Larger Corner-Free Sets from Combinatorial Degenerations

Christandl, M.; Fawzi, O.; Ta, H.; Zuiddam, J.

DOI
10.4230/LIPIcs.ITCS.2022.48

Publication date
2022

Document Version
Final published version

Published in
13th Innovations in Theoretical Computer Science Conference

License
CC BY

Citation for published version (APA):
Christandl, M., Fawzi, O., Ta, H., & Zuiddam, J. (2022). Larger Corner-Free Sets from Combinatorial Degenerations. In M. Braverman (Ed.), 13th Innovations in Theoretical Computer Science Conference: ITCS 2022, January 31-February 3, 2022, Berkeley, CA, USA Article 48 (Leibniz International Proceedings in Informatics; Vol. 215). Schloss Dagstuhl - Leibniz-Zentrum für Informatik. https://doi.org/10.4230/LIPIcs.ITCS.2022.48

General rights
It is not permitted to download or to forward/distribute the text or part of it without the consent of the author(s) and/or copyright holder(s), other than for strictly personal, individual use, unless the work is under an open content license (like Creative Commons).

Disclaimer/Complaints regulations
If you believe that digital publication of certain material infringes any of your rights or (privacy) interests, please let the Library know, stating your reasons. In case of a legitimate complaint, the Library will make the material inaccessible and/or remove it from the website. Please Ask the Library: https://uba.uva.nl/en/contact, or a letter to: Library of the University of Amsterdam, Secretariat, Singel 425, 1012 WP Amsterdam, The Netherlands. You will be contacted as soon as possible.

UvA-DARE is a service provided by the library of the University of Amsterdam (https://dare.uva.nl)
Larger Corner-Free Sets from Combinatorial Degenerations

Matthias Christandl
Department of Mathematical Sciences, University of Copenhagen, Denmark

Omar Fawzi
Univ. Lyon, ENS Lyon, UCBL, CNRS, Inria, LIP, France

Hoang Ta
Univ. Lyon, ENS Lyon, UCBL, CNRS, Inria, LIP, France

Jeroen Zuiddam
Korteweg-de Vries Institute for Mathematics, University of Amsterdam, The Netherlands

Abstract
There is a large and important collection of Ramsey-type combinatorial problems, closely related to central problems in complexity theory, that can be formulated in terms of the asymptotic growth of the size of the maximum independent sets in powers of a fixed small hypergraph, also called the Shannon capacity. An important instance of this is the corner problem studied in the context of multiparty communication complexity in the Number On the Forehead (NOF) model. Versions of this problem and the NOF connection have seen much interest (and progress) in recent works of Linial, Pitassi and Shraibman (ITCS 2019) and Linial and Shraibman (CCC 2021).

We introduce and study a general algebraic method for lower bounding the Shannon capacity of directed hypergraphs via combinatorial degenerations, a combinatorial kind of “approximation” of subgraphs that originates from the study of matrix multiplication in algebraic complexity theory (and which play an important role there) but which we use in a novel way.

Using the combinatorial degeneration method, we make progress on the corner problem by explicitly constructing a corner-free subset in \( \mathbb{F}_2^n \times \mathbb{F}_2^n \) of size \( \Omega(3.39^{n/poly(n)}) \), which improves the previous lower bound \( \Omega(2.82^n) \) of Linial, Pitassi and Shraibman (ITCS 2019) and which gets us closer to the best upper bound \( 4^n - o(n) \). Our new construction of corner-free sets implies an improved NOF protocol for the Eval problem. In the Eval problem over a group \( G \), three players need to determine whether their inputs \( x_1, x_2, x_3 \in G \) sum to zero. We find that the NOF communication complexity of the Eval problem over \( \mathbb{F}_2^n \) is at most \( 0.24n + O(\log n) \), which improves the previous upper bound \( 0.5n + O(\log n) \).

2012 ACM Subject Classification Theory of computation → Communication complexity; Theory of computation → Algebraic complexity theory; Mathematics of computing → Discrete mathematics

Keywords and phrases Corner-free sets, communication complexity, number on the forehead, combinatorial degeneration, hypergraphs, Shannon capacity, eval problem

Digital Object Identifier 10.4230/LIPIcs.ITCS.2022.48

Related Version Full Version: http://arxiv.org/abs/2111.08262 [17]

Funding Matthias Christandl: European Research Council (ERC Grant Agreement No. 818761), the VILLUM FONDEN via the QMATH Centre of Excellence (Grant No. 10059) and the QuantERA ERA-NET Cofund in Quantum Technologies implemented within the European Union’s Horizon 2020 Programme (QuantAlgo project) via the Innovation Fund Denmark.

Omar Fawzi: European Research Council (ERC Grant Agreement No. 851716).

Hoang Ta: European Research Council (ERC Grant Agreement No. 851716), LABEX MILYON (ANR-10-LABX-0070) of Université de Lyon, within the program “Investissements d’Avenir” (ANR-11-IDEX-0007) operated by the French National Research Agency (ANR).

© Matthias Christandl, Omar Fawzi, Hoang Ta, and Jeroen Zuiddam; licensed under Creative Commons License CC-BY 4.0

13th Innovations in Theoretical Computer Science Conference (ITCS 2022).
Editor: Mark Braverman; Article No. 48; pp. 48:1–48:20
Leibniz International Proceedings in Informatics
Schloss Dagstuhl – Leibniz-Zentrum für Informatik, Dagstuhl Publishing, Germany
Larger Corner-Free Sets from Combinatorial Degenerations

1 Introduction

This paper is about constructing special combinatorial objects, namely “corner-free sets” in $F_p^n \times F_p^n$, motivated (besides their inherent interest) by central problems in communication complexity, specifically in the study of the number on the forehead (NOF) model of communication introduced by Chandra, Furst and Lipton [15]. There has been much interest in (and progress on) the corner problem, variations of the problem, and connections to NOF communication, in particular in the recent works of Shraibman [45], Linial, Pitassi and Shraibman [34], Viola [50], Alon and Shraibman [5], and Linial and Shraibman [35, 36]. In the recent work of Linial and Shraibman [35] a construction of large corner-free sets in $[N] \times [N]$ was obtained in an elegant manner by designing efficient NOF communication protocols for a specific communication problem (much like the upcoming Eval problem). We take a different, algebraic approach to the corner problem, and make progress on the corner problem over $F_p^n \times F_p^n$ by introducing in this area a new algebraic method via combinatorial degeneration.

NOF communication complexity

The NOF model is very rich in terms of connections to Ramsey theory and additive combinatorics [9, 45, 34, 35, 36], as well as applications to boolean models of computation such as branching programs and boolean circuits [15, 10]. The goal in the NOF model is for $k$ players to compute a fixed given function $F : X_1 \times \cdots \times X_k \rightarrow \{0, 1\}$ on inputs $(x_1, \ldots, x_k) \in X_1 \times \cdots \times X_k$ where player $i$ has access to input $x_i$ for all $j \neq i$ but no access to input $x_j$. For $k = 2$, this model coincides with the standard two-party communication model of Yao [51], but when $k \geq 3$, the shared information between the players makes this model surprisingly powerful [29, 6, 1, 16], and fundamental problems remain open. For instance, a sufficiently strong lower bound for an explicit function $F$ for $k \geq \text{polylog}(n)$ players with $n = \log |X|$ would imply a breakthrough result in complexity theory, namely a lower bound on the complexity class $\text{ACC}^0$.

NOF complexity of the Eval problem

A central open problem in the theory of NOF communication is to construct an explicit function for which randomized protocols are significantly more efficient than deterministic ones [8]. A well-studied candidate for this separation (for $k = 3$) is the function $\text{Eval}_{F_2}$, which is defined by $\text{Eval}_{F_2}(x_1, x_2, x_3) = 1$ if and only if $x_1 + x_2 + x_3 = 0$, where the additions are all in $F_2$. Thus the Eval problem naturally generalizes the equality problem for $k = 2$. It is known that in the randomized setting, the standard protocol for the two-party equality problem that uses $O(1)$ bits of communication works in the same way for three parties for the Eval problem. However, in the deterministic setting, the communication complexity $D_3(\text{Eval}_{F_2})$ remains wide open: the best known lower bound $\Omega(\log \log n)$ follows from the work of Lacey and McClain [33] and, before this work, the best upper bound was $0.5n + O(\log n)$ [1].

Corner problem in combinatorics, and connection to the Eval problem

Chandra, Furst and Lipton [15] found that the deterministic communication complexity of many problems in the NOF model can be recast as Ramsey theory problems. In particular, and this leads to the problem of interest in this paper, the deterministic communication complexity of $\text{Eval}_{F_2}$ can be characterized in terms of corner-free subsets of $F_2^2 \times F_2^2$, as follows. We call any triple of elements $(x, y), (x + \lambda, y), (x, \lambda + y)$ for $x, y, \lambda \in F_2$ a corner. A subset $S \subseteq F_2^2 \times F_2^2$ is called corner-free if it does not contain any nontrivial corners (where nontrivial
means that \(\lambda \neq 0\). Denoting by \(r_c(F_2^n)\) the size of the largest corner-free set in \(F_2^n \times F_2^n\), the communication complexity of \(\text{Eval}_{F_2}^n\) equals \(\log(4^n/r_c(F_2^n))\) up to a \(\mathcal{O}(\log n)\) additive term, which provides the close connection between the \(\text{Eval}\) problem in NOF communication and the corner problem in combinatorics. In particular, large corner-free sets in \(F_2^n \times F_2^n\) correspond to efficient protocols for \(\text{Eval}_{F_2}^n\).

