# An additive combinatorics approach relating rank to communication complexity 

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#### Abstract

For a $\{0,1\}$-valued matrix $M$ let $\mathrm{CC}(M)$ denote the deterministic communication complexity of the boolean function associated with $M$. It is well-known since the work of Mehlhorn and Schmidt [STOC 1982] that $\mathrm{CC}(M)$ is bounded from above by $\operatorname{rank}(M)$ and from below by $\log \operatorname{rank}(M)$ where $\operatorname{rank}(M)$ denotes the rank of $M$ over the field of real numbers. Determining where in this range lies the true worst-case value of $\mathrm{CC}(M)$ is a fundamental open problem in communication complexity. The state of the art is $$
\log ^{1.631} \operatorname{rank}(M) \leq \mathrm{CC}(M) \leq 0.415 \operatorname{rank}(M),
$$ the lower bound is by Kushilevitz [unpublished, 1995] and the upper bound is due to Kotlov [Journal of Graph Theory, 1996]. Lovász and Saks [FOCS 1988] conjecture that $\mathrm{CC}(M)$ is closer to the lower bound, i.e., $\mathrm{CC}(M) \leq$ $\log ^{c}(\operatorname{rank}(M))$ for some absolute constant $c$ - this is the famous "log-rank conjecture" - but so far there has been no evidence to support it, even giving a slightly nontrivial ( $o(\operatorname{rank}(M))$ ) upper bound on the communication complexity.

Our main result is that, assuming the Polynomial Freiman-Ruzsa (PFR) conjecture in additive combinatorics, there exists a universal constant $c$ such that


$$
\mathrm{CC}(M) \leq c \cdot \operatorname{rank}(M) / \log \operatorname{rank}(M) .
$$

Although our bound is stated using the rank of $M$ over the reals, our proof goes by studying the problem over the finite field of size 2 , and there we bring to bear a number of new tools from additive combinatorics which we hope will facilitate further progress on this perplexing question.

In more detail, our proof is based on the study of the "approximate duality conjecture" which was suggested by Ben-Sasson and Zewi [STOC 2011] and studied there in connection to the PFR conjecture. First we improve the bounds on approximate duality assuming the PFR conjecture. Then we use the approximate duality conjecture (with improved bounds) to get our upper bound on the communication complexity of low-rank martices.

Index Terms-Communication Complexity; Log-rank Conjecture; Additive Combinatorics;

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## I. Introduction

This paper presents a new connection between communication complexity and additive combinatorics, showing that a well-known conjecture from additive combinatorics known as the Polynomial Freiman-Ruzsa Conjecture (PFR, in short), implies better upper bounds than currently known on the deterministic communication complexity of a boolean function in terms of the rank of its associated matrix. More precisely, our results show that the PFR Conjecture implies that every boolean function has communication complexity $O(\operatorname{rank}(M) / \log \operatorname{rank}(M))$ where $\operatorname{rank}(M)$ is the rank, over the reals, of the associated matrix. We view this result as interesting not only due to its being the first sublinear bound (and the first advance on this problem since 1997) but also because of its suggestion of a new connection between the two vibrant, yet seemingly unrelated, fields of communication complexity and additive combinatorics.

Our analysis relies on the study of approximate $d u$ ality, a concept closely related to the PFR Conjecture, which was introduced in [1]. Our main technical contribution improves the bounds on approximate duality, assuming the PFR Conjecture, and it does so with simpler proof than in [1]. We view this contribution as being of independent interest because of the growing number of applications of the "approximate duality method" to theoretical computer science. These include so-far the construction of bipartite Ramsey graphs and two-source extractors [1], communication complexity (this work), and the subsequent lower bounds for matching vector locally decodable codes [2].

## A. On communication complexity and matrix rank

In the two-party communication complexity model two parties - Alice and Bob - wish to compute a function $f: X \times Y \rightarrow\{0,1\}$ on inputs $x$ and $y$ where $x$ is known only to Alice and $y$ is known only to Bob. In order to compute the function $f$ they must exchange bits of information between each other according to some
(deterministic) protocol. The (deterministic) communication complexity of a protocol is the maximum total number of bits sent between the two parties, where the maximum is taken over all pairs of inputs $x, y$. We henceforth omit the adjective "deterministic" from our discourse because our results deal only with the deterministic model. The communication complexity of the function $f$, denoted by $\mathrm{CC}(f)$, is the minimum communication complexity of a protocol for $f$.

For many applications it is convenient to associate the function $f: X \times Y \rightarrow\{0,1\}$ with the matrix $M \in\{0,1\}^{X \times Y}$ whose $(x, y)$ entry equals $f(x, y)$. For a $\{0,1\}$-valued matrix $M$, let $\mathrm{CC}(M)$ denote the communication complexity of the boolean function associated with $M$. Let $\operatorname{rank}(M)$ denote the rank of $M$ over the reals. We will occasionally consider the rank of $M$ over the two-element field $\mathbb{F}_{2}$ and will denote this by $\operatorname{rank}_{\mathbb{F}_{2}}(M)$.

It is well-known since the work of Mehlhorn and Schmidt [3] that

$$
\begin{equation*}
\log \operatorname{rank}(M) \leq \mathrm{CC}(M) \leq \operatorname{rank}(M) \tag{1}
\end{equation*}
$$

and it is a fundamental question to find out what is the true worst-case dependency of $\mathrm{CC}(M)$ on the rank. The famous log-rank conjecture due to Lovász and Saks [4] postulates that the true answer is closer to the lower bound of (1).

Conjecture I. 1 (Log-rank). For every $\{0,1\}$-valued matrix $M \mathrm{CC}(M)=\log ^{O(1)} \operatorname{rank}(M)$.

Lovász and Saks also point out that the above conjecture has several other interesting equivalent formulations. One of them, due to Nuffelen [5] and Fajtlowicz [6], is the following:
Conjecture I.2. For every graph $G, \chi(\bar{G}) \leq$ $\log ^{O(1)} \operatorname{rank}(G)$, where $\chi(\bar{G})$ is the chromatic number of the complement of $G$, and $\operatorname{rank}(G)$ is the rank of the adjacency matrix of $G$ over the reals.

Though considerable effort has been made since 1982 in an attempt to narrow the gap between lower and upper bounds in (1), the state of the art is not far from where it was 30 years ago and currently stands at

$$
\begin{equation*}
\Omega\left(\log ^{\log _{3} 6} \operatorname{rank}(M)\right) \leq \mathrm{CC}(M) \leq \log (4 / 3) \operatorname{rank}(M) \tag{2}
\end{equation*}
$$

where $\log _{3} 6 \approx 1.63 \ldots$ and $\log (4 / 3) \approx 0.41 \ldots$ The upper bound is due to Kotlov [7] and improves on the previous best bound of $\mathrm{CC}(M) \leq$ $\operatorname{rank}(M) / 2$ by Kotlov and Lovász [8]. The lower bound is due to Kushilevitz (unpublished, cf. [9]) and improves on a previous bound of $\Omega\left(\log ^{\log _{2} 3} \operatorname{rank}(M)\right)=$ $\Omega\left(\log ^{1.58 \ldots} \operatorname{rank}(M)\right)$ due to Nisan and Wigderson [9].

Our main result is stated next. It assumes a wellknown conjecture from additive combinatorics - the

Polynomial Freiman-Ruzsa (PFR) conjecture - discussed in the next section.

Theorem I. 3 (Main). Assuming the PFR Conjecture I.5, for every $\{0,1\}$-valued matrix $M$

$$
\mathrm{CC}(M)=O(\operatorname{rank}(M) / \log \operatorname{rank}(M))
$$

## B. Additive combinatorics and the Polynomial FreimanRuzsa conjecture

Quoting the (current) Wikipedia definition, additive combinatorics studies "combinatorial estimates associated with the arithmetic operations of addition and subtraction". As such, it deals with a variety of problems that aim to 'quantify' the amount of additive structure in subsets of additive groups. One such a problem is that which is addressed by the Polynomial Freiman-Ruzsa conjecture (we shall encounter a different problem in additive combinatorics when we get to "approximate duality" later on).

