9$ , and  $\beta = 0.8$ . □

## 6 Multipart communication complexity of $\text{AC}^0$

### 6.1 A separating function for $\text{NP}_k^{\text{cc}}$ and $\text{BPP}_k^{\text{cc}}$ for $k = O(\log n)$

In this subsection we will show that  $\mathcal{F}_k^{\text{TRIBES}_{p,q}}$  separates  $\text{NP}_k^{\text{cc}}$  and  $\text{BPP}_k^{\text{cc}}$  for  $k = O(\log n)$  for some appropriately chosen values of  $p$  and  $q$ .

**Lemma 6.1.**  *$N^k(\mathcal{F}_k^{\text{TRIBES}_{p,q}})$  is  $O(\log q + p \log n)$  for any  $k \geq 2$ .*



*Proof.* The lemma is easy to see as follows. The 0-th player (who holds  $x$ ), guesses one of the  $q$  branches and sends this guess to all other players. Then he also broadcasts the positions of all the  $p$  bits in that branch. Finally any other player, who can see  $x$  and is given the  $p$  positions, can compute the output of  $\mathcal{F}_k^{\text{TRIBES}_{p,q}}$ . The communication cost is then  $O(\log q + p \log n)$  bits.  $\square$

**Lemma 6.2.** *Let  $0 < \epsilon < 1$ . Let  $q, p$  be sufficiently large positive integers with  $q > p \geq 2$  such that  $\lceil q^{1-\beta} \rceil < 2^p \leq \frac{1}{6}q^{\alpha+\epsilon-1} \ln 2$  for some fixed constants  $1 > \alpha > \beta > 1 - \epsilon$ . Let  $m = pq$  and  $n \geq pqs$  for  $s = \lceil 4e \rceil^{k-1}$  and  $k \leq a \log_2 n$  for some constant  $a > 0$  depending only on  $\alpha$ . Then for any  $\delta > 0$ ,  $R_{1/3}^k(\mathcal{H}_k^{\text{TRIBES}_{p,q}, \text{OR}_s}) \geq R_{1/3}^k(\mathcal{F}_k^{\text{TRIBES}_{p,q}})$  is  $\Omega(q^{(1-\epsilon)/2}/2^k)$ , which is  $\Omega(n^{(1-\epsilon)/2-\delta}/c^k)$  for  $c = 2(4e)^{(1-\epsilon)/2}$ .*

*Proof.* Applying Corollary 5.4, we get that for large enough  $q$ ,  $\text{TRIBES}_{p,q}$  has  $(5/6, \alpha)$ -approximate degree of  $\Omega(q^{(1-\epsilon)/2})$ . Then it follows from Theorem 3.1 that  $R_{1/3}^k(\mathcal{F}_k^{\text{TRIBES}_{p,q}})$  is  $\Omega(q^{(1-\epsilon)/2}/2^k)$ , when  $k$  is  $O(\log n)$ .

Since  $q = m/p$  and  $p$  is  $O(\log q)$ , we get that  $q$  is  $\Omega(m/(\log m))$ . Moreover,  $m \geq n/(4e)^k$ . Hence  $R_{1/3}^k(\mathcal{F}_k^{\text{TRIBES}_{p,q}})$  is  $\Omega(n^{(1-\epsilon)/2-\delta}/c^k)$ , for  $c = 2(4e)^{(1-\epsilon)/2}$  and any  $\delta > 0$ .  $\square$

In particular, we get the following corollary.

**Corollary 6.3.** *Let  $q, p$  be positive integers with  $q > p \geq 2$  such that  $\lceil q^{0.2} \rceil < 2^p \leq \frac{1}{6}q^{0.3} \ln 2$ . Then for  $k \leq a \log_2 n$  for some constant  $a > 0$  the following holds. For  $q$  large enough and any  $\delta > 0$ ,  $R_{1/3}^k(\mathcal{H}_k^{\text{TRIBES}_{p,q}, \text{OR}_s}) \geq R_{1/3}^k(\mathcal{F}_k^{\text{TRIBES}_{p,q}})$  is  $\Omega(q^{0.3}/2^k)$  which is  $\Omega(n^{0.3-\delta}/c^k)$  for  $c = 2(4e)^{0.3}$ .*

*Proof.* Follows from Lemma 6.2 with  $\epsilon = 0.4$ ,  $\alpha = 0.9$ , and  $\beta = 0.8$ .  $\square$

Combining Lemma 6.1 and Corollary 6.3 we obtain our desired separation.