**General paradigm: Shannon capacity of hypergraphs**

The point of view we will take (and the general setting in which the methods we introduce will apply) is to regard the corner problem as a Shannon capacity problem of directed hypergraphs. Namely, the size \(r_c(F_2^n)\) of the largest corner-free set in \(F_2^n \times F_2^n\) can be characterized as the independence number of a (naturally defined) directed 3-uniform hypergraph with \(4^n\) vertices.\(^{1}\)

This hypergraph has a recursive form: it is obtained by taking the \(r\)-th power of a fixed (directed) hypergraph \(H_{\text{cor},F_2}\) on 4 vertices. (We discuss this in more detail in Section 2.) The asymptotic growth of \(r_c(F_2^n)\) as \(n \to \infty\) is characterized by the Shannon capacity \(\Theta(H_{\text{cor},F_2})\) of the corner hypergraph \(H_{\text{cor},F_2}^{\otimes r}\).\(^{2}\) That is, we have \(r_c(F_2^n) = \Theta(H_{\text{cor},F_2}^{n-o(1)})\). In this way, proving the strict upper bound \(\Theta(H_{\text{cor},F_2}) < 4\) is equivalent to proving a linear lower bound on the communication complexity of \(\text{Eval}_{F_2}^n\). Many other Ramsey type problems can be expressed as the Shannon capacity of some fixed hypergraph, such as the Cap Set problem that saw a recent breakthrough by Ellenberg and Gijswijt [25] following Croot, Lev and Pach [22], and the Uniquely Solvable Puzzle (USP) problems that were put forward in the “group-theoretic approach” to the matrix multiplication problem [20, 4].

1.1 Is the complexity of the \(\text{Eval}\) problem maximal?

Let us discuss the open problem that motivates our work, and that is central in NOF communication complexity and combinatorics (through the aforementioned connections). This problem asks whether or not the complexity of the \(\text{Eval}\) problem is “maximal”, or in other words, whether or not there are corner-free sets in \(F_2^n \times F_2^n\) that have “sub-maximal” size:

- **Problem 1.** Are the following three statements (which we know are equivalent\(^3\)) true?
  - \(D_3(\text{Eval}_{F_2}^n) = \Omega(n)\)
  - \(r_c(F_2^n) \leq O(c^n)\) for some \(c < 4\)
  - \(\Theta(H_{\text{cor},F_2}) < 4\).

Here the best capacity lower bound before our work was \(\Theta(H_{\text{cor},F_2}) \geq \sqrt{8}\) by Linial, Pitassi and Shraibman [34, Cor. 24 in the ITCS version], obtained by explicit construction of an independent set in the second power of the relevant hypergraph, which in turn leads to the bounds \(D_3(\text{Eval}_{F_2}^n) \leq 0.5n + \mathcal{O}(\log n)\) and \(r_c(F_2^n) \geq \sqrt{8^n}\).

In the above we may naturally generalize \(F_2^n\) to \(F_p^n\) or even to \(G^n\), where \(G\) is an arbitrary abelian group, so that Problem 1 is a special case of the more general problem:

\(^{1}\) As usual an independent set of a hypergraph is a subset \(S\) of vertices such that no hyperedge has all its vertices in \(S\).

\(^{2}\) In the setting of directed graphs, also the term Sperner capacity (typically applied to the complement graph) [27, 26] is used for what we call Shannon capacity.

\(^{3}\) The equivalence among the three formulations is standard and follows from Lemma 17, Proposition 16 and Lemma 14 further on in the paper. We will mainly use the formulation in terms of Shannon capacity (see Definition 12 for a precise definition).
Problem 2. Are the following three statements (which we know are equivalent) true?

- \( D_3(\text{Eval}_{G^n}) = \Omega(n) \)
- \( r_\triangle(G^n) \leq O(c^n) \) for some \( c < |G|^2 \)
- \( \Theta(H_{\text{cor},G}) < |G|^2 \).

Our goal in this paper, motivated by the connections as remarked earlier, is to make progress on above problems via new algebraic methods.

1.2 Lower bounds for the corner problem (and other problems) from combinatorial degeneration

Our main result is progress on Problem 2 by proving new lower bounds for the corner problem over the groups \( \mathbb{F}_2 \) and \( \mathbb{F}_3 \), which we arrive at via a new method to lower bound the Shannon capacity of directed hypergraphs. Equivalently, in the language of communication complexity, we obtain improved protocols for the Eval problem.

The lower bound of Linial, Pitassi and Shraibman [34] for the corner problem was obtained by explicit construction of an independent set (i.e. a set that does not contain edges) in the second power of a hypergraph, which is the natural approach for such lower bounds. We improve on this bound by observing that it is actually sufficient to construct a set which does not contain “cycles”. For graphs, the notion of cycle is clear but for hypergraphs there are many possible definitions, and we initiate a careful study of this (and believe that this will be a worthwhile avenue for further study independently). Here, to get new bounds we use the notion of combinatorial degeneration to model such a “cycle”. We will say more about this in a moment.

Using the combinatorial degeneration method on the corner hypergraphs that characterize the corner problem we find new bounds for Problem 2 for the groups \( \mathbb{F}_2^n \) and \( \mathbb{F}_3^n \). These are as follows (in the three equivalent forms):

**Theorem 3** (Thm. 26). For the corner and Eval problem over \( \mathbb{F}_2^n \) we have:

- \( D_3(\text{Eval}_{\mathbb{F}_2^n}) \leq 0.24n + O(\log n) \)
- \( r_\triangle(\mathbb{F}_2^n) \geq \frac{\sqrt{3}n}{\text{poly}(n)} \)
- \( \Theta(H_{\text{cor},\mathbb{F}_2^n}) \geq \sqrt{3} \)

**Theorem 4** (Thm. 25). For the corner and Eval problem over \( \mathbb{F}_3^n \) we have:

- \( D_3(\text{Eval}_{\mathbb{F}_3^n}) \leq 0.37n + O(\log n) \)
- \( r_\triangle(\mathbb{F}_3^n) \geq \frac{7^n}{\text{poly}(n)} \)
- \( \Theta(H_{\text{cor},\mathbb{F}_3^n}) \geq 7 \).

Let us discuss on a high level the history and ideas behind the combinatorial degeneration method. Combinatorial degeneration is an existing concept from algebraic complexity theory. It was (in a slightly different form) introduced and studied by Strassen in [46, Section 6].

Degeneration of tensors is a powerful approximation notion in the theory of tensors. Combinatorial degeneration is the “combinatorial” or “torus” version of this kind of approximation. Combinatorial degeneration was introduced by Bürgisser, Clausen and Shokrollahi [14, Definition 15.29] based on the notion of M-degeneration for tensors defined and studied by Strassen in [46].
be reduced (in an appropriate algebraic manner) to matrix multiplication [46, Theorem 6.6].\(^5\)

Strassen then used this result to prove his Laser method [46, Section 7], vastly generalizing the method that Coppersmith and Winograd had introduced in their construction of matrix multiplication algorithms [21].\(^6\)

Combinatorial degeneration was used more broadly to construct large induced matchings in the setting of important combinatorial problems, namely the Sunflower problem by Alon, Shpilka and Umans [4, Lemma 3.9] and the Cap Set problem by Kleinberg, Sawin and Speyer [31]. These results are often referred to as the “multicolored” versions of the problem at hand, as opposed to the “single color” version. These ideas were developed further in the context of matrix multiplication barriers by Alman and Williams [2, Lemma 6] and in the study of tensors by Christandl, Vrana and Zuiddam [18, Theorem 4.11].

Crucially, all of the above applications use combinatorial degeneration to construct induced matchings in \((k\text{-uniform } k\text{-partite})\) hypergraphs. However, we use combinatorial degeneration in a novel manner to construct independent sets in hypergraphs instead of induced matchings. In this context an independent set should be thought of as a symmetric induced matching. Constructing large independent sets is a much harder task than constructing large induced matchings, as witnessed by the fact that the “multicolored” cap set problem is solved [31] while its “single color” version is not. Similarly, for the corner problem, as we will discuss in Section 1.4, the asymptotic growth of the largest induced matching can be shown to be maximal, whereas the main question of study of this paper is whether the same holds for the largest independent set. We expect our new way of using combinatorial degeneration to be useful in the study of other problems besides the corner problem as well, and thus think it is of independent interest.

On a more technical level, combinatorial degeneration is a notion that compares sets of \(k\)-tuples by means of algebraic conditions. Our “universe” is \(I_1 \times \cdots \times I_k\) where \(I_1, \ldots, I_k\) are finite sets. Then for sets \(\Phi \subseteq \Psi \subseteq I_1 \times \cdots \times I_k\) we say that \(\Phi\) is a combinatorial degeneration of \(\Psi\), and write \(\Psi \succeq \Phi\), if there are maps \(u_i : I_i \to \mathbb{Z}\) such that for every \(x = (x_1, \ldots, x_k) \in I_1 \times \cdots \times I_k\), if \(x \in \Psi \setminus \Phi\), then \(\sum_{i=1}^k u_i(x_i) > 0\), and if \(x \in \Phi\), then \(\sum_{i=1}^k u_i(x_i) = 0\). Thus the maps \(u_i\) together are able to distinguish between the elements in the set \(\Phi\) (which may be thought of as our “goal” set, i.e. a set with good properties) and the elements in the difference \(\Psi \setminus \Phi\). As a quick example of a combinatorial degeneration, let

\[
\begin{align*}
\Phi &= \{(0, 0, 0), (1, 1, 0), (1, 0, 1)\}, \\
\Psi &= \{(0, 0, 0), (1, 1, 0), (1, 0, 1), (0, 1, 1)\}.
\end{align*}
\]

Then we find a combinatorial degeneration \(\Psi \succeq \Phi\) by defining the maps \(u_i : \{0, 1\} \to \mathbb{Z}\) simply by setting \(u_1(0) = u_2(0) = u_3(0) = 0\), and \(u_1(1) = -1, u_2(1) = u_3(1) = 1\).