For $A \subseteq \mathbb{F}_{2}^{n}$, let $A+A$ denote the sum-set of $A$

$$
A+A:=\left\{a+a^{\prime} \mid a, a^{\prime} \in A\right\}
$$

where addition is over $\mathbb{F}_{2}$. It is easy to see that $|A+A|=$ $|A|$ if and only if $A$ is an affine subspace of $\mathbb{F}_{2}^{n}$. The question addressed by the Freiman-Ruzsa Theorem is whether the ratio of $|A+A|$ to $|A|$ also 'approximates' the closeness of $A$ to being a subspace, or in other words, whether the fact that $A+A$ is small with respect to the size of $A$ also implies that span $(A)$ is small with respect to the size of $A$. The Freiman-Ruzsa Theorem [10] says that this is indeed the case.

Theorem $\mathbf{I} 4$ (Freiman-Ruzsa Theorem [10]). If $A \subseteq \mathbb{F}_{2}^{n}$ has $|A+A| \leq K|A|$, then $|\operatorname{span}(A)| \leq K^{2} 2^{K^{4}}|A|$.

The above theorem was improved in a series of works [11]-[13], culminating in the recent work [14] which proved an upper bound on the ratio $\frac{|\operatorname{span}(A)|}{|A|}$ of the form $2^{2 k} /(2 k)$. This bound can be seen to be tight (up to a multiplicative factor of 2 ) by letting $A=\left\{u_{1}, u_{2}, \ldots, u_{t}\right\}$, where $u_{1}, u_{2}, \ldots, u_{t} \in \mathbb{F}_{2}^{n}$ are linearly independent vectors. Then in this case we have $|A+A| \approx \frac{t}{2}|A|$, while $|\operatorname{span}(A)|=2^{t}$.

This example also shows that the ratio $\frac{|\operatorname{span}(A)|}{|A|}$ must depend exponentially on $K$. However, it does not rule out the existence of a large subset $A^{\prime} \subseteq A$ for which the ratio $\frac{\left|\operatorname{span}\left(A^{\prime}\right)\right|}{\left|A^{\prime}\right|}$ is just polynomial in $K$, and this is exactly what is suggested by the PFR Conjecture:
Conjecture I. 5 (Polynomial Freiman-Ruzsa (PFR)). There exists an absolute constant $r$, such that if $A \subset \mathbb{F}_{2}^{n}$ has $|A+A| \leq K|A|$, then there exists a subset $A^{\prime} \subseteq A$ of size at least $K^{-r}|A|$ such that $\left|\operatorname{span}\left(A^{\prime}\right)\right| \leq|A|$.

Note that the above conjecture implies that $\left|\operatorname{span}\left(A^{\prime}\right)\right| \leq|A| \leq K^{r}\left|A^{\prime}\right|$. The PFR conjecture
has many other interesting equivalent formulations, see the survey of Green [15] for some of them. It is conjectured to hold for subsets of general groups as well and not only for subsets of the group $\mathbb{F}_{2}^{n}$ but we will be interested only in the latter case. Significant progress on this conjecture has been achieved recently by Sanders [16], using new techniques developed by Croot and Sisask [17]. Sanders proved an upper bound on the ratio $\frac{\left|\operatorname{span}\left(A^{\prime}\right)\right|}{\left|A^{\prime}\right|}$ which is quasi-polynomial in $K$ :
Theorem I. 6 (Quasi-polynomial Freiman-Ruzsa Theorem (QFR) [16]). Let $A \subset \mathbb{F}_{2}^{n}$ be a set such that $|A+A| \leq K|A|$. Then there exists a subset $A^{\prime} \subseteq A$ of size at least $K^{-O\left(\log ^{3} K\right)}|A|$ such that $\left|\operatorname{span}\left(A^{\prime}\right)\right| \leq$ $|A|$.

We end this section by mentioning several other recent applications of the PFR Conjecture to theoretical computer science. The first application, due to Samorodnitsky [18], is to the area of low-degree testing, with further results by Lovett [19] and Green and Tao [20]. The second application is to the construction of twosource extractors due to Ben-Sasson and Zewi [1]. The latter paper also introduced the notion of approximate duality which plays a central role in our proof method as well. The approximate duality method has recently found another application to proving lower bounds on locally decodable matching vector codes in the subsequent work by Bhowmick, Dvir and Lovett [2]. In the next section we describe the approximate duality conjecture and our new contributions to its study.

## C. Approximate duality

Our main technical contribution (Lemma I.10) is improving the bounds on approximate duality, assuming the PFR conjecture. The new bound lies at the heart of our proof of the Main Theorem I.3. We believe that Lemma I. 10 and its proof are of independent interest since they improve and simplify the proof of [1], and have already found new interesting applications to the study of locally decodable codes [2].

For $A, B \subseteq \mathbb{F}_{2}^{n}$, we define the duality measure of $A, B$ in (3) as an estimate of how 'close' this pair is to being dual

$$
\begin{equation*}
D(A, B):=\left|\mathbb{E}_{a \in A, b \in B}\left[(-1)^{\langle a, b\rangle_{2}}\right]\right| \tag{3}
\end{equation*}
$$

where $\langle a, b\rangle_{2}$ denotes the binary inner-product of $a, b$ over $\mathbb{F}_{2}$, defined by $\langle a, b\rangle_{2}=\sum_{i=1}^{n} a_{i} \cdot b_{i}$ where all arithmetic operations are in $\mathbb{F}_{2}$.
Remark I.7. The duality measure can be alternatively defined as the discrepancy of the inner product function on the rectangle $A \times B$ (up to a normalization factor of $\frac{2^{n}}{|A||B|}$ ). Nevertheless we chose to use the term 'duality
measure' instead of 'discrepancy' because of the algebraic context in which we use it, as explained below.

It can be verified that if $D(A, B)=1$ then $A$ is contained in an affine shift of $B^{\perp}$ which is the space dual to the linear $\mathbb{F}_{2}$-span of $B$. The question is what can be said about the structure of $A, B$ when $D(A, B)$ is sufficiently large, but strictly smaller than 1 . The following theorem from [1] says that if the duality measure is a constant very close to 1 (though strictly smaller than 1) then there exist relatively large subsets $A^{\prime} \subseteq A, B^{\prime} \subseteq B$, such that $D\left(A^{\prime}, B^{\prime}\right)=1$.

Theorem I. 8 (Approximate duality for nearly-dual sets, [1]). For every $\delta>0$ there exists a constant $\epsilon>0$ that depends only on $\delta$, such that if $A, B \subseteq \mathbb{F}_{2}^{n}$ satisfy $D(A, B) \geq 1-\epsilon$, then there exist subsets $A^{\prime} \subseteq A,\left|A^{\prime}\right| \geq \frac{1}{4}|A|$ and $B^{\prime} \subseteq B,\left|B^{\prime}\right| \geq 2^{-\delta n}|B|$, such that $D\left(A^{\prime}, B^{\prime}\right)=1$.

It is conjectured that a similar result holds also when the duality measure is relatively small, and in particular when it tends to zero as $n$ goes to infinity. Furthermore, the following theorem from [1] gives support to this conjecture, by showing that such bounds indeed follow from the PFR conjecture.

Theorem I. 9 (Approximate duality assuming PFR, exponential loss [1]). Assuming the PFR Conjecture I.5, for every pair of constants $\alpha>\delta>0$ there exists a constant $\zeta>0$, depending only on $\alpha$ and $\delta$, such that the following holds. If $A, B \subseteq \mathbb{F}_{2}^{n}$ satisfy $|A|,|B|>2^{\alpha n}$ and $D(A, B) \geq 2^{-\zeta n}$, then there exist subsets $A^{\prime} \subseteq A$, $\left|A^{\prime}\right| \geq 2^{-\delta n}|A|$ and $B^{\prime} \subseteq B,\left|B^{\prime}\right| \geq 2^{-\delta n}|B|$ such that $D\left(A^{\prime}, B^{\prime}\right)=1$.

Our main technical contribution is the following generalization of the above theorem.