**Theorem 6.4.** *Let  $q, p$  be large enough positive integers with  $q > p \geq 2$  such that  $\lceil q^{0.2} \rceil < 2^p \leq \frac{1}{6}q^{0.3} \ln 2$ . Then  $\mathcal{F}_k^{\text{TRIBES}_{p,q}} \in \text{NP}_k^{\text{cc}} - \text{BPP}_k^{\text{cc}}$  for  $k \leq a \log n$  for some constant  $a > 0$ .*

## 6.2 Lower bound for $\text{DISJ}_{k,n}$

In this subsection we reduce  $\mathcal{H}_k^{\text{TRIBES}_{p,q}, \text{OR}_s}$  to  $\text{DISJ}_{k,n}$  for a suitable value of  $n$  to obtain a NOF communication complexity lower bound on  $\text{DISJ}_{k,n}$  for  $k$  up to  $\Theta(\sqrt{\log n})$  players.

**Theorem 6.5.** *Given constants  $1 > \epsilon, \alpha > 0$  such that  $1 > \alpha > 1 - \epsilon$ . Then for  $k \leq a\sqrt{\log_2 n}$  for some constant  $a > 0$  depending only on  $\alpha, \epsilon$ ,  $R_{1/3}^k(\text{DISJ}_{n,k})$  is  $\Omega(n^{c/k}/2^k)$ , where  $c = \frac{1-\epsilon}{7\alpha+7\epsilon-6}$ . In particular,  $R_{1/3}^k(\text{DISJ}_{n,k})$  is  $\Omega(n^{1/(6k)}/2^k)$ .*

*Proof.* Recall that

$$\text{DISJ}_{k,n}(x) = \bigvee_{i=1}^n \bigwedge_{j=1}^k x_{i,j}.$$

For any  $x \in \{0, 1\}^{Nk}$ , where  $N = pqs$  for integers  $p, q$ , and  $s$  we rewrite  $\mathcal{H}_k^{\text{TRIBES}_{p,q}, \text{OR}_s}$  as

$$\begin{aligned} \mathcal{H}_k^{\text{TRIBES}_{p,q}, \text{OR}_s}(x) &= \bigvee_{i=1}^q \bigwedge_{j=1}^p \bigvee_{u=1}^s \bigwedge_{v=1}^k x_{i,j,u,v} \\ &= \bigvee_{i=1}^q \bigvee_{I \in [s]^p} \bigwedge_{j=1}^p \bigwedge_{v=1}^k x_{i,j,I(j),v} \end{aligned}$$

by expanding the second “ $\wedge$ ”, where  $I(j)$  is the  $j$ -th index of  $I$ . This in turn equals

$$\begin{aligned}
&= \bigvee_{i=1}^q \bigvee_{I \in [s]^p} \bigwedge_{v=1}^k \bigwedge_{j=1}^p x_{i,j,I(j),v} \\
&= \bigvee_{i=1}^q \bigvee_{I \in [s]^p} \bigwedge_{v=1}^k y_{i,I,v} \\
&= \bigvee_{i \in [q], I \in [s]^p} \bigwedge_{v=1}^k y_{i,I,v} \\
&= \text{DISJ}_{n,k}(y),
\end{aligned}$$

where the bits of vector  $y \in \{0, 1\}^{nk}$  for  $n = qs^p$ , indexed by  $i \in [q]$ ,  $I \in [s]^p$ , and  $v \in [k]$ , are given by

$$y_{i,I,v} = \bigwedge_{j=1}^p x_{i,j,I(j),v}.$$

Observe that for any two players  $v \neq v'$ , player  $v'$  can compute any value  $y_{i,I,v}$ . Thus the  $k$  players can compute  $\mathcal{H}_k^{\text{TRIBES}_{p,q}, \text{OR}_s}$  by executing a NOF randomized communication protocol for  $\text{DISJ}_{n,k}$  on  $y$  of length  $nk$ , where  $n = qs^p$ .