We apply the idea of combinatorial degeneration in the following fashion to get Shannon capacity lower bounds:

**Theorem 5** (Combinatorial degeneration method, Theorem 23). Let \(H = (V, E)\) be a directed \(k\text{-uniform} \ k\text{-partite}\) hypergraph. Let \(S \subseteq V\) be a subset of vertices. Define the sets

\[
\Psi = E \cup \{(v, \ldots, v) : v \in V\}
\]

\(^5\) Strassen’s result is asymptotically optimal. Strassen’s proof resembles Behrend’s construction of arithmetic-progression-free sets. Also note that this is precisely the opposite of the problem of reducing matrix multiplication to as few independent scalar multiplications as possible. The latter corresponds precisely to the arithmetic complexity of matrix multiplication.

\(^6\) The book [14, Definition 15.29 and Lemma 15.31] gives a different proof of the Laser method which relies even more strongly on combinatorial degeneration.
Suppose that $\Psi \supseteq \Phi$ is a combinatorial degeneration. Then we get the Shannon capacity lower bound $\Theta(H) \geq |S|$.

In other words, whereas $S$ in the statement of Theorem 5 may not be an independent set, we can via the algebraic conditions of combinatorial degeneration construct an independent set in the $n$th power of the hypergraph of size approaching $|S^*|$. Namely, the algebraic conditions allow us to select such an independent set using a natural type analysis of the labels given by the maps $u_i$. Thus we may think of a set $S$ as above as an approximative independent set, which asymptotically we can turn into an actual independent set by means of Theorem 5.

We note that, whereas it is relatively simple to verify for a given set $S$ that $\Psi \supseteq \Phi$ holds (with the notation of Theorem 5) via linear programming, it is seems much harder to find a large set $S \subseteq V$ for which $\Psi \supseteq \Phi$, given $H$. We obtain our best lower bounds via an integer linear programming approach. The resulting combinatorial degenerations that we find are explicit and checkable by hand.

We have yet to develop structural understanding of how the above combinatorial degenerations that exhibit the new capacity lower bounds arise (and we feel that deeper understanding of this may lead to more progress or even solve the corner problem), and leave the investigation of further generalizations and improvements to future work. As a partial remedy to our limited understanding, we introduce the acyclic method as a tool to construct combinatorial degenerations. While the acyclic method does not recover the bounds of Theorem 3 and Theorem 4, it has the merits of being transparent and simple to apply. The acyclic method involves another notion of a set without “cycles”, which implies a combinatorial degeneration, but whose conditions are simpler to check.

### 1.3 Lower bounds for the corner problem from the probabilistic method

We employ the probabilistic method to find the following very general bound for the corner problem over arbitrary abelian groups.

**Theorem 6 (Prop. 19).** For the corner and Eval problem over an arbitrary abelian group $G$ we have

- $D_3(\text{Eval}_{G^n}) \leq \frac{\log |G|}{2} n + O(\log n)$
- $r_{\subseteq}(G^n) \geq \frac{|G|^{3n/2}}{\text{poly}(n)}$
- $\Theta(H_{\text{cor}.G}) \geq |G|^{3/2}$

This general bound applied to the special cases $G = F_2$ and $G = F_3$ does not quite match the bounds in Theorem 3 and Theorem 4, respectively. However, applied to the special case $G = F_2$ we do recover the lower bound $\sqrt{8}$ of [34, Cor. 24 in the ITCS version].

Using the same techniques we gain insight about the high-dimensional version of the corner problem and Eval problem and what happens when the number of players grows. For an arbitrary abelian group $G$, a $k$-dimensional corner over $G$ is naturally defined as a set of $(k+1)$ points in $(G)^\times k$ of the form \( \{(x_1, x_2, \ldots, x_k), (\lambda + x_1, x_2, \ldots, x_k), \ldots, (x_1, x_2, \ldots, \lambda + x_k)\} \) where $x_i, \lambda \in G^n$. A subset $S \subseteq G^{xk}$ is called corner-free if it does not contain any nontrivial corners (where nontrivial again means $\lambda \neq 0$). We denote by $r_{k,\subseteq}(G)$ the size of the largest $(k$-dimensional) corner-free set. Just like the $k = 2$ case corner-free sets correspond to independent sets in a naturally defined $(k+1)$-uniform directed hypergraph $H_{k,\text{cor}.G}$. With
the probabilistic method (extending Theorem 6), we find that when the \( k \) goes to infinity, the capacity of \( H_{k,\text{cor},G} \) becomes essentially maximal. As a consequence if \( k \) grows with \( n \) (e.g., \( k = \log n \)) we find that the NOF complexity of the corresponding \( k \)-player Eval problem becomes sub-linear.

\[ \Theta(H_{k,\text{cor},G}) \geq |G|^{k-1/k} \]

Thus we learn that to prove a linear lower bound on \( D_k(\text{Eval}_{G^n}) \) for any given \( G \) (say for \( G = \mathbb{F}_p \)) it is important to keep \( k \) constant.

1.4 Limitations of tensor methods for proving upper bounds for the corner problem

Our second result is a strong limitation of current tensor methods to effectively upper bound the Shannon capacity of hypergraphs. This limitation is caused by induced matchings and applies to various combinatorial problems including the corner problem. We use a method of Strassen to show that these limitations are indeed very strong for the corner problem.

In order to elaborate on these results let us first give an overview of upper bound methods. The general question of upper bounds on the Shannon capacity of hypergraphs is particularly well-studied in the special setting of undirected graphs, from which the name “Shannon capacity” comes: it in fact corresponds to the zero-error capacity of a channel [42]. Even for undirected graphs, it is not clear how to compute the Shannon capacity in general, but some methods were developed to give upper bounds. The difficulty is to find a good upper bound on the largest independent set that behaves well under the product \( \boxtimes \). For undirected graphs, the best known methods are the Lovász theta function [37] and the Haemers bound which is based on the matrix rank [30]. For hypergraphs, we only know of algebraic methods that are based on various notions of tensor rank, and in particular the slice rank [49] (which was used and studied extensively in combinatorics, in the context of cap sets [48, 31], sunflowers [41] and right-corners [40]), and similar notions like the analytic rank [28, 38, 13], the geometric rank [32], and the G-stable rank [23]. Even though the slice rank is not multiplicative under \( \boxtimes \) it is possible to give good upper bounds on the asymptotic slice rank via an asymptotic analysis [49], which is closely related to the Strassen support functionals [47] or the more recent quantum functionals [18].

Most of the rank-based bounds actually give upper bounds on the size of induced matchings and not only on the size of independent sets. It is simple and instructive to see this argument in the setting of undirected graphs. For a given graph \( H = (V, E) \), let \( A \) be the adjacency matrix in which we set all the diagonal coefficients to 1. Then for any independent set \( I \subseteq V \), the submatrix \( (A_{i,j})_{i,j \in I} \) of \( A \) is the identity matrix and as a result \( |I| \leq \text{rank}(A) \). As such, the matrix rank is an upper bound on the asymptotic maximum induced matching. Tensor rank methods such as the subrank, slice rank, analytic rank, geometric rank and G-stable rank also provide upper bounds on the asymptotic maximum induced matching.
Table 1 Independence number and induced matching number for small powers of the cap set hypergraph $H_{\cap}$ and corner hypergraph $H_{\cap,F_2}$. Interestingly, the independence number and induced matching number of powers of the cap set hypergraph are exactly equal for the powers $n = 1, 2, 3$. For the corner hypergraph we see that they are different already for the second and third power.

| $n$ | Independence number $H_{\cap,F_2}^n$ | Induced matching number $H_{\cap,F_2}^n$ | Independence number $H_{\cap,F_2}^n$ | Induced matching number $H_{\cap,F_2}^n$ |
|-----|-------------------------------------|------------------------------------------|-------------------------------------|------------------------------------------|
| 1   | 2                                   | 2                                        | 1                                   | 2                                        |
| 2   | 4                                   | 4                                        | 2                                   | 8                                        |
| 3   | 9                                   | 9                                        | 3                                   | 24                                       |

Using a result of Strassen [47], we show that there is an induced matching of the $n$-th power of $H_{\cap,F_2}$ of size $4^{n-o(1)}$. This establishes a barrier on many existing tensor methods (such as slice rank, subrank, analytic rank, etc.) to make progress on Problem 1. In fact, this result holds more generally for any abelian group $G$:

Theorem 8 (Cor. 37). For any abelian group $G$, the hypergraph $H_{\cap,G}^n$ has an induced matching of size $|G|^{2n-o(n)}$. In other words, for any $n \geq 1$, there exist lists $I_1, I_2, I_3 \subseteq G^n \times G^n$ of size $s(n) = |G|^{2n-o(n)}$ such that the following holds. For any $\alpha, \beta, \gamma \in \{1, \ldots, s(n)\}$ we have $(I_1(\alpha), I_2(\beta), I_3(\gamma))$ forms a corner $\iff \alpha = \beta = \gamma$.