Lemma I. 10 (Main technical lemma). Assuming the PFR Conjecture I. 5 there exists a universal integer $r$ such that the following holds. Suppose that $A, B \subseteq$ $\{0,1\}^{n}$ satisfy $D(A, B) \geq \epsilon$. Then for every $K \geq 1$ and $t=n / \log K$, there exist subsets $A^{\prime}, B^{\prime}$ of $A, B$ respectively such that $D\left(A^{\prime}, B^{\prime}\right)=1$, and
$\frac{\left|A^{\prime}\right|}{|A|} \geq\left(\left(\frac{(\epsilon / 2)^{2^{t}}}{n K}\right)(4 n)^{-t}\right)^{r}, \frac{\left|B^{\prime}\right|}{|B|} \geq\left(\left(\frac{(\epsilon / 2)^{2^{t}}}{n K}\right) 2^{-t}\right)^{r}$

The proof of the above lemma appears in Section II. To see that it is indeed a generalization of Theorem I. 9 set $K=2^{\delta n /(3 r)}, t=3 r / \delta, \zeta=\delta /\left(3 r \cdot 2^{t}\right)=$ $\delta /\left(3 r \cdot 2^{3 r / \delta}\right), \epsilon=2^{-\zeta n}$, and note that in this case the above lemma assures the existence of $\left|A^{\prime}\right| \geq 2^{-\delta n}|A|$, $\left|B^{\prime}\right| \geq 2^{-\delta n}|B|$ such that $D\left(A^{\prime}, B^{\prime}\right)=1$. Note that Lemma I. 10 actually improves on the previous Theorem I. 9 even in this exponential range of parameters in that
its parameters do not depend on the sizes of the sets $A$ and $B$ as was the case in Theorem I.9.

However, the main significance of Lemma I. 10 is that it allows one to tradeoff the loss in the sizes of $A^{\prime}$ and $B^{\prime}$ with the value of $\epsilon$ for a wider range of parameters. More specifically it allows one to achieve a loss in the sizes of $A^{\prime}$ and $B^{\prime}$ which is only sub-exponential in $n$ by requiring $\epsilon$ be a bit larger. In particular, the following corollary of Lemma I. 10 will enable us to prove the new upper bound of $O(\operatorname{rank}(M) / \log \operatorname{rank}(M))$ on the communication complexity of $\{0,1\}$-valued matrices assuming the PFR conjecture.

Corollary 1.11 (Approximate duality assuming PFR, sub-exponential loss). Suppose that $A, B \subseteq \mathbb{F}_{2}^{n}$ satisfy $D(A, B) \geq 2^{-\sqrt{n}}$. Then assuming the PFR conjecture I.5, there exist subsets $A^{\prime}, B^{\prime}$ of $A, B$ respectively such that $D\left(A^{\prime}, B^{\prime}\right)=1$, and $\left|A^{\prime}\right| \geq 2^{-c n / \log n}|A|,\left|B^{\prime}\right| \geq$ $2^{-c n / \log n}|B|$ for some absolute constant $c$.

Proof of Corollary I.11: Follows from Lemma I. 10 by setting $K=2^{4 n / \log n}, t=\frac{\log n}{4}, \epsilon=2^{-\sqrt{n}}$.

Note that in Corollary I. 11 the ratios $\left|A^{\prime}\right| /|A|$, $\left|B^{\prime}\right| /|B|$ are bounded from below by $2^{-c n / \log n}$, whereas in Theorem I. 9 we only get a smaller bound of the form $2^{-\delta n}$ for some constant $\delta>0$. However, this improvement comes with a requirement that the duality measure $D(A, B)$ is larger - in the above corollary we require that it is at least $2^{-\sqrt{n}}$ while in Theorem I. 9 we only require it to be at least $2^{-\zeta n} \ll 2^{-\sqrt{n}}$. We note that the bound $D(A, B) \geq 2^{-\sqrt{n}}$ can be replaced by $D(A, B) \geq \exp \left(-n^{1-\epsilon}\right)$ for any $\epsilon>0$ at the price of a larger constant $c=c(\epsilon)$.

Remark I. 12 (Exponential loss necessary). Generally speaking, the bound on $\min \left\{\frac{\left|A^{\prime}\right|}{|A|}, \frac{\left|B^{\prime}\right|}{|B|}\right\}$ - which in Corollary I. 11 above is $2^{-c n / \log n}$ - cannot be improved beyond $2^{-O(\sqrt{n})}$ even if we assume $D(A, B)>$ 0.99. To see this take $A=B=\binom{n}{c^{\prime} \sqrt{n}}$ to be the set of all $\{0,1\}$-vectors with exactly $c^{\prime} \sqrt{n}$ ones, where $c^{\prime}$ is a sufficiently small positive constant that guarantees $D(A, B) \geq 0.99$. It was shown in [21] that if $A^{\prime} \subset$ $A, B^{\prime} \subset B$ satisfy $D\left(A^{\prime}, B^{\prime}\right)=1$ then the smaller set of $A^{\prime}, B^{\prime}$ is of size $2^{-\Omega(\sqrt{n})} \cdot|A|$.

We stress that a benefit of the proof of Lemma I. 10 is that it simplifies the original proof of Theorem I. 9 in [1]. Indeed, we believe that the presentation of the proof that appears in this paper is clearer and less involved than that in [1]. Also, the fact that the parameters in Lemma I .10 do not depend on the sizes of $A$ and $B$ allows us to deduce new equivalence between approximate duality and the PFR conjecture in the exponential range that was not previously known. We elaborate on this equivalence in the full version of this paper [22].

## D. Proof overview

First we show how our Main Theorem I. 3 is deduced from the improved bounds on approximate duality in Corollary I.11. Then we give an overview of the proof of Lemma I. 10 itself.
a) From approximate duality to communication complexity upper bounds.: We follow the approach of Nisan and Wigderson from [9]. Let the size of a matrix $M$ be the number of entries in it and if $M$ is $\{0,1\}$ valued let $\delta(M)$ denote its (normalized) discrepancy, defined as the absolute value of the difference between the fraction of zero-entries and one-entries in $M$. Informally, discrepancy measures how "unbalanced" is $M$, with $\delta(M)=1$ when $M$ is monochromatic - all entries have the same value - and $\delta(M)=0$ when $M$ is completely balanced.

Returning to the work of [9], they observed that to prove the log-rank conjecture it suffices to show that a $\{0,1\}$-valued matrix $M$ of rank $r$ always contains a monochromatic sub-matrix of size $|M| / \operatorname{qpoly}(r)$ where qpoly $(r)=r^{\log ^{O(1)} r}$ means quasi-polynomial in $r$. Additionally, they used spectral techniques (i.e., arguing about the eigenvectors and eigenvalues of $M$ ) to show that any $\{0,1\}$-valued matrix $M$ of rank $r$ contains a relatively large submatrix $M^{\prime}$ - of size at least $|M| / r^{3 / 2}$ - that is somewhat biased - its discrepancy is at least $1 / r^{3 / 2}$. We show, using tools from additive combinatorics, that $M^{\prime}$ in fact contains a pretty large monochromatic submatrix (though not large enough to deduce the log-rank conjecture).

To this end we start by working over the two-element field $\mathbb{F}_{2}$. This seems a bit counter-intuitive because the log-rank conjecture is false over $\mathbb{F}_{2}$. The canonical counterexample is the inner product function $I P(x, y)=$ $\langle x, y\rangle_{2}$ - It is well-known (see e.g. [23][Chapters 1.3., 2.5.]) that $\operatorname{rank}_{\mathbb{F}_{2}}\left(M_{I P}\right)=n$ while $\mathrm{CC}(I P)=n$. However, rather than studying $M$ over $\mathbb{F}_{2}$ we focus on the biased submatrix $M^{\prime}$ and things change dramatically. (As a sanity-check notice that $M_{I P}$ does not contain large biased submatrices and this does not contradict the work of [9] because the rank of $M_{I P}$ over the reals is $2^{n}-1$.)