Let  $\beta$  be a constant such that  $\alpha > \beta > 1 - \epsilon$ . Let  $q > p \geq 2$  be sufficiently large and satisfy  $\lceil q^{1-\beta} \rceil < 2^p \leq \frac{1}{6}q^{\alpha+\epsilon-1} \ln 2$ . Let  $s = \lceil 4e \rceil^{k-1}$ . From Lemma 6.2, we know that for  $k \leq a \log_2 q$  for some constant  $a > 0$  depending only on  $\alpha$ ,  $R_{1/3}^k(\mathcal{H}_k^{\text{TRIBES}_{p,q}, \text{OR}_s})$  is  $\Omega(q^{(1-\epsilon)/2}/2^k)$ . Observe that since  $2^p \leq \frac{1}{6}q^{\alpha+\epsilon-1} \ln 2 \leq q^{\alpha+\epsilon-1}$  we have

$$n = qs^p < q \lceil 4e \rceil^{kp} < q^{1+(\alpha+\epsilon-1)k \log_2 \lceil 4e \rceil}.$$

Since  $\log_2 \lceil 4e \rceil < 3.5$ , letting  $b = 7(\alpha + \epsilon - 1) + 1$ , we get  $q > n^{2/(bk)}$ . Therefore  $R_{1/3}^k(\text{DISJ}_{n,k})$  is

$$\Omega(n^{\frac{1-\epsilon}{bk}}/2^k).$$

The rest of the theorem follows by taking  $\epsilon = 0.4$  and  $\alpha = 0.9$ . □

### 6.3 The randomized communication complexity of depth-3 AC<sup>0</sup>

In the last two subsections, we showed that there is a depth-4 read-once AC<sup>0</sup> function separating  $\text{NP}_k^{\text{cc}}$  and  $\text{BPP}_k^{\text{cc}}$  for  $k$  up to  $\Theta(\log n)$ , and there is a depth-2 read-once AC<sup>0</sup> function separating  $\text{NP}_k^{\text{cc}}$  and  $\text{BPP}_k^{\text{cc}}$  for  $k$  up to  $\Theta(\sqrt{\log n})$ . In this subsection we show that there is a depth-3 AC<sup>0</sup> function that is hard for randomized NOF communication complexity for  $k$  up to  $\Theta(\log n)$ .

**Corollary 6.6.** *Let  $q, p$  be positive integers with  $q > p \geq 2$  such that  $\lceil q^{0.2} \rceil < 2^p \leq \frac{1}{6}q^{0.3} \ln 2$ , then for  $q$  large enough and any  $\epsilon > 0$ ,  $R_{1/3}^k(\mathcal{H}_k^{f,t}) \geq R_{1/3}^k(\mathcal{F}_k^f)$  is  $\Omega(q^{0.3}/2^k)$  which is  $\Omega(n^{0.3-\epsilon/c^k})$  when  $k$  is  $O(\log n)$ , and  $c = 2(4e)^{0.3}$ , where  $f$  is  $\text{TRIBES}'_{p,q}$ , the dual of the  $\text{TRIBES}_{p,q}$  function on  $m = pq$  bits, and  $t$  is the OR function on  $s$  bits and  $n = m \cdot s$  for  $s = \lceil 4e \rceil^{k-1}$ . Moreover,*

$$\mathcal{H}_k^{f,t}(x) = \bigwedge_{i \in [q]} \bigvee_{j \in [p], u \in [s]} \bigwedge_{v \in [k]} x_{i,j,u,v}$$

is a depth 3 read-once formula.

*Proof.* The first part follows directly from the proof of Lemma 6.2. Since the second layer of  $f$  can be combined with  $t$  into one layer, the second part follows. □

Although we have shown non-trivial lower bounds for  $\text{DISJ}_{k,n}$  for  $k$  up to  $\Theta(\sqrt{\log n})$  it is open whether one can prove similar lower bounds to Corollary 6.6 for  $k = \omega(\sqrt{\log n})$  players for  $\text{DISJ}_{k,n}$

or any other depth-2  $AC^0$  function. The difficulty of extending our lower bound methods is our inability to apply Lemma 3.2 to OR since the constant function 1 approximates OR on all but one point.