We prove this result by establishing in Theorem 36 that the adjacency tensor of the hypergraph $H_{\cap,G}$ is tight (see Definition 34). Strassen showed in [47] that for tight sets, the asymptotic induced matching number is characterized by the support functionals. By computing the support functionals for the relevant tensors, we establish the claimed result in Corollary 37. Note that if we could ensure that $I_1 = I_2 = I_3$, this would solve Problem 1. We computed the maximum independent set and maximum induced matching for $H_{\cap,F_2}^n$ for small powers $n = 1, 2, 3$ (see Table 1) and we found that the maximum independent set is strictly smaller than the maximum induced matching for $n = 2$ and $n = 3$. This motivates the search for methods that go beyond the maximum induced matching barrier. For comparison, we also give the analogous numbers for the cap set hypergraph $H_{\cap}$ (which is an undirected hypergraph), where, interestingly, the maximum independent set and the maximum induced matching are equal.

2 Lower bounds from the combinatorial degeneration method

In this section we discuss three methods to prove lower bounds on the Shannon capacity of directed 3-uniform hypergraphs: the probabilistic method, the combinatorial degeneration method and the acyclic set method. We apply these methods to the corner problem — the problem of constructing large corner-free sets — which as a consequence gives new NOF communication protocols for the Eval problem. We begin by discussing the corner problem and its relation to NOF communication complexity.
2.1 Corner problem, cap set problem and number on the forehead communication

Hypergraphs

We recall the definition of directed $k$-uniform hypergraphs and basic properties of Shannon capacity on directed $k$-uniform hypergraphs.

Definition 9. A directed $k$-uniform hypergraph $H$ is a pair $H = (V, E)$ where $V$ is a finite set of elements called vertices, and $E$ is a set of $k$-tuples of elements of $V$ which are called hyperedges or edges. If the set of edges $E$ is invariant under permuting the $k$ coefficients of its elements, then we may also think of $H$ as an undirected $k$-uniform hypergraph.

Let $H = (V, E)$ be a directed $k$-uniform hypergraph with $n$ vertices. The adjacency tensor $A$ of $H$ is defined as

$$A_{i_1,\ldots,i_k} = \begin{cases} 1 & \text{if } i_1 = i_2 = \cdots = i_k \text{ or } (i_1,\ldots,i_k) \in E, \\ 0 & \text{otherwise}. \end{cases}$$

Definition 10. The strong product of a pair of directed $k$-uniform hypergraphs $G = (V_G, E_G)$ and $H = (V_H, E_H)$ is denoted $G \boxtimes H$ and defined as follows. It is a directed $k$-uniform hypergraph with vertex set $V_G \times V_H$ and the following edge set: Any $k$ vertices $(g_1,h_1),\ldots,(g_k,h_k) \in V_G \times V_H$ form an edge $((g_1,h_1),\ldots,(g_k,h_k))$ if one of the following three conditions holds:
1. $g_1 = \cdots = g_k$ and $(h_1,\ldots,h_k) \in E_H$
2. $(g_1,\ldots,g_k) \in E_G$ and $h_1 = \cdots = h_k$
3. $(g_1,\ldots,g_k) \in E_G$ and $(h_1,\ldots,h_k) \in E_H$

Definition 11. An independent set in a directed $k$-uniform hypergraph $H = (V, E)$ is a subset $S$ of the vertices $V$ that induces no edges, meaning for every $(e_1,\ldots,e_k) \in E$ there is an $i \in [k]$ such that $e_i \notin S$. The independence number of $H$, denoted by $\alpha(H)$, is the maximal size of an independent set in $H$.

If $S$ and $T$ are independent sets in two directed $k$-uniform hypergraphs $G$ and $H$, respectively, then $S \times T$ is an independent set in the strong product $G \boxtimes H$. Therefore, we have the supermultiplicativity property $\alpha(G)\alpha(H) \leq \alpha(G \boxtimes H)$. For any directed $k$-uniform hypergraph $H$, let $H^{\boxtimes n}$ denote the $n$-fold product of $H$ with itself.

Definition 12. The Shannon capacity of a directed $k$-uniform hypergraph $H$ is defined as

$$\Theta(H) := \lim_{n \to \infty} (\alpha(H^{\boxtimes n}))^{1/n}.$$  

By Fekete’s lemma we can write $\Theta(H) = \sup_m (\alpha(H^{\boxtimes n}))^{1/n}$. The following proposition can be deduced directly from the definition of Shannon capacity.

Proposition 13. Suppose $H$ is a directed $k$-uniform hypergraph with $m$ vertices and there is an independent set of size $s$ in $H^{\boxtimes n}$. Then $s^{\frac{1}{n}} \leq \Theta(H) \leq m$.

Corner problem

Let $(G, +)$ be a finite Abelian group. A corner in $G \times G$ is a three-element set of the form $\{ (x,y), (x + \lambda, y), (x, y + \lambda) \}$ for some $x, y, \lambda \in G$ and $\lambda \neq 0$. The element $(x,y)$ is called the center of this corner. Let $r_\Delta(G)$ be the size of the largest subset $S \subseteq G \times G$ such that no three elements in $S$ form a corner. The corner problem asks to determine $r_\Delta(G)$ given $G$. 

ITCS 2022
Trivially, we have the upper bound \( r_c(G) \leq |G|^2 \). The best-known general upper bound on \( r_c(G) \) comes from [43, 44], and reads
\[
r_c(G) \leq \frac{|G|^2}{(\log \log |G|)^3},
\]
where \( 0 < c < \frac{1}{10} \) is an absolute constant. In the finite field setting, in [33] the following better upper bound for \( r_c(G) \) with \( G = \mathbb{F}_2^n \) was obtained:
\[
r_c(\mathbb{F}_2^n) \leq O\left(\frac{|G|^2 \log \log |G|}{\log |G|}\right).
\]

We may phrase the corner problem as a hypergraph independence problem. We define \( H_{\text{cor}, G} = (V, E) \) to be the directed 3-uniform hypergraph with \( V = \{(g_1, g_2) : g_1, g_2 \in G\} \) and \( E = \{(g_1, g_2), (g_1 + \lambda, g_2), (g_1, g_2 + \lambda) : g_1, g_2, \lambda \in G, \lambda \neq 0\} \). Then by construction:

**Lemma 14.** \( r_c(G^n) = \alpha(H_{\text{cor}, G}^n) \).

As a consequence, \( r_c(G^n) = \Theta(H_{\text{cor}, G}^n)^{n-o(n)} \).

**Example 15.** Let \( G \) correspond to addition in \( \mathbb{F}_2 \). Then \( H_{\text{cor}, G} = (V, E) \) with
\[
E = \{(0,0), (1,0), (0,1), (0,1), (1,0), (0,0), (1,1), (1,1), (0,1), (1,0)\}.
\]
Under the labeling \((0,0) = 0, (0,1) = 1, (1,0) = 2 \) and \((1,1) = 3 \) we will think of \( H_{\text{cor}, \mathbb{F}_2} \) as the hypergraph \( H_{\text{cor}, \mathbb{F}_2} = (V, E) \) with \( V = \{0, 1, 2, 3\} \) and \( E = \{(0,2), (1,3), (2,0), (3,1)\} \).

Closely related to \( r_c(G) \) is the minimum number of colors needed to color \( G \times G \) so that no corner is monochromatic, which we denote by \( c_c(G) \). Then:

**Proposition 16 ([15, 34]).** Let \((G, +)\) be a finite Abelian group. There is a constant \( c \), such that for every \( n \in \mathbb{N} \),
\[
\frac{|G|^{2n}}{r_c(G^n)} \leq c \frac{n|G|^{2n} \log |G|}{r_c(G^n)}.
\]

For \( G = \mathbb{F}_2 \), the current upper bound in the literature is \( c_c(\mathbb{F}_2^n) \leq O(n^{2n/2}) \) [34], which we will improve on.

**Number on the forehead communication**

The corner problem is closely related to the Number On the Forehead (NOF) communication model [15]. In this model, \( k \) players wish to evaluate a function \( F : X_1 \times \cdots \times X_k \rightarrow \{0, 1\} \) on a given input \( x_1, \ldots, x_k \). The input is distributed among the players in a way that player \( i \) sees every \( x_j \) for \( j \neq i \). This scenario is visualized as \( x_i \) being written on the forehead of Player \( i \).

The computational power of everyone is unlimited, but the number of exchanged bits has to be minimized. Let \( D_k(F) \) be the minimum number of bits they need to communicate to compute the function \( F \) in the NOF model with \( k \) players. Many questions that have been thoroughly analyzed for the two-player case remain open in the general case of 3 or more players, where lower bounds on communication complexity are much more difficult to prove. The difficulty in proving lower bounds arises from the overlap in the inputs known to different players.