Thus, our starting point is a large submatrix $M^{\prime}$ that has large discrepancy. It is well-known that $\operatorname{rank}_{\mathbb{F}_{2}}\left(M^{\prime}\right) \leq \operatorname{rank}\left(M^{\prime}\right) \leq r$ and that this implies $M^{\prime}$ can be written as $M^{\prime}=A^{\top} \cdot B$ where $A, B$ are matrices whose columns are vectors in $\mathbb{F}_{2}^{r}$. Viewing each of $A, B$ as the set of its columns, we have in hand two sets that have a large duality measure as defined in (3), namely, $D(A, B)=\delta\left(M^{\prime}\right) \geq 1 / r^{3 / 2}$. This is the setting in which we apply Corollary I. 11 and deduce that $A, B$ contain relatively large subsets $A^{\prime}, B^{\prime}$ with $D\left(A^{\prime}, B^{\prime}\right)=1$. One can now verify that the submatrix of $M^{\prime}$ whose rows and columns are indexed by $A^{\prime}, B^{\prime}$ respectively is indeed
monochromatic, as needed. We point out that to get our bounds we need to be able to find monochromatic submatrices of $M^{\prime}$ even when $M^{\prime}$ is both small and skewed (i.e., has many more columns than rows or vice versa). Fortunately, Corollary I. 11 is robust enough to use in such settings.
b) Improved bounds on approximate duality assuming PFR.: We briefly sketch the proof of our Main Technical Lemma I.10. We use the spectrum of a set as defined in [24, Chapter 4]:

Definition I. 13 (Spectrum). For a set $B \subseteq \mathbb{F}_{2}^{n}$ and $\alpha \in$ $[0,1]$ let the $\alpha$-spectrum of $B$ be the set

$$
\begin{equation*}
\operatorname{Spec}_{\alpha}(B):=\left\{x \in \mathbb{F}_{2}^{n}| | \mathbb{E}_{b \in B}\left[(-1)^{\langle x, b\rangle_{2}}\right] \mid \geq \alpha\right\} \tag{5}
\end{equation*}
$$

Notice that $A \subseteq \operatorname{Spec}_{\epsilon}(B)$ implies $D(A, B) \geq \epsilon$ (cf. (3)). In the other direction, Markov's inequality can be used to deduce that $D(A, B) \geq \epsilon$ implies the existence of $A^{\prime} \subseteq A$ of relatively large size - $\left|A^{\prime}\right| \geq \frac{\epsilon}{2}|A|$ such that $A^{\prime} \subseteq \operatorname{Spec}_{\epsilon / 2}(B)$. To prove our lemma we start with $A_{1}=A^{\prime}$ and establish a sequence of sets

$$
A_{2} \subseteq A_{1}+A_{1}, \quad A_{3} \subseteq A_{2}+A_{2}, \ldots
$$

such that $A_{i} \subseteq \operatorname{Spec}_{\epsilon_{i}}(B)$ for all $i$. This holds by construction for $A_{1}$ with $\epsilon_{1}=\epsilon / 2$, and we show that it is maintained throughout the sequence for increasingly smaller values of $\epsilon_{i}$ (we shall use $\epsilon_{i}=\epsilon_{i-1}^{2}$ ).

Moving our problem from the field of real numbers to the two-element field $\mathbb{F}_{2}$ now pays off. Each $A_{i}$ is of size at most $2^{n}$ so there must be an index $i \leq n / \log K$ for which $\left|A_{i+1}\right| \leq K\left|A_{i}\right|$, let $t$ be the minimal such index. We use the PFR conjecture together with the Balog-Szemerédi-Gowers Theorem II. 1 from additive combinatorics to show that our assumption that $\left|A_{t+1}\right| \leq K\left|A_{t}\right|$ implies that a large subset $A_{t}^{\prime \prime}$ of $A_{t}$ has small span (over $\mathbb{F}_{2}$ ).

We now have in hand a set $A_{t}^{\prime \prime}$ which is a relatively large fraction of its span and additionally satisfies $D\left(A_{t}^{\prime \prime}, B\right) \geq \epsilon_{t}$ because by construction $A_{t}^{\prime \prime} \subseteq$ $\operatorname{Spec}_{\epsilon_{t}}(B)$. We use an approximate duality claim from [1] (Lemma II.2) which applies when one of the sets is a large fraction of its span (in our case the set which is a large fraction of its span is $A_{t}^{\prime \prime}$ ). This claim says that $A_{t}^{\prime \prime}$ and $B$ each contain relatively large subsets $A_{t}^{\prime}, B_{t}^{\prime}$ satisfying $D\left(A_{t}^{\prime}, B_{t}^{\prime}\right)=1$. Finally, recalling $A_{t}^{\prime}$ is a (carefully chosen) subset of $A_{t-1}+A_{t-1}$, we argue that $A_{t-1}$ contains a relatively large subset $A_{t-1}^{\prime}$ that is "dual" to a large subset $B_{t-1}^{\prime}$ of $B_{t}^{\prime}$, where by "dual" we mean $D\left(A_{t-1}^{\prime}, B_{t-1}^{\prime}\right)=1$ (in other words $A_{t-1}^{\prime}$ is contained in an affine shift of the space dual to $\operatorname{span}\left(B_{t-1}^{\prime}\right)$ ). We continue in this manner to find pairs of "dual" subsets for $t-2, t-3, \ldots, 1$ at which point we have found a pair of "dual" subsets of $A, B$ that have relatively large size, thereby completing the proof.

## E. Discussion and directions for future research

The new connection between additive combinatorics and communication complexity seems to us worthy of further study. In particular, the exciting recent advances in additive combinatorics [14], [16], [17] use a rich palette of tools that may yield further insights into problems in communication complexity. We end this section by briefly pointing out a few directions we find interesting.
c) Improved unconditional bounds on communication complexity: Given the recent QFR result of [16] (Theorem I.6) which comes very close to proving the PFR conjecture, it is interesting to see if it implies any unconditional improvement on communication complexity of low-rank matrices. Looking at our proof of Lemma I.10, we apply the PFR conjecture to a subset $A_{t}^{\prime}$ of $A_{t}$ which satisfies $\left|A_{t}^{\prime}+A_{t}^{\prime}\right| \leq K^{\prime}\left|A_{t}^{\prime}\right|$ for $K^{\prime} \approx K / \epsilon^{2^{t}}$. For $\epsilon<\frac{1}{2}$ this gives a non-trivial bound only if $t=O(\log n)$. Since $t$ could be as large as $n / \log K$ we are forced to choose $K=2^{\Omega(n / \log n)}$ which implies in turn $K^{\prime}=2^{\Omega(n / \log n)}$. Thus, Sander's QFR Theorem I. 6 does not yield any non-trivial bounds in our case. However, for purposes of improving the unconditional upper bound of Kotlov (cf. 2) say, to $C C(M) \leq \operatorname{rank}(M) / 4$, it suffices to improve the loss in the size of $A$ in Theorem I. 6 from $K^{-O\left(\log ^{3} K\right)}$ to $K^{-c \log K}$ for a sufficiently small constant $c$.
d) Improved conditional bounds: The bounds on approximate duality in Corllary I. 11 can possibly be significantly improved. For all we know, the exponential loss of $2^{-O(\sqrt{n})}$ shown in Remark I. 12 may be tight, and this would lead to an improved version of Corollary I. 11 in which the sizes of $\left|A^{\prime}\right|,\left|B^{\prime}\right|$ are a $2^{-O(\sqrt{n})}$ fraction of $A$ and $B$ respectively, instead of the $2^{-O(n / \log n)}$ loss we currently have. Such a result would translate directly to an upper bound on communication complexity of the form $\mathrm{CC}(M) \leq O(\sqrt{\operatorname{rank}(M)})$. In order to make further progress one might want to also consider working over finite fields that are larger than 2 , or over the reals. As a first step in this direction, one may wish to investigate whether there are interesting approximate duality statements over such fields.
e) Does the log-rank conjecture imply the PFR conjecture?: Alternatively, does it have any other nontrivial consequences in additive combinatorics? We believe the answer to this question is positive and make a step in this direction by showing an equivalence between approximate duality and PFR statements in the exponential range, namely, when the losses in the sizes of sets in both approximate duality and PFR is exponential in $n$ (See Section 4 in the full version of this paper [22] for an exact statement and details of the proof.)

## F. Paper organization.

The next section contains the proof of the Main Technical Lemma I.10. The proof of Main Theorem I. 3 given Corollary I. 11 appears in Section III.

## II. IMPROVED BOUNDS ON APPROXIMATE DUALITY ASSUMING PFR

In this section we prove our Main Technical Lemma I.10. We start with some additive combinatorics preliminaries.
f) Additive combinatorics preliminaries: In what follows all arithmetic operations are taken over $\mathbb{F}_{2}$. For the proof of Lemma I. 10 we need two other theorems from additive combinatorics. The first is the well-known Balog-Szemerédi-Gowers Theorem of [25], [26].
Theorem II. 1 (Balog-Szemerédi-Gowers). There exist fixed polynomials $f(x, y), g(x, y)$ such that the following holds for every subset $A$ of an abelian additive group. If A satisfies $\operatorname{Pr}_{a, a^{\prime} \in A}\left[a+a^{\prime} \in S\right] \geq 1 / K$ for $|S| \leq C|A|$, then one can find a subset $A^{\prime} \subseteq A$ such that $\left|A^{\prime}\right| \geq$ $|A| / f(K, C)$, and $\left|A^{\prime}+A^{\prime}\right| \leq g(K, C)|A|$.

