High rate locally-correctable and locally-testable codes with sub-polynomial query complexity

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Abstract

In this work, we construct the first locally-correctable codes (LCCs), and locally-testable codes (LTCs) with constant rate, constant relative distance, and sub-polynomial query complexity. Specifically, we show that there exist binary LCCs and LTCs with block length $n$, constant rate (which can even be taken arbitrarily close to 1), constant relative distance, and query complexity $\exp(O(\sqrt{\log n}))$. Previously such codes were known to exist only with $\Omega(n^\beta)$ query complexity (for constant $\beta > 0$), and there were several, quite different, constructions known.

Our codes are based on a general distance-amplification method of Alon and Luby [AL96]. We show that this method interacts well with local correctors and testers, and obtain our main results by applying it to suitably constructed LCCs and LTCs in the non-standard regime of sub-constant relative distance.

Along the way, we also construct LCCs and LTCs over large alphabets, with the same query complexity $\exp(O(\sqrt{\log n}))$, which additionally have the property of approaching the Singleton bound: they have almost the best-possible relationship between their rate and distance. This has the surprising consequence that asking for a large alphabet error-correcting code to further be an LCC or LTC with $\exp(O(\sqrt{\log n}))$ query complexity does not require any sacrifice in terms of rate and distance! Such a result was previously not known for any $o(n)$ query complexity.

Our results on LCCs also immediately give locally-decodable codes (LDCs) with the same parameters.

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1 Introduction

Locally-correctable codes $\text{BFLS91 STV01 KT00}$ and locally-testable codes $\text{FS95 RS96 GS06}$ are codes that admit local algorithms for decoding and testing respectively. More specifically:

- We say that a code $C$ is a **locally-correctable code (LCC)** if there is a randomized algorithm that, when given a string $z$ that is close to a codeword $c \in C$, and a coordinate $i$, computes $c_i$ while making only a small number of queries to $z$.

- We say that a code $C$ is a **locally-testable code (LTC)** if there is a randomized algorithm that, when given a string $z$, decides whether $z$ is a codeword of $C$, or far from $C$, while making only a small number of queries to $z$.

The number of queries that are used by the latter algorithms is called the **query complexity**.

Besides being interesting in their own right, LCCs and LTCs have also played important roles in different areas of complexity theory, such as hardness amplification and derandomization (see e.g. $\text{STV01}$), and probabilistically checkable proofs $\text{AS98 ALM+98}$. It is therefore a natural and well-known question to determine what are the best parameters that LCCs and LTCs can achieve.

LCCs and LTCs were originally studied in the setting where the query complexity was either constant or poly-logarithmic. In those settings, it is believed that LCCs and LTCs must be very redundant, since every bit of the codeword must contain, in some sense, information about every other bit of the codeword. Hence, we do not expect such codes to achieve a high rate. In particular, in the setting of constant query complexity, it is known that linear LCCs cannot have constant rate $\text{KT00 WdW05 Woo07}$, and that LTCs with certain restrictions cannot have constant rate $\text{DK11 BSV12}$. On the other hand, the best-known constant-query LCCs have exponential length $\text{KSY14}$, and the best-known constant-query LTCs have quasi-linear length (see e.g. $\text{BS08 Din07 Vid15}$).

It turns out that the picture is completely different when allowing the query complexity to be much larger. In this setting, it has long been known that one can have LCCs and LTCs with constant rate and query complexity $O(n^\beta)$ for constant $\beta > 0$ $\text{BFLS91 RS96}$. More recently, it has been discovered that both LCCs $\text{KSY14 GKS13 HOW13}$ and LTCs $\text{Vid11 GKS13}$ can simultaneously achieve rates that are arbitrarily close to 1 and query complexity $O(n^\beta)$ for an arbitrary constant $\beta > 0$. This is in contrast with the general belief that local correctability and testability require much redundancy.

In this work, we show that there are LCCs and LTCs with constant rate (which can in fact be taken to be arbitrarily close to 1) and constant relative distance, whose associated local algorithms have $n^\omega(1)$ query complexity and running time. We find it quite surprising in light of the fact that there were several quite different constructions of LCCs and LTCs $\text{BFLS91 RS96 KSY14 Vid11 GKS13 HOW13}$ with constant rate and constant relative distance, all of which had $\Omega(n^\beta)$ query complexity.

Furthermore, we show that over large alphabets, such codes can approach the Singleton bound: they achieve a tradeoff between rate and distance which is essentially as good as possible for general

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1 There is a closely related notion of locally decodable codes (LDCs) that is more popular and very well studied. All our results for LCCs hold for LDCs as well, see discussion at the end of the introduction.