One interesting function in this context is the family of Eval functions. The function \( \text{Eval}_{G^n} : (G^n)^3 \rightarrow \{0, 1\} \) outputs 1 on inputs \( x_1, x_2, x_3 \in G^n \) if and only if \( x_1 + x_2 + x_3 = 0 \). The trivial algorithm gives that \( D_3(\text{Eval}_{G^n}) \leq \lceil n \log(|G|) \rceil + 1 \). For two players Yao [51] proved that \( D_2(\text{Eval}_{G^n}) = \Omega(n) \) (for nontrivial \( G \)). But, for three players it is an open problem whether \( D_3(\text{Eval}_{G^n}) = \Omega(n) \).
Lemma 17 ([9]). \( \log(c_L(G^n)) \leq D_3(\text{Eval}_{G^n}) \leq 2 + \log(c_L(G^n)) \).

From Lemma 17 and Proposition 16 it follows that \( \Theta(H_{\text{cor}, G}) < |G|^2 \) would imply that \( D_3(\text{Eval}_{G^n}) = \Omega(n) \), and also that lower bounds on \( r_L(G^n) \) give upper bounds on \( D_3(\text{Eval}_{G^n}) \). For \( G = \mathbb{F}_2 \), the best-known upper bound on \( D_3(\text{Eval}_{\mathbb{F}_2^n}) \) is \( 0.5n + O(\log n) \) [1] which we improve on.

Three-term arithmetic progressions and the cap set problem

A three-term arithmetic progression in \( G \) is a three-element set of the form \( \{x, x + \lambda, x + 2\lambda\} \) for some \( x, \lambda \in G \) and \( \lambda \neq 0 \). Let \( r_3(G) \) be the size of the largest subset \( S \subseteq G \) such that no three elements in \( S \) form a three-term arithmetic progression.

Following [52, Corollary 3.24] there is a simple relation between corner-free sets and three-term-arithmetic-progression-free sets:

\[ p^n r_3(\mathbb{F}_p^n) \leq r_L(\mathbb{F}_p^n) \]

A three-term-arithmetic-progression-free subset of \( \mathbb{F}_3^n \) is also called a cap set. The notorious cap set problem is to determine how \( r_3(\mathbb{F}_3^n) \) grows when \( n \) goes to infinity. A priori we have that \( 2^n \leq r_3(\mathbb{F}_3^n) \leq 3^n \). Using Fourier methods and the density increment argument of Roth, the upper bound \( r_3(\mathbb{F}_3^n) \leq \mathcal{O}(3^n/n) \) was obtained by Meshulam [39], and improved only as late as 2012 to \( \mathcal{O}(3^n/n^{1+\epsilon}) \) for some positive constant \( \epsilon \) by Bateman and Katz in [7]. Until recently it was not known whether \( r_3(\mathbb{F}_3^n) \) grows like \( 3^n - o(n) \) or like \( c^n - o(n) \) for some \( c < 3 \). Gijswijt and Ellenberg solved this question in 2017, showing that \( r_3(\mathbb{F}_3^n) \leq 2.756^n + o(n) \) [25].

The best lower bound is \( 2.2174^n \leq r_3(\mathbb{F}_3^n) \) by Edel [24]. In particular, using Lemma 18, this implies the lower bound \( 3^n \cdot 2.2174^n = 6.6522^n \leq r_L(\mathbb{F}_3^n) \) for the corner problem. We will improve this lower bound in Theorem 25.

We may phrase the cap set problem as a hypergraph independence problem by defining the undirected 3-uniform hypergraph \( H_{\text{cap}} \) consisting of three vertices \( \{0, 1, 2\} \) and a single edge \( e = \{0, 1, 2\} \). The independence number \( \alpha(H_{\text{cap}}) \) equals \( r_3(\mathbb{F}_3^n) \), and thus the Shannon capacity of \( H_{\text{cap}} \) determines the rate of growth of \( r_3(\mathbb{F}_3^n) \).

2.2 Probabilistic method

We start off with a simple and general method for obtaining lower bounds on the Shannon capacity. For any element \( g \in G \), the set \( \{(g, g + \lambda) : \lambda \in G\} \) is an independent set of \( H_{\text{cor}, G} \), and therefore we have \( \Theta(H_{\text{cor}, G}) \geq |G| \), which we think of as the trivial lower bound. By using a simple probabilistic argument (which does not use much of the structure of \( H_{\text{cor}, G} \)), we show the following nontrivial lower bound for \( \Theta(H_{\text{cor}, G}) \).

Proposition 19. For any finite Abelian group \( G \), we have \( \Theta(H_{\text{cor}, G}) \geq |G|^{3/2} \).

The idea in the proof of Proposition 19 to apply the probabilistic method to lower bound the number of remaining elements after a “pruning” procedure (in this case, pruning vertices that induce edges) goes back to [21]. A similar probabilistic method construction is the driving component in the recent new upper bound on the matrix multiplication exponent \( \omega \) [3].

In terms of the corner problem, the lower bound on the Shannon capacity in Proposition 19 for \( G = \mathbb{F}_2 \) corresponds to the upper bound \( c_L(\mathbb{F}_2^n) \leq \mathcal{O}(n2^n/2) \) (via Proposition 16). This upper bound is similar to the bound provided in [34, Corollary 26 in the ITCS version].

Remark 20. The proof of Proposition 19 directly extends from 2-dimensional corners to \( k \)-dimensional corners, which are sets of the form

\[ \{(x_1, x_2, \ldots, x_k), (x_1 + \lambda, x_2, \ldots, x_k), \ldots, (x_1, x_2, \ldots, x_k + \lambda)\} \]
Just like the Eval problem on 3 players is closely related to 2-dimensional corners in \((G^n)^2\), the Eval function on \(k + 1\) players is closely related to \(k\)-dimensional corners in \((G^n)^k\). By a similar argument as the proof of Lemma 17 we have that the \(k + 1\) player NOF complexity is upper bounded by \(D_{k+1}(\text{Eval}_{G^n}) \leq k + c_{k,\leq}(G^n)\), where \(c_{k,\leq}(G^n)\) is minimum number of colors that we can use to color \((G^n)^k\) such that no \(k\)-dimensional corner is monochromatic. Letting \(r_{k,\leq}(G^n)\) denote the size of the largest \(k\)-dimensional corner free set in \(G^n\), we have similar to Proposition 16 the relation between \(r_{k,\leq}(G^n)\) and \(c_{k,\leq}(G^n)\) given by

\[
\frac{|G|^{kn}}{r_{k,\leq}(G^n)} \leq c_{k,\leq}(G^n) \leq nk|G|^n \log(|G|) - r_{k,\leq}(G^n),
\]

which is proved in [34]. From a similar probabilistic method argument as in the proof of Proposition 19, choosing each \((x_1, \ldots, x_k) \in (G^n)^k\) independently at random with probability \(p = \frac{1}{(k+1)(|G|^{n-1})^{k+1}}\), we get

\[
r_{k,\leq}(G^n) \geq \frac{k|G|^n}{|G|^n/k(k+1)^{k+1}},
\]

as a consequence one has \(\Theta(H_{k,\text{cor},G} \geq |G|^{1-k/k})\), where \(H_{k,\text{cor},G}\) is directed \((k+1)\)-uniform hypergraph that construct for the \(k\)-dimensional corner. Furthermore from the lower bound of \(r_{k,\leq}(G^n)\), we have

\[
D_{k+1}(\text{Eval}_{G^n}) \leq \frac{n}{k} \log |G| + \log n + \log \log |G| + (1 + \frac{1}{k}) \log (1 + k) + k.
\]

If we take \(k = \log n\) (for instance), then \(D_{k+1}(\text{Eval}_{G^n}) \leq \frac{n}{\log n} \log |G| + O(log n)\), that is, we obtain a sublinear upper bound for \(D_{\log n}(\text{Eval}_{G^n})\) in \(n\).

### 2.3 Combinatorial degeneration method

We now introduce the combinatorial degeneration method for lower bounding Shannon capacity. Combinatorial degeneration is an existing concept from algebraic complexity theory introduced by Strassen in [46, Section 6, in particular 'Theorem 6.1']. In that original setting it was used as part of the construction of fast matrix multiplication algorithms [14, Definition 15.29 and Lemma 15.31], and, in a broader setting, combinatorial degeneration was used to construct large induced matchings in [4, Lemma 3.9], [2, Lemma 5.1] and [18, Theorem 4.11]. However, we will be using it in a novel manner in order to construct independent sets instead of induced matchings. We will subsequently apply the combinatorial degeneration method to get new bounds for the corner problem. We expect the method to be useful in the study of other problems besides the corner problem as well. First we must define combinatorial degeneration.

**Definition 21** (Combinatorial degeneration). Let \(I_1, \ldots, I_k\) be finite sets. Let \(\Phi \subseteq \Psi \subseteq I_1 \times \cdots \times I_k\). We say that \(\Phi\) is a combinatorial degeneration of \(\Psi\), and write \(\Psi \geq \Phi\), if there are maps \(u_i : I_i \rightarrow \mathbb{Z}\) (\(i \in [k]\)) such that for every \(x = (x_1, \ldots, x_k) \in I_1 \times \cdots \times I_k\), if \(x \in \Psi \setminus \Phi\), then \(\sum_{i=1}^{k} u_i(x_i) > 0\), and if \(x \in \Phi\), then \(\sum_{i=1}^{k} u_i(x_i) = 0\).