The second is a lemma from [1] which can be seen as an approximate duality statement which applies when one of the sets has small span:

Lemma II. 2 (Approximate-duality for sets with small span, [1]). If $D(A, B) \geq \epsilon$, then there exist subsets $A^{\prime} \subseteq$ $A, B^{\prime} \subseteq B,\left|A^{\prime}\right| \geq \frac{\epsilon}{4}|A|,\left|B^{\prime}\right| \geq \frac{\epsilon^{2}}{4} \frac{|A|}{|\operatorname{span}(A)|}|B|$, such that $D\left(A^{\prime}, B^{\prime}\right)=1$. If $A \subseteq \operatorname{Spec}_{\epsilon}(B)$ then we have $\left|A^{\prime}\right| \geq|A| / 2$ and $\left|B^{\prime}\right| \geq \epsilon^{2} \frac{|A|}{|\operatorname{span}(A)|}|B|$ in the statement above.

Recall the definition of the spectrum given in (5):

$$
\operatorname{Spec}_{\alpha}(B):=\left\{x \in \mathbb{F}_{2}^{n}| | \mathbb{E}_{b \in B}\left[(-1)^{\langle x, b\rangle_{2}}\right] \mid \geq \alpha\right\} .
$$

Finally, for $S \subset \mathbb{F}_{2}^{n}$ and $x \in \mathbb{F}_{2}^{n}$ let $\operatorname{rep}_{S}(x)$ be the number of different representations of $x$ as an element of the form $s+s^{\prime}$ where $s, s^{\prime} \in S . \operatorname{rep}_{S}(x)$ can also be written, up to a normalization factor, as $1_{S} * 1_{S}(x)$ where $1_{S}$ is the indicating function of the set $S$ and * denotes convolution.
g) Proof overview: We construct a decreasing sequence of constants

$$
\epsilon_{1}=\epsilon / 2, \epsilon_{2}=\epsilon_{1}^{2} / 2, \epsilon_{3}=\epsilon_{2}^{2} / 2, \ldots
$$

and a sequence of sets $A_{1}:=A \cap \operatorname{Spec}_{\epsilon_{1}}(B), \quad A_{2} \subseteq$ $\left(A_{1}+A_{1}\right) \cap \operatorname{Spec}_{\epsilon_{2}}(B), \quad A_{3} \subseteq\left(A_{2}+A_{2}\right) \cap$ $\operatorname{Spec}_{\epsilon_{3}}(B), \ldots$

Since each of the sets in the sequence is of size at most $2^{n}$ there must be an index $i \leq n / \log K$ for which

$$
\begin{equation*}
\left|A_{i+1}\right| \leq K\left|A_{i}\right| \tag{6}
\end{equation*}
$$

and let $t$ be the minimal such index. The PFR Conjecture I. 5 together with the Balog-Szemerédi-Gowers

Theorem II. 1 will be used to deduce from (6) that a large subset $A_{t}^{\prime \prime}$ of $A_{t}$ has small span. Applying Lemma II. 2 to the sets $A_{t}^{\prime \prime}$ and $B$ implies the existence of large subsets $A_{t}^{\prime} \subseteq A_{t}$ and $B_{t}^{\prime} \subseteq B$ such that $D\left(A_{t}^{\prime}, B_{t}^{\prime}\right)=1$. Finally we argue inductively for $i=t-1, t-2, \ldots, 1$ that there exist large subsets $A_{i}^{\prime} \subseteq A_{i}$ and $B_{i}^{\prime} \subseteq B$ such that $D\left(A_{i}^{\prime}, B_{i}^{\prime}\right)=1$. The desired conclusion will follow from the $i=1$ case. To be able to "pull back" and construct a pair of large sets $A_{i-1}^{\prime}, B_{i-1}^{\prime}$ from the pair $A_{i}^{\prime}, B_{i}^{\prime}$ we make sure every element in $A_{i}$ is the sum of roughly the same number of pairs in $A_{i-1} \times A_{i-1}$.
h) The sequence of sets: Let $\epsilon_{1}:=\epsilon / 2, A_{1}:=$ $A \cap \operatorname{Spec}_{\epsilon_{1}}(B)$. Assuming $A_{i-1}, \epsilon_{i-1}$ have been defined set $\epsilon_{i}=\epsilon_{i-1}^{2} / 2$ and let $j_{i} \in\{0, \ldots, n-1\}$ be an integer index which maximizes the size of

$$
\left\{\begin{array}{l|l}
\left(a, a^{\prime}\right) \in A_{i-1} & \begin{array}{l}
a+a^{\prime} \in \operatorname{Spec}_{\epsilon_{i}}(B) \text { and } \\
2^{j_{i}} \leq \operatorname{rep}_{A_{i-1}}\left(a+a^{\prime}\right) \leq 2^{j_{i}+1}
\end{array} \tag{7}
\end{array}\right\}
$$

and set
$A_{i}:=\left\{\begin{array}{l|l}a+a^{\prime} & \begin{array}{l}a, a^{\prime} \in A_{i-1}, a+a^{\prime} \in \operatorname{Spec}_{\epsilon_{i}}(B) \text { and } \\ 2^{j_{i}} \leq \operatorname{rep}_{A_{i-1}}\left(a+a^{\prime}\right) \leq 2^{j_{i}+1}\end{array}\end{array}\right\}$

Claim II.3. For $i=1$ we have $\left|A_{1}\right| \geq(\epsilon / 2)|A|$. For $i>1$ we have

$$
\begin{equation*}
\operatorname{Pr}_{a, a^{\prime} \in A_{i-1}}\left[a+a^{\prime} \in A_{i}\right] \geq \epsilon_{i} / n \tag{9}
\end{equation*}
$$

and additionally

$$
\begin{equation*}
\left|A_{i}\right| \geq \frac{\epsilon_{i}}{2^{j_{i}+1} n}\left|A_{i-1}\right|^{2} \tag{10}
\end{equation*}
$$

Proof: The case of $i=1$ follows directly from Markov's inequality. For larger $i$ we argue that

$$
\operatorname{Pr}_{a, a^{\prime} \in A_{i-1}}\left[a+a^{\prime} \in \operatorname{Spec}_{\epsilon_{i}}(B)\right] \geq \epsilon_{i}
$$

To see this use Cauchy-Schwarz to get

$$
\begin{aligned}
\mathbb{E}_{a, a^{\prime} \in A_{i-1}} & \left|\mathbb{E}_{b \in B}(-1)^{\left\langle a+a^{\prime}, b\right\rangle}\right| \\
= & E_{b \in B}\left(\mathbb{E}_{a \in A_{i-1}}\left[(-1)^{\langle a, b\rangle}\right]\right)^{2} \\
\geq & \left(\mathbb{E}_{a \in A_{i-1}, b \in B}\left[(-1)^{\langle a, b\rangle}\right]\right)^{2}=\epsilon_{i-1}^{2}
\end{aligned}
$$

and apply Markov's inequality to deduce that an $\epsilon_{i}{ }^{-}$ fraction of $\left(a, a^{\prime}\right) \in A_{i-1} \times A_{i-1}$ sum to an element of $\operatorname{Spec}_{\epsilon_{i}}(B)$. Selecting $j_{i}$ to maximize (7) yields inequality (9). Since every element $x \in A_{i}$ can be represented as $x=a+a^{\prime}$ with $a, a^{\prime} \in A_{i-1}$ in at most $2^{j_{i}+1}$ different ways we deduce (10) from (9) and complete the proof.
i) The inductive claim: Let $t$ be the minimal index such that $\left|A_{t+1}\right| \leq K\left|A_{t}\right|$ and note that $t \leq n / \log K$ because all sets $A_{i}$ are contained in $\mathbb{F}_{2}^{n}$. We shall prove the following claim by backward induction.