The above function is far from unique. In particular, since it contains the above function as a subfunction, a similar  $m^{\Omega(1)}/2^{O(k)}$  lower bound applies to the  $k$ -party NOF communication complexity of the extension of the Minsky-Papert function that is the  $k$ -party analogue of the function considered by Razborov and Sherstov [RS08], namely  $\bigwedge_{i=1}^m \bigvee_{j=1}^{m^2} \bigwedge_{v=1}^k x_{i,j,v}$ .

## 7 Outside of $AC^0$

### 7.1 Generalized Inner Product (GIP)

We can use Theorem 3.1 to prove a  $\Omega(n^{1-\epsilon}/c^k)$ -lower bound on  $k$ -party NOF randomized communication complexity for the GIP function defined on  $n \cdot k$  bits as follows:

$$\text{GIP}_{n,k}(x_1, \dots, x_k) := \bigoplus_{i=1}^n \left( \bigwedge_{j=1}^k x_{j,i} \right),$$

for any constant  $\epsilon > 0$ , where  $c > 4$  is some constant. With the lower bound of  $\Omega(n/4^k)$  proved in [BNS92], our technique shows that we do not lose too much by applying the pattern matrix method with a simple masking scheme.

**Lemma 7.1.** *For any  $0 < \gamma < \alpha \leq 1$ ,  $\text{PARITY}_m$  has  $(5/6, \alpha)$ -approximation degree of  $\Omega(m^\gamma)$ .*

*Proof.* We define the witnesses for the  $(5/6, \alpha)$ -approximate degree of  $\text{PARITY}_m$  to be the set  $R_m^{m^\gamma}$  and  $\nu$  taken to be the uniform distribution on the set. It follows that for any  $\pi \sim \nu$ ,  $\text{deg}_{5/6}(\text{PARITY}_m |_\pi) = \Omega(m^\gamma)$ . Thus what remains is to show that for any integers  $m \geq i \geq \Omega(m^{\gamma\alpha})$ , if we pick uniformly at random  $\rho$  such that  $|\rho| = i$  and  $\pi \sim \nu$ , then

$$\Pr[C_\rho \cap C_\pi \neq \emptyset] < 2^{i^\alpha - i},$$

for sufficiently large  $m$ .

We consider two cases: (1)  $i \leq m^\gamma$  and (2)  $i > m^\gamma$ . These are exactly like the cases in the proof of the claim in Lemma 5.1 with  $q = m$ ,  $r = m^\gamma$ ,  $p = 1$  and  $k = i$ . The bound follows since  $2 \cdot 2^{-i} e^{8m^{\gamma-1}i} < 2^{i^\alpha - i}$ .  $\square$

**Corollary 7.2.**  *$R_{1/3}^k(\text{GIP}_{n,k})$  is  $\Omega(n^{1-\epsilon}/c^k)$  for any constant  $\epsilon > 0$  and  $c = 8e$ .*

*Proof.* It is easy to see that  $\text{GIP}_{n,k} = \mathcal{H}_k^{f,t}$ , where  $f = \text{PARITY}_m$  and  $t = \text{PARITY}_s$  and  $n = m \cdot s$ . From Lemma 7.1, the  $(5/6, \alpha)$ -approximate degree of  $f$  is  $d = \Omega(m^\gamma)$  for any  $0 < \gamma < \alpha < 1$ . Then by applying Theorem 3.1, since  $m = n/(4e)^k$  and for  $\alpha > \gamma = 1 - \epsilon$ , we have  $R_{1/3}^k(\mathcal{H}_k^{f,t}) \geq R_{1/3}^k(F_k^f)$  and the latter is  $\Omega(d/2^k)$  which is  $\Omega(n^{1-\epsilon}/c^k)$  for  $c = 8e$ .  $\square$

## References

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