---

7 The precise connection to [46] is as follows. Strassen defines the notion of \(M\)-degeneration on tensors. In our terminology, a tensor is an \(M\)-degeneration of another tensor, if the support of the first is a combinatorial degeneration of the support of the second. The terminology “combinatorial degeneration”, which does not refer to tensors, but rather directly to their supports (hence the adjective “combinatorial”), was introduced in [14, Definition 15.29].
Example 22. As a quick example of a combinatorial degeneration, let
\[ \Phi = \{(0, 0, 0), (1, 1, 0), (1, 0, 1)\}, \]
\[ \Psi = \{(0, 0, 0), (1, 1, 0), (1, 0, 1), (0, 1, 1)\}. \]
Then we have a combinatorial degeneration \( \Psi \supseteq \Phi \) by picking the maps \( u_1(0) = u_2(0) = u_3(0) = 0 \), and \( u_1(1) = -1, u_2(1) = u_3(1) = 1 \).

We apply combinatorial degeneration in the following fashion to get Shannon capacity lower bounds:

Theorem 23 (Combinatorial degeneration method). Let \( H = (V, E) \) be a directed \( k \)-uniform hypergraph. Let \( S \subseteq V \). Let \( \Psi = E \cup \{(v, \ldots, v) : v \in V\} \) and let \( \Phi = \{(v, \ldots, v) : v \in S\} \) and suppose that \( \Psi \supseteq \Phi \). Then \( \Theta(H) \geq |S| \).

Motivated by Theorem 23 we have the following definition:

Definition 24. For any directed \( k \)-uniform hypergraph \( H = (V, E) \), we define \( \beta(H) \) to be the size of the largest subset \( S \subseteq V \) such that \( \{(v, \ldots, v) : v \in S\} \) is a combinatorial degeneration of \( E \cup \{(v, \ldots, v) : v \in V\} \).

Clearly, \( \Theta(H) \geq \beta(H) \) by Theorem 23.

In order to construct combinatorial degenerations we employ integer programming. To state the integer program, we let \( t \) be a variable that takes values in \( \{0, 1\}^{|V|} \) and let \( u_1, \ldots, u_k \) be variables that take values in \( \mathbb{Z}^{|V|} \). We choose \( M \in \mathbb{N} \) large enough. The parameter \( \beta(H) \) can be then computed by the following integer linear program:

\[
\begin{align*}
\text{max} & \quad \sum_{i \in V} t(i) \\
\text{subject to} & \quad u_1(i_1) + \cdots + u_k(i_k) \geq 1, \quad \forall (i_1, \ldots, i_k) \in E, \\
& \quad 1 - t(i) \leq u_1(i) + \cdots + u_k(i) \leq M(1 - t(i)), \quad \forall i \in V.
\end{align*}
\]

Indeed, if \( (t, u_1, \ldots, u_k) \) is a feasible solution of the program (1), then \( \{(v, \ldots, v) : v \in S\} \) is a combinatorial degeneration of \( E \cup \{(v, \ldots, v) : v \in V\} \) by choosing \( k \) integer maps \( u_1, \ldots, u_k \), where \( S = \{i \in V : t(i) = 1\} \). Therefore, one has \( \beta(H) \geq A \) (\( A \) is a maximum value of program (1)). On the other hand, for any \( S \subseteq V \) such that if there is a combinatorial degeneration from \( E \cup \{(v, \ldots, v) : v \in V\} \) to \( \{(v, \ldots, v) : v \in S\} \) with \( k \) integer maps \( u_1, \ldots, u_k \), by defining \( t \in \{0, 1\}^{|V|} \) so that \( t(i) = 1 \) iff \( i \in S \), we have \( (t, u_1, \ldots, u_k) \) is a feasible solution of the program (1). Thus, \( \beta(H) \leq A \).

As a first application of the combinatorial degeneration method (Theorem 23), we prove the following new bound for corners over \( \mathbb{F}_3^3 \) by lower bounding \( \beta \) (Definition 24).

Theorem 25. \( \beta(H_{\text{cor}, \mathbb{F}_3}) \geq 7 \) and thus \( \Theta(H_{\text{cor}, \mathbb{F}_3}) \geq 7 \).

In other words, \( 7^n / \text{poly}(n) \leq r_{\mathbb{C}}(\mathbb{F}_3^n) \). This improves on the lower bound 6.6522\( ^n \leq r_{\mathbb{C}}(\mathbb{F}_3^n) \) that can be obtained from Edel’s construction of cap sets [24] and Lemma 18. As a consequence of the new lower bound, we find the bounds \( c_{\mathbb{C}}(\mathbb{F}_3^n) \leq O(\text{poly}(n)(3/2)^n) \) and \( D_3(\text{Eval}_{\mathbb{F}_3^n}) \leq n \log(9/7) + O(\log n) \leq 0.37n + O(\log n) \). Previously, only the weaker bound \( D_3(\text{Eval}_{\mathbb{F}_3^n}) \leq n + O(\log n) \) was known [34].

In the previous proof we only considered the first power of the relevant hypergraph. For the next result we will be able to get good bounds by considering higher powers.

\[ ^8 \text{We note that, as far as we know, the NOF protocol for } \text{Eval}_{\mathbb{F}_3^n} \text{ given in [1] does not generalize to } \text{Eval}_{\mathbb{F}_3^n} \text{ in any direct way.} \]
Theorem 26. \( \beta(H^{2\log n}_{\text{cor},F_2}) \geq 11 \) and \( \beta(H^{2\log n}_{\text{cor},F_2}) \geq 39 \), as a consequence \( \Theta(H_{\text{cor},F_2}) \geq \sqrt{39} \).

In other words, \( (\sqrt{39})/n \) poly\((n) \leq r_{\text{c}}(F_2^n) \). As a consequence, we have the upper bound \( c_{\text{c}}(F_2^n) \leq \mathcal{O}(\text{poly}(n)(1/n)^n) \leq \mathcal{O}(\text{poly}(n)1.18^n) \) for the corner problem and the upper bound \( D_3(\text{Eval}_{F_2}) \leq \log((\sqrt{39}/n) \log n) \leq 0.24n + \mathcal{O}(\log n) \) for the eval problem.

We have yet to develop structural understanding of how the above combinatorial degenerations that exhibit the new capacity lower bounds arise, and leave the investigation of further generalizations and improvements to future work. As a partial remedy to our limited understanding, we introduce in the next section the acyclic method as a tool to construct combinatorial degenerations. While the acyclic method does not recover the bounds of Theorem 26 and Theorem 25, it has the merits of being transparent and simple to apply.

Remark 27. The above proof of Theorem 23 gives in fact the precise lower bound
\[
\alpha(H^{2\log n}) \geq \frac{|S|^n}{(n+1)|S|}.
\]
This lower bound is optimal up to a poly\((n)\) factor. A more careful analysis improves this poly\((n)\) factor, for which we refer to the full version [17].

2.4 Acyclic set method

The acyclic set method that we are about to introduce is modeled on the fact that the Shannon capacity of a directed graph \( G \) is at least the size of any induced acyclic subgraph of \( G \) [11]. We introduce the concept of an acyclic set in a directed \( k \)-uniform hypergraph as an extension of the notion of an induced acyclic subgraph.

Definition 28. Let \( H \) be a directed \( k \)-uniform hypergraph. We associate to \( H \) the directed graph \( G_H \) with vertices \( V(G) = V(H) \) and edges \( E(G) = \{(a_1,a_2) : (a_1,a_2,\ldots,a_k) \in E \text{ for some } a_3,\ldots,a_k\} \). For any subset \( A \subseteq V \) let \( H[A] \) denote the subhypergraph of \( H \) induced by \( A \), that is, \( H[A] \) is the directed \( k \)-uniform hypergraph with vertices \( S \) and edges \( E \cap A^k \). We call a subset \( A \subseteq V \) an acyclic set of \( H \) if the directed graph \( G_{H[A]} \) is a directed acyclic graph.

Note that, if \( A \) is an independent set of \( H \), then \( E(H[A]) = \emptyset \) and thus \( E(G_{H[A]}) = \emptyset \), and in particular \( A \) is an acyclic set of \( H \). On the other hand, acyclic sets are not necessarily independent sets. However, the existence of an acyclic set does imply strong lower bounds on the Shannon capacity (via combinatorial degeneration, as we will see):

Theorem 29. Let \( H \) be a directed \( k \)-uniform hypergraph. For any acyclic set \( A \) of \( H \), we have \( \Theta(H) \geq |A| \).

Theorem 29 follows directly from the combinatorial degeneration method (Theorem 23) and the following lemma:

Lemma 30. Let \( H = (V,E) \) be a directed \( k \)-uniform hypergraph. Let \( A \) be an acyclic set of \( H \). Then there is a combinatorial degeneration from \( E \cup \{(v,\ldots,v) : v \in V\} \) to \( \Phi = \{(v,\ldots,v) : v \in A\} \).

As can be seen from the proof of Lemma 30, the combinatorial degenerations that result from acyclic sets have a special form, and in particular the acyclic set method does not recover the full power of the combinatorial degeneration method. However the acyclic set method is much easier to apply than the combinatorial degeneration method. For example,
we can use the acyclic set method to quickly see that \( \Theta(H_{cor, \mathbb{R}_2}) \geq 3 \). Namely, it is verified directly that the set \( S = \{0, 1, 2\} \) of size three is an acyclic set in \( H_{cor, \mathbb{R}_2} \), which implies the claim by Theorem \( 29 \).

Finally, we note that for directed graphs \( (k = 2) \) the combinatorial degeneration method can be used to characterize whether the Shannon capacity is full or not.

**Theorem 31.** Let \( G = (V, E) \) be a directed graph. Then \( \Theta(G) = |V| \) if and only if there is a combinatorial degeneration from \( E \cup \{(v, v) : v \in V\} \) to \( \{(v, v) : v \in V\} \).