Claim II. 4 (Inductive claim). For $i=t, t-1, \ldots, 1$ there exist subsets

$$
A_{i}^{\prime} \subseteq A_{i}, \quad B_{i}^{\prime} \subseteq B
$$

such that $D\left(A_{i}^{\prime}, B_{i}^{\prime}\right)=1$ and $A_{i}^{\prime}, B_{i}^{\prime}$ are not too small:

$$
\begin{gathered}
\left|A_{i}^{\prime}\right| \geq \operatorname{poly}\left(\frac{\epsilon_{t+1}}{n K}\right)(4 n)^{-(t-i)}\left(\prod_{\ell=i}^{t} \epsilon_{\ell+1}\right)\left|A_{i}\right| \\
\left|B_{i}^{\prime}\right| \geq \operatorname{poly}\left(\frac{\epsilon_{t+1}}{n K}\right) 2^{-(t-i)}|B|
\end{gathered}
$$

We split the proof of the claim to two parts. The base case (Proposition II.5) is proved using the tools from additive combinatorics listed in the beginning of this section. The inductive step is proved in Proposition II. 6 using a graph construction. Before proving Claim II. 4 we show how it implies Lemma I.10.

Proof of Main Technical Lemma I.10: Set $i=1$ in Claim II. 4 above. Recall that $\epsilon_{i+1}=\epsilon_{i}^{2} / 2$ for all $i$, so

$$
\epsilon_{\ell+1}=\epsilon^{2^{\ell}} / 2^{2^{\ell}-1} \geq(\epsilon / 2)^{2^{\ell}}
$$

Thus we have $\epsilon_{t+1} \geq(\epsilon / 2)^{2^{t}}$ and $\prod_{\ell=1}^{t} \epsilon_{\ell+1} \geq$ $(\epsilon / 2)^{2^{t+1}}$. This gives the bounds on $A^{\prime}, B^{\prime}$ stated in (4).

Proposition II. 5 (Base case of Claim II. $4(i=t)$ ). There exist subsets $A_{t}^{\prime} \subseteq A_{t}, B_{t}^{\prime} \subseteq B_{t}$ such that $D\left(A_{t}^{\prime}, B_{t}^{\prime}\right)=$ 1 and $A_{t}^{\prime}, B_{t}^{\prime}$ are not too small:

$$
\begin{aligned}
& \left|A_{t}^{\prime}\right| \geq \operatorname{poly}\left(\frac{\epsilon_{t+1}}{n K}\right)\left|A_{t}\right| \\
& \left|B_{t}^{\prime}\right| \geq \operatorname{poly}\left(\frac{\epsilon_{t+1}}{n K}\right)|B|
\end{aligned}
$$

Proof:
By assumption $\left|A_{t+1}\right| \leq K\left|A_{t}\right|$ and $P r_{a, a^{\prime} \in A_{t}}[a+$ $\left.a^{\prime} \in A_{t+1}\right] \geq \epsilon_{t+1} / n$ by (10). Hence we can apply the Balog-Szemerédi-Gowers Theorem (Theorem II.1) to the set $A_{t}$ to obtain a subset $\tilde{A}_{t} \subseteq A_{t}$ such that

$$
\left|\tilde{A}_{t}\right| \geq \operatorname{poly}\left(\frac{\epsilon_{t+1}}{n K}\right)\left|A_{t}\right|
$$

and

$$
\left|\tilde{A}_{t}+\tilde{A}_{t}\right| \leq \operatorname{poly}\left(\frac{n K}{\epsilon_{t+1}}\right)\left|A_{t}\right|=\operatorname{poly}\left(\frac{n K}{\epsilon_{t+1}}\right)\left|\tilde{A}_{t}\right|
$$

Now we can apply the PFR Conjecture I. 5 to the set $\tilde{A}_{t}$ which gives a subset $A_{t}^{\prime \prime} \subseteq \tilde{A}_{t}$ such that

$$
\left|A_{t}^{\prime \prime}\right| \geq \operatorname{poly}\left(\frac{\epsilon_{t+1}}{n K}\right)\left|\tilde{A}_{t}\right|=\operatorname{poly}\left(\frac{\epsilon_{t+1}}{n K}\right)\left|A_{t}\right|
$$

and

$$
\left|\operatorname{span}\left(A_{t}^{\prime \prime}\right)\right| \leq\left|\tilde{A}_{t}\right|=\operatorname{poly}\left(\frac{n K}{\epsilon_{t+1}}\right)\left|A_{t}^{\prime \prime}\right|
$$

Recall that $A_{t}^{\prime \prime} \subseteq \operatorname{Spec}_{\epsilon_{t}}(B)$, and in particular $D\left(A_{t}^{\prime \prime}, B\right) \geq \epsilon_{t}$. Applying Lemma II. 2 to the sets $A_{t}^{\prime \prime}$ and $B$ we conclude that there exist subsets $A_{t}^{\prime} \subseteq A_{t}^{\prime \prime}$, $B^{\prime} \subseteq B$ such that $D\left(A_{t}^{\prime}, B^{\prime}\right)=1$, and which satisfy $\left|A_{t}^{\prime}\right| \geq \frac{1}{2}\left|A_{t}^{\prime \prime}\right|$ and

$$
\left|B_{t}^{\prime}\right| \geq \epsilon_{t}^{2} \frac{\left|A_{t}^{\prime \prime}\right|}{\left|\operatorname{span}\left(A_{t}^{\prime \prime}\right)\right|}|B|=\operatorname{poly}\left(\frac{\epsilon_{t+1}}{n K}\right)|B|
$$

This completes the proof of the base case.
Proposition II. 6 (Inductive step of Claim II.4). For every $i=t-1, \ldots, 1$ there exist subsets $A_{i}^{\prime} \subseteq A_{i}$, $B_{i}^{\prime} \subseteq B$ such that $D\left(A_{i}^{\prime}, B_{i}^{\prime}\right)=1$ and $A_{i}^{\prime}, B_{i}^{\prime}$ are not too small:

$$
\begin{gathered}
\left|A_{i}^{\prime}\right| \geq \operatorname{poly}\left(\frac{\epsilon_{t+1}}{n K}\right)(4 n)^{-(t-i)}\left(\prod_{\ell=i}^{t} \epsilon_{\ell+1}\right)\left|A_{i}\right| \\
\left|B_{i}^{\prime}\right| \geq \operatorname{poly}\left(\frac{\epsilon_{t+1}}{n K}\right) 2^{-(t-i)}|B|
\end{gathered}
$$

Proof: Suppose that the claim is true for $i$ and argue it holds for index $i-1$. Let $G=\left(A_{i-1}, E\right)$ be the graph whose vertices are the elements in $A_{i-1}$, and $\left(a, a^{\prime}\right)$ is an edge if $a+a^{\prime} \in A_{i}^{\prime}$. We bound the number of edges in this graph from below. Recall from (8) that every $a \in A_{i}^{\prime}$ (where $A_{i}^{\prime} \subseteq A_{i}$ ) satisfies $2^{j_{i}} \leq \operatorname{rep}_{A_{i-1}}(a) \leq 2^{j_{i}+1}$. Using this we get

$$
\begin{aligned}
|E| & \geq 2^{j_{i}} \cdot\left|A_{i}^{\prime}\right| \\
& \geq 2^{j_{i}}\left(\frac{\epsilon_{t+1}}{n K}\right)^{O(1)} \frac{\left|A_{i}\right|}{(4 n)^{(t-i)}} \prod_{\ell=i}^{t} \epsilon_{\ell+1} \\
& \geq 2^{j_{i}}\left(\frac{\epsilon_{t+1}}{n K}\right)^{O(1)} \frac{\left|A_{i-1}\right|^{2}}{(4 n)^{(t-i)}} \frac{\epsilon_{i}}{2^{j_{i}+1} n} \prod_{\ell=i}^{t} \epsilon_{\ell+1} \\
& =2 \cdot\left(\frac{\epsilon_{t+1}}{n K}\right)^{O(1)} \frac{\left|A_{i-1}\right|^{2}}{(4 n)^{(t-(i-1))}} \prod_{\ell=i-1}^{t} \epsilon_{\ell+1}
\end{aligned}
$$

The first inequality follows because $\operatorname{rep}_{A_{i-1}}(x) \geq 2^{j_{i}}$ for all $x \in A_{i}^{\prime}$, the second uses the induction hypothesis and the third follows by (10).

$$
\text { Let } M:=\operatorname{poly}\left(\frac{\epsilon_{t+1}}{n K}\right)(4 n)^{-(t-(i-1))}\left(\prod_{\ell=i-1}^{t} \epsilon_{\ell+1}\right)
$$

Since our graph has at least $2 M\left|A_{i-1}\right|^{2}$ edges and $\left|A_{i-1}\right|$ vertices, it has a connected component with at least $2 M\left|A_{i-1}\right|$ vertices and denote by $A_{i-1}^{\prime \prime}$ the set of vertices in it.