2 $\text{KT00 WdW05 Woo07}$ proved a lower bound for the related notion of LDCs. Since every linear LCC is also an LDC, their lower bound applies to linear LCCs as well.

3 For example, a constant-degree Reed-Muller code is such an LCC.
error-correcting codes. Such a result was previously not known for any $o(n)$ query complexity. This means that, remarkably, local correctability and local testability with $n^{o(1)}$ queries over large alphabets is not only possible with constant rate and constant relative distance, but it also does not require “paying” anything in terms of rate and relative distance.

We first state our theorems for the binary alphabet.

**Theorem 1.1** (Binary LCCs with sub-polynomial query complexity). For every $r \in (0, 1)$, there exist $\delta > 0$ and an explicit infinite family of binary linear codes $\{C_n\}_n$ satisfying:

1. $C_n$ has block length $n$, rate at least $r$, and relative distance at least $\delta$,
2. $C_n$ is locally correctable from $\frac{\delta}{2}$-fraction of errors with query complexity and running time at most $\exp(\sqrt{\log n \cdot \log \log n})$.

**Theorem 1.2** (Binary LTCs with sub-polynomial query complexity). For every $r \in (0, 1)$, there exist $\delta > 0$ and an explicit infinite family of binary linear codes $\{C_n\}_n$ satisfying:

1. $C_n$ has block length $n$, rate at least $r$, and relative distance at least $\delta$,
2. $C_n$ is locally testable with query complexity and running time at most $\exp(\sqrt{\log n \cdot \log \log n})$.

The binary LCCs and LTCs in the above theorems are obtained by first constructing LCCs and LTCs over large alphabets, and then concatenating them with binary codes. The following theorems describe these large alphabet LCCs and LTCs, which in addition to having sub-polynomial query complexity, also approach the Singleton bound.

**Theorem 1.3** (LCCs with sub-polynomial query complexity approaching the Singleton bound). For every $r \in (0, 1)$, there exists an explicit infinite family of linear codes $\{C_n\}_n$ satisfying:

1. $C_n$ has block length $n$, rate at least $r$, and relative distance at least $1 - r - o(1)$,
2. $C_n$ is locally correctable from $\frac{1-r-o(1)}{2}$-fraction of errors with query complexity and running time at most $\exp(\sqrt{\log n \cdot \log \log n})$,
3. The alphabet of $C_n$ is of size at most $\exp(\exp(\sqrt{\log n \cdot \log \log n}))$.

**Theorem 1.4** (LTCs with sub-polynomial query complexity approaching the Singleton bound). For every $r \in (0, 1)$, there exists an explicit infinite family of linear codes $\{C_n\}_n$ satisfying:

1. $C_n$ has block length $n$, rate at least $r$, and relative distance at least $1 - r - o(1)$,
2. $C_n$ is locally testable with query complexity and running time at most $\exp(\sqrt{\log n \cdot \log \log n})$,
3. The alphabet of $C_n$ is of size at most $\exp(\exp(\sqrt{\log n \cdot \log \log n}))$.

The above theorems are proved in Sections 3 and 4.

**Remark 1.5.** If we were only interested in LCCs and LTCs with $O(n^\beta)$ query complexity (for arbitrary $\beta$), we could have constructed binary codes that meet the Zyablov bound, which is the best-known rate-distance tradeoff for explicit binary codes. Furthermore, we could have constructed codes over constant-size alphabet that approach the Singleton bound (rather than having alphabet of super-constant size).

Moreover, our results imply the existence of non-explicit binary LCCs/LTCs with query complexity $\exp(\sqrt{\log n \cdot \log \log n})$ that meet the Zyablov bound. This follows by concatenating the codes of Theorems 1.3 and 1.4 with (non-explicit) Gilbert-Varshamov codes [Gil52, Var57].
The Alon-Luby distance-amplification. Our constructions are based on the distance-amplification technique of [AL96]. This distance amplifier, based on a $d$-regular expander, converts an error-correcting code with relative distance $\gg 1/d$ into an error-correcting code with larger relative distance $\delta$, while reducing the rate only by a factor of $\approx (1 - \delta)$. Thus for a large enough constant $d$, if we start with a code of rate $1 - \varepsilon$ and relative distance $\gg 1/d$, where $\varepsilon \ll \delta$, then after distance amplification with a $d$-regular expander, we get a code with rate $(1 - \delta)(1 - \varepsilon) \approx (1 - \delta)$ and relative distance $\delta$.