## 3 Upper bounds from tensor methods, and their limitations

In this section we discuss methods to obtain upper bounds on the Shannon capacity of directed \( k \)-uniform hypergraphs and we discuss limitations of these methods for hypergraphs like \( H_{cor, G} \). We will not be discussing all available methods, but rather some of the main ones: subrank and slice rank. The main point is to introduce the induced matching barrier and apply it to the corner problem.

We recall some standard tensor notation and definitions that we will use in the rest of the section. For \( d \in \mathbb{N} \) let \( [d] = \{1, \ldots, d\} \). Let \( \mathcal{P}(\{d\}) \) be the set of all probability distributions on \( \{d\} \). Let \( f \in \mathbb{F}^{d_1} \otimes \cdots \otimes \mathbb{F}^{d_k} \) be a \( k \)-tensor over a field \( \mathbb{F} \). Let \( \{e_1, \ldots, e_{d_j}\} \) denote the standard basis of \( \mathbb{F}^{d_j} \). We may then write \( f = \sum f_{i_1 \ldots i_k} e_{i_1} \otimes \cdots \otimes e_{i_k} \), where the sum goes over \( i \in [d_1] \times \cdots \times [d_k] \). In this way \( f \) corresponds to a \( k \)-way array \( f \in \mathbb{F}^{d_1 \times \cdots \times d_k} \). For \( f \in \mathbb{F}^{d_1} \otimes \cdots \otimes \mathbb{F}^{d_k} \) and \( f' \in \mathbb{F}^{d_1'} \otimes \cdots \otimes \mathbb{F}^{d_k'} \), we define the tensor product as \( (f \otimes f')_{i_1 \ldots i_k, j_1 \ldots j_k} = f_{i_1 \ldots i_k} f'_{j_1 \ldots j_k} \). We define the support of \( f \) as the set

\[
\text{supp}(f) := \{(i_1, \ldots, i_k) : f_{i_1 \ldots i_k} \neq 0\} \subseteq [d_1] \times \cdots \times [d_k].
\]

For \( r \in \mathbb{N} \), we call \( (r) := \sum_{i=1}^{r} e_i \otimes e_i \) the unit tensor of size \( r \).

### 3.1 Tensor methods: subrank, slice rank (and more)

We focus on two tensor methods here: subrank and slice rank. We begin by defining subrank, for which we need the notion of restriction of tensors [46]. We say that the tensor \( f \in \mathbb{F}^{d_1} \otimes \cdots \otimes \mathbb{F}^{d_k} \) restricts to \( f' \in \mathbb{F}^{d_1'} \otimes \cdots \otimes \mathbb{F}^{d_k'} \), and write \( f' \leq f \) if there exist linear maps \( A^{(i)} : \mathbb{F}^{d_i} \to \mathbb{F}^{d_i'} \) such that \( f' = (A^{(1)} \otimes \cdots \otimes A^{(k)}) \cdot f \). Written in the standard basis, this corresponds to having for all \( i_1 \in [d_1'], \ldots, i_k \in [d_k'] \) that

\[
f'_{i_1 \ldots i_k} = \sum_{j_1 \in [d_1] \ldots j_k \in [d_k]} A^{(1)}_{i_1, j_1} \cdots A^{(k)}_{i_k, j_k} f_{j_1 \ldots j_k}.
\]

**Example 32.** Here we see restriction in action in a small example. For the tensors

\[
f = e_0 \otimes e_0 \otimes e_0 + e_1 \otimes e_1 \otimes e_1, \quad f' = e_0 \otimes (e_0 \otimes e_0 + e_1 \otimes e_1),
\]

we have \( f' \leq f \) by letting \( A^{(1)} : e_0 \mapsto e_0, e_1 \mapsto e_0 \) and letting \( A^{(2)} \) and \( A^{(3)} \) both be the identity map.

Let \( (n) = \sum_{i \in [n]} e_i \otimes \cdots \otimes e_i \) be the unit tensor of rank \( n \). Strassen [46] defined the subrank of \( f \) as \( Q(f) := \max \{r \in \mathbb{N} : (r) \leq f\} \). Similarly, one may define the “opposite” of the subrank as \( R(f) := \min \{r \in \mathbb{N} : f \leq (r)\} \), which is called the rank and which coincides with the usual notion of tensor rank in terms of a rank-one decomposition. For \( k = 2 \), the
subrank and rank of $f$ are the usual matrix rank: $Q(f) = R(f) = \text{rank}(f)$. When $k \geq 3$, however, there are $f$ for which $Q(f) < \text{rank}(f)$. In fact, the tensor rank can be larger than the dimensions $d_1, \ldots, d_k$, whereas the subrank cannot exceed $\min_i d_i$.

Applications require us to understand the rate of growth of the subrank as we take tensor product powers of a fixed tensor. Strassen [46] defined the asymptotic subrank of $f \in \mathbb{F}^{d_1} \times \cdots \times \mathbb{F}^{d_k}$ as $\tilde{Q}(f) := \lim_{n \to \infty} Q(f^\otimes n)^{1/n}$. Since the subrank is super-multiplicative, we can, by Fekete’s lemma, replace the limit by a supremum.

The second tool we focus on is slice rank. Slice rank was introduced by Tao [48] and developed further in [49] and [12] as a variation on tensor rank to study cap sets and approaches to fast matrix multiplication algorithms. A tensor in $\mathbb{F}^{d_1} \times \cdots \times \mathbb{F}^{d_k}$ has slice rank one if it has the form $u \otimes v$ for $u \in \mathbb{F}^{d_i}$ and $v \in \mathbb{F}^{d_j}$ for some $i \in [k]$. The slice rank of $f$, denoted by $\text{SR}(f)$, is the smallest number $r$ such that $f$ can be written as sum of $r$ slice rank one tensors. Since slice rank is not sub-multiplicative and not super-multiplicative, the limit $\lim_{n \to \infty} \text{SR}(f^\otimes n)^{1/n}$ might not exist [18]. We define

$$\tilde{\text{SR}}(f) = \limsup_{n \to \infty} \text{SR}(f^\otimes n)^{1/n}. $$

Since slice rank is monotone under the restriction order and normalized on $\langle n \rangle$ [48], it follows that $Q(f) \leq \text{SR}(f)$ and $Q(f) \leq \tilde{\text{SR}}(f)$.

### 3.2 Induced matchings and tightness

Now we discuss the notion of induced matchings, and we will discuss Strassen’s theorem that gives a construction of large induced matchings under a tightness condition.

Let $H = (V, E)$ be a directed $k$-uniform hypergraph with adjacency tensor $A$. Let $\Phi_H$ be the support of $A$. A subset $D \subseteq \Phi_H$ is called a matching if any two distinct elements $a, b \in D$ differ in all $k$ coordinates, that is, $a_i \neq b_i$ for all $i \in [k]$. We call a matching $D \subseteq \Phi_H$ an induced matching if $D = \Phi_H \cap (D_1 \times \cdots \times D_k)$, where $D_i = \{a_i : a \in D\}$ is the projection of $D$ onto the $i$-th coordinate. We denote by $Q_{\text{IM}}(\Phi_H)$ the maximum size of an induced matching $D \subseteq \Phi_H$.

For two directed $k$-uniform hypergraphs $G = (V_G, E_G)$ and $H = (V_H, E_H)$, let $\Phi_G$ and $\Phi_H$ be the support of the adjacency tensors of $G$ and $H$, respectively. We define the product $\Phi_G \times \Phi_H \subseteq (V_G \times V_H) \times \cdots \times (V_G \times V_H)$ by $\Phi_G \times \Phi_H := \{(a_1, b_1), \ldots, (a_k, b_k) : a \in \Phi_G, b \in \Phi_H\}$. The asymptotic induced matching number of $H$ is defined as $\tilde{Q}_{\text{IM}}(\Phi_H) := \lim_{n \to \infty} Q_{\text{IM}}(\Phi_H^{\otimes n})^{1/n} = \sup_n Q_{\text{IM}}(\Phi_H^{\otimes n})^{1/n}$.

The induced matching number should be thought of as the combinatorial version of the subrank, as follows. Let $\Phi_H$ be the support of the adjacency tensor $A_H$ of a directed $k$-uniform hypergraph $H$. Then the induced matching number $Q_{\text{IM}}(\Phi_H)$ is the largest number $n$ such that $\langle n \rangle$ can be obtained from $A_H$ using a restriction that consists of matrices that have at most one nonzero entry in each row and in each column. Therefore, $Q_{\text{IM}}(\Phi_H) \leq Q(A_H)$.

**Lemma 33.** Let $H$ be a directed $k$-uniform hypergraph and $A_H$ its adjacency tensor with support $\Phi_H = \text{supp}(A_H)$. Then

$$\Theta(H) \leq \tilde{Q}_{\text{IM}}(\Phi_H) \leq \tilde{Q}(A_H).$$

Next, we discuss tight sets, a notion introduced by Strassen [47].

**Definition 34** ([47], see also [18]). Let $I_1, \ldots, I_k$ be finite sets. We call any subset $\Phi \subseteq I_1 \times \cdots \times I_k$ tight if there are injective maps $u_i : I_i \to \mathbb{Z}$ for every $i \in [k]$ such that:

$$u_1(a_1) + \cdots + u_k(a_k) = 0 \text{ for every } (a_1, \ldots, a_k) \in \Phi.$$
When $\Phi_H$ is tight, the asymptotic induced matching number is essentially known, and can be described as a simple optimization. To explain the precise formula we recall some definitions.