Choose an arbitrary element $a$ in $A_{i-1}^{\prime \prime}$. Partition $B_{i}^{\prime}$ into two sets $B_{i, 0}^{\prime}$ and $B_{i, 1}^{\prime}$ such that all elements in $B_{i, 0}^{\prime}$ have inner product 0 with $a$, and all elements in $B_{i, 1}^{\prime}$ have inner product 1 with $a$. Let $B_{i-1}^{\prime}$ be the larger of $B_{i, 0}^{\prime}, B_{i, 1}^{\prime}$, and note that $\left|B_{i-1}^{\prime}\right| \geq\left|B_{i}^{\prime}\right| / 2$. Recall that our assumption was that $D\left(A_{i}^{\prime}, B_{i}^{\prime}\right)=1$. Abusing notation, let $\left\langle A_{i}^{\prime}, B_{i}^{\prime}\right\rangle_{2}$ denote the value of $\left\langle a^{\prime}, b^{\prime}\right\rangle_{2}$ for
some $a^{\prime} \in A_{i}^{\prime}, B_{i}^{\prime}$ (the choice of $a^{\prime}, b^{\prime}$ does not matter because $D\left(A_{i}^{\prime}, B_{i}^{\prime}\right)=1$ ). Next we consider two cases - the case where $\left\langle A_{i}^{\prime}, B_{i}^{\prime}\right\rangle_{2}=0$, and the case where $\left\langle A_{i}^{\prime}, B_{i}^{\prime}\right\rangle_{2}=1$.
In the first case we have that for every $a, a^{\prime} \in A_{i-1}^{\prime \prime}$ which are neighbors in the graph, $a+a^{\prime} \in A_{i}^{\prime}$, and therefore $\left\langle a+a^{\prime}, b\right\rangle_{2}=0$ for every $b \in B_{i-1}^{\prime}$. This implies in turn that $\langle a, b\rangle_{2}=\left\langle a^{\prime}, b\right\rangle_{2}$ for all elements $a, a^{\prime} \in A_{i-1}^{\prime \prime}$ which are neighbors in the graph, $b \in B_{i-1}^{\prime}$. Since $A_{i-1}^{\prime \prime}$ induces a connected component, and due to our choice of $B_{i-1}^{\prime}$, this implies that $D\left(A_{i-1}^{\prime \prime}, B_{i-1}^{\prime}\right)=1$ so we set $A_{i-1}^{\prime}=A_{i-1}^{\prime \prime}$.

In the second case we have that $\left\langle a+a^{\prime}, b\right\rangle_{2}=1$ for every $a, a^{\prime} \in A_{i-1}^{\prime \prime}$ which are neighbors in the graph, $b \in$ $B_{i-1}^{\prime}$. In particular this implies that $\langle a, b\rangle_{2}=\left\langle a^{\prime}, b\right\rangle_{2}+1$ for every elements $a, a^{\prime} \in A_{i-1}^{\prime \prime}$ which are neighbors in the graph, $b \in B_{i-1}^{\prime}$. This means that $A_{i-1}^{\prime \prime}$ can be partitioned into two sets $A_{i-1,0}^{\prime}, A_{i-1,1}^{\prime}$, where the first one contains all elements in $A_{i-1}^{\prime \prime}$ that have inner product 0 with all elements in $B_{i-1}^{\prime}$, while the second set contains all elements in $A_{i-1}^{\prime}$ that have inner product 1 with all elements in $B_{i-1}^{\prime}$. We set $A_{i-1}^{\prime}$ to be the larger of these two sets and get $D\left(A_{i-1}^{\prime}, B_{i-1}^{\prime}\right)=1$ and $\left|A_{i-1}^{\prime}\right| \geq M\left|A_{i-1}\right|$.

Concluding, in both cases we obtained subsets $A_{i-1}^{\prime}, B_{i-1}^{\prime}$ of $A_{i-1}, B$ respectively, such that $D\left(A_{i-1}^{\prime}, B_{i-1}^{\prime}\right)=1$ and $A_{i-1}^{\prime}, B_{i-1}^{\prime}$ are not too small:

$$
\frac{\left|A_{i-1}^{\prime}\right|}{\left|A_{i-1}\right|} \geq\left(\frac{\epsilon_{t+1}}{n K}\right)^{O(1)}(4 n)^{-(t-(i-1))}\left(\prod_{\ell=i-1}^{t} \epsilon_{\ell+1}\right)
$$

and

$$
\begin{aligned}
\frac{\left|B_{i-1}^{\prime}\right|}{|B|} & \geq \frac{1}{2} \frac{\left|B_{i}^{\prime}\right|}{|B|} \geq \frac{1}{2} \operatorname{poly}\left(\frac{\epsilon_{t+1}}{n K}\right) 2^{-(t-i)} \\
& =\operatorname{poly}\left(\frac{\epsilon_{t+1}}{n K}\right) 2^{-(t-(i-1))}
\end{aligned}
$$

This concludes the proof of the inductive claim.

## III. From approximate duality to COMMUNICATION COMPLEXITY UPPER BOUNDS

In this section we prove our main theorem, Theorem I. 3 given Corollary I.11. The proof of the main technical lemma is deferred to Section II.

We start by repeating the necessary definitions. For a $\{0,1\}$-valued matrix $M$, let $\mathrm{CC}(M)$ denote the communication complexity of the boolean function associated with $M$. Let $\operatorname{rank}(M)$ and $\operatorname{rank}_{\mathbb{F}_{2}}(M)$ denote the rank of $M$ over the reals and over $\mathbb{F}_{2}$, respectively. We denote by $|M|$ the total number of entries in $M$, and by $\left|M_{0}\right|$ and $\left|M_{1}\right|$ the number of zero and non-zero entries of $M$, respectively. We say that $M$ is monochromatic if either $|M|=\left|M_{0}\right|$ or $|M|=\left|M_{1}\right|$. Finally, we define the discrepancy $\delta(M)$ of $M$ to be the ratio $\frac{\frac{\| M_{0}\left|-\left|M_{1}\right|\right|}{|M|} \text {. }}{|M|}$

Recall the statements of Theorem I. 3 and Corollary I.11.

Main Theorem I. 3 (restated). Assuming the PFR conjecture (Conjecture I.5), for every $\{0,1\}$-valued matrix M,

$$
\mathrm{CC}(M)=O(\operatorname{rank}(M) / \log \operatorname{rank}(M))
$$

Corollary 1.11 (restated). Suppose that $A, B \subseteq \mathbb{F}_{2}^{n}$ satisfy $D(A, B) \geq 2^{-\sqrt{n}}$. Then assuming the PFR conjecture, there exist subsets $A^{\prime}, B^{\prime}$ of $A, B$ respectively such that $D\left(A^{\prime}, B^{\prime}\right)=1$, and $\left|A^{\prime}\right| \geq 2^{-c n / \log n}|A|$, $\left|B^{\prime}\right| \geq 2^{-c n / \log n}|B|$ for some absolute constant $c$.

We first prove that the above corollary is equivalent to the following one:

Lemma III. 1 (Main technical lemma, equivalent matrix form). Let $M$ be a $\{0,1\}$-valued matrix with no identical rows or columns, of rank at most $r$ over $\mathbb{F}_{2}$, and of discrepancy at least $2^{-\sqrt{r}}$. Then assuming the PFR conjecture (Conjecture I.5), there exists a monochromatic submatrix $M^{\prime}$ of $M$ of size at least $2^{-c r / \log r}|M|$ for some absolute constant $c$.

Proof: We prove only the Corollary I. $11 \Rightarrow$ Lemma III. 1 implication. The proof of the converse implication is similar. Denote the number of rows and columns of $M$ by $k, \ell$ respectively. It is well known that the rank of $M$ over a field $\mathbb{F}$ equals $r$ if and only if $M$ can be written as the sum of $r$ rank one matrices over the field $\mathbb{F}$. Since $\operatorname{rank}_{\mathbb{F}_{2}}(M) \leq r$ this implies in turn that there exist subsets $A, B \subseteq \mathbb{F}_{2}^{r}, A=\left\{a_{1}, a_{2}, \ldots, a_{k}\right\}$, $B=\left\{b_{1}, b_{2}, \ldots, b_{\ell}\right\}$ such that $M_{i, j}=\left\langle a_{i}, b_{j}\right\rangle_{2}$ for all $1 \leq i \leq k, 1 \leq j \leq \ell$. Since $M$ has no identical rows or columns we know that $|A|=k,|B|=\ell$. Note that $D(A, B)=\delta(M) \geq 2^{-\sqrt{r}}$.