For any finite set $X$, let $\mathcal{P}(X)$ be the set of all distributions on $X$. For any probability distribution $P \in \mathcal{P}(X)$ the Shannon entropy of $P$ is defined as $H(P) := -\sum_{x \in X} P(x) \log_2 P(x)$ with $0 \log_2 0 = 0$. Given finite sets $I_1, \ldots, I_k$ and a probability distribution $P \in \mathcal{P}(I_1 \times \cdots \times I_k)$ on the product set $I_1 \times \cdots \times I_k$ we denote the marginal distribution of $P$ on $I_i$ by $P_i$, that is, $P_i(a) = \sum_{x : x_i = a} P(x)$ for any $a \in I_i$.

▶ Theorem 35 ([47]). Let $H$ be a directed 3-uniform hypergraph. If $\Phi_H$ is tight, then

$$\tilde{Q}_{\text{IM}}(\Phi_H) = \max_{P \in \mathcal{P}(\Phi_H)} \min_{i \in [3]} 2^{H(P_i)}.$$

In particular, Theorem 35 implies that, for any directed 3-uniform hypergraph $H = (V, E)$ if there is a distribution $P$ on $\Phi_H$ such that every marginal distribution $P_i$ is uniform on $V$, then $\Phi_H$ has asymptotically maximal induced matchings.

Note that Theorem 35 only applies to directed $k$-uniform hypergraphs for $k = 3$. For the higher-order case $k > 3$ an extension of the lower bound of Theorem 35 was proven in [19, Theorem 1.2.4].

### 3.3 The corner hypergraph is tight

We will now apply Theorem 35 to the corner problem. First we see how the tightness property is satisfied by the corner problem by a simple construction.

▶ Theorem 36. For any finite Abelian group $(G, +)$, let $\Phi_{H_{\text{cor}}, G}$ be the support of the adjacency tensor of $H_{\text{cor}}, G$. Then the set $\Phi_{H_{\text{cor}}, G}$ is tight.

As a consequence of Theorem 36 and Theorem 35, we find almost directly that the asymptotic induced matching number of the corner hypergraph is maximal:

▶ Corollary 37. For any group $G$, $\tilde{Q}_{\text{IM}}(H_{\text{cor}}, G) = |G|^2$.

In particular, Corollary 37 implies that no better upper bound on $\Theta(H_{\text{cor}}, G)$ can be obtained via methods that also upper bound the asymptotic induced matching number $\tilde{Q}_{\text{IM}}(H_{\text{cor}}, G)$. Such methods include the slice rank, the analytic rank, the geometric rank and the G-stable rank.

---

References

1. Anil Ada, Arkadev Chattopadhyay, Omar Fawzi, and Phuong Nguyen. The NOF multiparty communication complexity of composed functions. *Computational Complexity*, 24(3):645–694, 2015. doi:10.1007/s00037-013-0078-4.
2. Josh Alman and Virginia Vassilevska Williams. Further limitations of the known approaches for matrix multiplication. In *9th Innovations in Theoretical Computer Science Conference (ITCS 2018)*, pages 25:1–25:15, 2018. doi:10.4230/LIPIcs.ITCS.2018.25.
3. Josh Alman and Virginia Vassilevska Williams. A refined laser method and faster matrix multiplication. In *Proceedings of the 2021 ACM-SIAM Symposium on Discrete Algorithms (SODA 2021)*, pages 522–539, 2021. doi:10.1137/1.9781611976465.32.
4. Noga Alon, Amir Shpilka, and Christopher Umans. On sunflowers and matrix multiplication. *computational complexity*, 22(2):219–243, 2013. doi:10.1007/s00037-013-0060-1.
5. Noga Alon and Adi Shraibman. Algorithmic number on the forehead protocols yielding dense Ruzsa-Szemeredi graphs and hypergraphs, 2020. arXiv:2001.00387.
László Babai, Anna Gál, Peter G. Kimmel, and Satyanarayana V. Lokam. Communication complexity of simultaneous messages. *SIAM J. Comput.*, 33(1):137–166, 2004. doi:10.1137/S0097539700375944.

Michael Bateman and Nets Hawk Katz. New bounds on cap sets. *J. Amer. Math. Soc.*, 25(2):585–613, 2012. doi:10.1090/S0894-0347-2011-00725-X.

Paul Beame, Matei David, Toniann Pitassi, and Philipp Woelfel. Separating deterministic from nondeterministic NOF multiparty communication complexity. In *International Colloquium on Automata, Languages, and Programming (ICALP 2007)*, pages 134–145, 2007. doi:10.1007/978-3-540-73420-8_14.

Richard Beigel, William Gasarch, and James Glenn. The multiparty communication complexity of Exact-T: Improved bounds and new problems. In *International Symposium on Mathematical Foundations of Computer Science (MFCS 2006)*, pages 146–156, 2006. doi:10.1007/11821069_13.

Richard Beigel and Jun Tarui. On ACC. *Computational Complexity*, 4(4):350–366, 1994. doi:10.1007/BF01263423.

E. Bidamon and H. Meyniel. On the Shannon capacity of a directed graph. *European J. Combin.*, 6(4):289–290, 1985. doi:10.1016/S0195-6698(85)80042-1.

Jonah Blasiak, Thomas Church, Henry Cohn, Joshua A. Grochow, Eric Naslund, William F. Sawin, and Chris Umans. On cap sets and the group-theoretic approach to matrix multiplication. *Discrete Anal.*, 2017. doi:10.19086/da.1245.

Jop Briët. Subspaces of tensors with high analytic rank, 2019. arXiv:1908.04169.

Peter Bürgisser, Michael Clausen, and M. Amin Shokrollahi. *Algebraic complexity theory*, volume 315 of *Grundlehren der Mathematischen Wissenschaften*. Springer-Verlag, Berlin, 1997. doi:10.1007/978-3-662-03338-8.

Ashok K. Chandra, Merrick L. Furst, and Richard J. Lipton. Multi-party protocols. In *Proceedings of the 15th Annual ACM Symposium on Theory of Computing (STOC 1983)*, pages 94–99, 1983. doi:10.1145/800061.808737.

Arkadev Chattopadhyay and Michael E. Saks. The power of super-logarithmic number of players. In *Approximation, Randomization, and Combinatorial Optimization. Algorithms and Techniques (APPROX/RANDOM 2014)*, volume 28, pages 596–603, 2014. doi:10.4230/LIPIcs.APPROX-RANDOM.2014.596.

Matthias Christandl, Omar Fawzi, Hoang Ta, and Jeroen Zuiddam. Larger corner-free sets from combinatorial degenerations, 2021. arXiv:2111.08262.

Matthias Christandl, Péter Vrana, and Jeroen Zuiddam. Universal points in the asymptotic spectrum of tensors. In *Proceedings of the 50th Annual ACM SIGACT Symposium on Theory of Computing (STOC 2018)*, pages 289–296, 2018. doi:10.1145/3188745.3188766.

Matthias Christandl, Péter Vrana, and Jeroen Zuiddam. Asymptotic tensor rank of graph tensors: beyond matrix multiplication. *Comput. Complex.*, 28(1):57–111, 2019. doi:10.1007/s00037-018-0172-8.

H. Cohn, R. Kleinberg, B. Szegedy, and C. Umans. Group-theoretic algorithms for matrix multiplication. In *Proceedings of the 46th Annual IEEE Symposium on Foundations of Computer Science (FOCS 2005)*, pages 379–388, 2005. doi:10.1109/SFCS.2005.39.

Don Coppersmith and Shmuel Winograd. Matrix multiplication via arithmetic progressions. In *Proceedings of the nineteenth annual ACM symposium on Theory of computing*, pages 1–6. ACM, 1987.

Ernie Croot, Vsevolod F. Lev, and Péter Pál Pach. Progression-free sets in $\mathbb{Z}_q^n$ are exponentially small. *Annals of Mathematics*, pages 331–337, 2017. doi:10.4007/annals.2017.185.1.7.

Harm Derksen. The G-stable rank for tensors, 2020. arXiv:2002.08435.

Yves Edel. Extensions of generalized product caps. *Designs, Codes and Cryptography*, 31(1):5–14, 2004. doi:10.1023/A:1027365901231.

Jordan S. Ellenberg and Dion Gijswijt. On large subsets of $\mathbb{F}_q^n$ with no three-term arithmetic progression. *Ann. of Math.*, 185(1):339–343, 2017. doi:10.4007/annals.2017.185.1.8.
Terence Tao and Will Sawin. Notes on the “slice rank” of tensors. Tao’s blog post, 2016. URL: https://terrytao.wordpress.com/2016/08/24/notes-on-the-slice-rank-of-tensors/.

Emanuele Viola. Guest column: Non-abelian combinatorics and communication complexity. SIGACT News, 50(3):52–74, 2019. doi:10.1145/3364626.3364637.

Andrew Chi-Chih Yao. Some complexity questions related to distributive computing. In Proceedings of the Eleventh Annual ACM Symposium on Theory of Computing (STOC 1979), pages 209–213, 1979. doi:10.1145/800135.804414.

Yufei Zhao. Graph theory and additive combinatorics. Lecture Notes, 2019. URL: https://yufeizhao.com/gtac/.