Corollary I. 11 now implies the existence of subsets $A^{\prime} \subseteq A, B^{\prime} \subseteq B,\left|A^{\prime}\right| \geq 2^{-c r / \log r}|A|,\left|B^{\prime}\right| \geq$ $2^{-c r /} \log r|B|$, such that $D\left(A^{\prime}, B^{\prime}\right)=1$. Let $M^{\prime}$ be the submatrix of $M$ whose rows and columns correspond to the indices in $A^{\prime}$ and $B^{\prime}$ respectively. The fact that $D\left(A^{\prime}, B^{\prime}\right)=1$ implies that $M_{i, j}=\left\langle a_{i}, b_{j}\right\rangle_{2} \equiv \mathrm{const}$ for all $a_{i} \in A^{\prime}, b_{j} \in B^{\prime}$. Therefore $M^{\prime}$ is a monochromatic submatrix of $M$ of which satisfies

$$
\begin{aligned}
\left|M^{\prime}\right| & =\left|A^{\prime}\right|\left|B^{\prime}\right| \geq 2^{-2 c r / \log r}|A||B| \\
& =2^{-2 c r / \log r}|M|
\end{aligned}
$$

as required.
In order to prove Theorem I. 3 we follow the highlevel approach of Nisan and Wigderson [9] which was explained in the previous section. They showed that in order to prove the log-rank conjecture it suffices to prove that every $\{0,1\}$-valued matrix of low rank has a large
monochromatic submatrix. We start with the following lemma.

Lemma III. 2 (Existence of large monochromatic submatrix assuming PFR). Assuming the PFR conjecture, every $\{0,1\}$-valued matrix $M$ with no identical rows or columns has a monochromatic submatrix of size at least $2^{-O(\operatorname{rank}(M) / \log \operatorname{rank}(M))}|M|$.

In order to prove the above lemma we use Lemma III.1, together with the following theorem from [9], which says that every $\{0,1\}$-valued matrix $M$ contains a submatrix of high discrepancy:

Theorem III. 3 (Existence of submatrix with high discrepancy [9]). Every $\{0,1\}$-valued matrix $M$ has a submatrix $M^{\prime}$ of size at least $(\operatorname{rank}(M))^{-3 / 2}|M|$ and with $\delta\left(M^{\prime}\right) \geq(\operatorname{rank}(M))^{-3 / 2}$.

Proof of Lemma III.2: Let $r=\operatorname{rank}(M)$. Theorem III. 3 implies the existence of a submatrix $M^{\prime}$ of $M$ with $\left|M^{\prime}\right| \geq(\operatorname{rank}(M))^{-3 / 2}|M|$, and $\delta\left(M^{\prime}\right) \geq r^{-3 / 2} \gg$ $2^{-\sqrt{r}}$. Note also that

$$
\operatorname{rank}_{\mathbb{F}_{2}}\left(M^{\prime}\right) \leq \operatorname{rank}\left(M^{\prime}\right) \leq \operatorname{rank}(M)=r
$$

Lemma III. 1 then implies the existence of a monochromatic submatrix $M^{\prime \prime}$ of $M^{\prime}$ of size at least $2^{-c r / \log r}\left|M^{\prime}\right|$ for some absolute constant $c$. So we have that $M^{\prime \prime}$ is a monochromatic submatrix of $M$ which satisfies

$$
\begin{aligned}
\left|M^{\prime \prime}\right| & \geq 2^{-c r / \log r}\left|M^{\prime}\right| \geq 2^{-c r / \log r} r^{-3 / 2}|M| \\
& =2^{-O(r / \log r)}|M|
\end{aligned}
$$

## Proof of Theorem I.3:

Let $M$ be a $\{0,1\}$-valued matrix. We will construct a deterministic protocol for $M$ with communication complexity $O(\operatorname{rank}(M) / \log \operatorname{rank}(M))$. We may assume w.l.o.g that $M$ has no repeated rows or columns, otherwise we can eliminate the repeated row or column and the protocol we construct for the "compressed" matrix (with no repeated rows/columns) will also be a protocol for $M$.

We follow the high level approach of the proof of Theorem 2 from [9]. We will show a protocol with $2^{O(r / \log r)}$ leaves. This will suffice since it is wellknown that a protocol with $t$ leaves has communication complexity at most $O(\log t)$ (cf. [23, Chapter 2, Lemma 2.8]).

Now we describe the protocol. Let $Q$ be the largest monochromatic submatrix of $M$. Then $Q$ induces a natural partition of $M$ into 4 submatrices $Q, R, S, T$ with $R$ sharing the rows of $Q$ and $S$ sharing the columns of $Q$.

$$
M=\left(\begin{array}{cc}
Q & R \\
S & T
\end{array}\right)
$$

Let $U_{1}$ be a subset of the rows of $(Q \mid R)$ whose restriction to the columns of $R$ span the rows of $R$. Similarly, let $U_{2}$ be a subset of the rows of $(S \mid T)$ whose restriction to the columns of $S$ span the rows of $S$. Note that if $Q$ is the all zeros matrix then the rows of $U_{1}$ are independent of the rows of $U_{2}$. Otherwise, if $Q$ is the all ones matrix then the rows of $U_{1}$ are independent of all the rows of $U_{2}$ except possibly for the vector in $U_{2}$ whose restriction to the columns of $S$ is the all ones vector (if such vector exists). Thus since $Q$ is monochromatic we have that $\operatorname{rank}(R)+\operatorname{rank}(S)=\left|U_{1}\right|+\left|U_{2}\right| \leq \operatorname{rank}(M)+1$.

If $\operatorname{rank}(R) \leq \operatorname{rank}(S)$ then the row player sends a bit saying if his input belongs to the rows of $Q$ or not. The players continue recursively with a protocol for the submatrix $(Q \mid R)$ or the submatrix $(S \mid T)$ according to the bit sent. If $\operatorname{rank}(R) \geq \operatorname{rank}(S)$ the roles of the row and column players are switched.

Suppose without loss of generality that $\operatorname{rank}(R) \leq$ $\operatorname{rank}(S)$. Then after sending one bit we continue with either the matrix $(Q \mid R)$ which is of rank at most $\operatorname{rank}(M) / 2$ or with the matrix $(S \mid T)$ which - thanks to Lemma III. 2 - is of size at most $(1-\delta)|M|$ for $\delta \geq 2^{-c r / \log r}$.

Let $L(m, r)$ denote the number of leaves in the protocol starting with a matrix of area at most $m$ and rank at most $r$. Then we get the following recurrence relation:

$$
L(m, r) \leq \begin{cases}L(m, r / 2)+L(m(1-\delta), r) & r>1 \\ 1 & r=1\end{cases}
$$

It remains to show that in the above recursion $L(m, r)=2^{O(r / \log r)}$. Applying the recurrence iteratively $1 / \delta$ times to the right-most summand we get

$$
\begin{aligned}
L(m, r) & \leq \delta^{-1} L(m, r / 2)+L\left(m(1-\delta)^{1 / \delta}, r\right) \\
& \leq 2^{c r / \log (r)} L(m, r / 2)+L(m / 2, r)
\end{aligned}
$$

Set $A(m, r):=2^{-2 c r / \log r} L(m, r)$. Then we have $A(m, r) \leq A(m, r / 2)+A(m / 2, r)$ which together with $A(1, r), A(m, 1) \leq 1$ imply $A(m, r) \leq\binom{\log m+\log r}{\log r}$ since we may apply the recursion iteratively at most $\log r$ times to the left term and $\log m$ times to the right term before we reach $A(1, r)$ or $A(m, 1)$. This in turn implies $A(m, r) \leq\binom{\log m+\log r}{\log r} \leq r^{O(\log r)}$ due to the fact that $r \leq m \leq 2^{2 r}$, since we may assume there are no identical rows or columns in the matrix $M$.

Concluding, we have $L(m, r) \leq 2^{2 c r / \log r+O\left(\log ^{2} r\right)}$, which implies in turn $\mathrm{CC}(M)=O(r / \log r)$ as claimed.
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