The original application of this technique in [AL96] was to construct linear-time erasure-decodable codes approaching the Singleton bound. In addition to the above distance-amplification technique, [AL96] constructed a linear-time erasure-decodable code (not approaching the Singleton bound) which could be used as the input code to the amplifier. The main result of [AL96] then follows from the fact that distance amplification via a constant-degree expander preserves linear-time erasure-decodability.

Subsequent applications of this distance-amplification technique followed a similar outline. One first constructs codes with high rate with some (possibly very small) constant relative distance and a certain desirable property. Then, applying distance amplification with a (possibly very large) constant-degree expander, one obtains a code with a much better tradeoff between its rate and relative distance. Finally one shows that the distance amplification with a constant degree expander preserves the desirable property. This scheme was implemented in [GI05], who constructed codes that can be decoded in linear time from errors (rather than erasures), and in [GI02, GR08], who constructed capacity-achieving list-decodable codes with constant alphabet.

Our observations. The first main observation of this paper is that the distance-amplification technique also preserves the property of being an LCC or an LTC. Specifically, if we start with an LCC or LTC with query complexity $q$, and then apply distance amplification with a $d$-regular expander, then the resulting code is an LCC/LTC with query complexity $q \cdot \text{poly}(d)$.

The next main observation is that this connection continues to hold even if we take $d$ to be super-constant, and take the LCC or LTC to have sub-constant relative distance $\Theta(1/d)$ (and then we only require the LCC to be able to correct strings whose distance from the code is within some constant fraction of the minimum distance of the code). This is potentially useful, since we only blow up the query complexity by a factor of poly($d$), and perhaps LCCs/LTCs with high rate and sub-constant relative distance can have improved query complexity over their constant relative distance counterparts.

Finally, we show that existing families of high rate LCCs and LTCs can achieve sub-polynomial query complexity if we only require them to have sub-constant relative distance. Specifically, multiplicity codes [KSY14] in a super-constant number of variables give us the desired LCCs, and super-constant-wise tensor products [Vid11] give us the desired LTCs.

As far as we are aware, there have been no previous uses of this distance-amplification technique using an expander of super-constant degree.

More generally, we wish to draw attention to the technique of [AL96]. We believe that it should be viewed as a general scheme for improving the rate-distance tradeoff for codes with certain desirable properties. In particular, it may transfer properties that codes with constant rate and sub-constant relative distance are known to have, to codes with constant rate and constant relative distance, and even to codes approaching the Singleton bound. We believe that this is a good “take-home message” from this work.
Correctable and testable codes. Using the above method, it is also possible to construct improved codes that are simultaneously locally correctable and locally testable. This can be done by applying the distance-amplification technique to the lifted Reed-Solomon codes of [GKS13]. The codes of [GKS13] are both locally correctable and testable, and achieve rates that are arbitrarily close to 1. Using these codes of [GKS13] in the sub-constant relative distance regime, and combining with our framework, we get codes of constant rate and constant relative distance (which over large alphabets approach the Singleton bound) that are both locally correctable and locally testable with $n^{O(1/\log \log n)}$ queries.

Locally decodable codes. An important variant of LCCs are locally decodable codes (LDCs). Those codes are defined similarly to LCCs, with the following difference: Recall that in the definition of LCCs, the decoder gets access to a string $z$ which is close to a codeword $c$, and is required to decode a coordinate of $c$. In the definition of LDCs, we view the codeword $c$ as the encoding of some message $x$, and the decoder is required to decode a coordinate of $x$. LDCs were studied extensively in the literature, perhaps more so than LCCs (see [Yek12] for a survey). One notable fact about LDCs is that there are constructions of LDCs with a constant query complexity and sub-exponential length [Yek08, Rag07, KY09, Efr12].

If we restrict ourselves to linear codes, then LDCs are a weaker object than LCCs, since every linear LCC can be converted into an LDC by choosing a systematic encoding map. Since the LCCs we construct in this paper are linear, all our results apply to LDCs as well.

Organization of this paper. We review the required preliminaries in Section 2, construct our LCCs in Section 3, and construct our LTCs in Section 4. We conclude with some open questions in Section 5.

Version. A preliminary version of this paper appeared as [Mei14], where the distance-amplification technique was used to construct codes approaching the Singleton bound with query complexity $O(n^{\beta})$ (for arbitrary $\beta > 0$).

2 Preliminaries

All logarithms in this paper are in base 2. For any $n \in \mathbb{N}$ we denote $[n] \overset{\text{def}}{=} \{1, \ldots, n\}$. We denote by $\mathbb{F}_2$ the finite field of two elements. For any finite alphabet $\Sigma$ and any pair of strings $x, y \in \Sigma^n$, the relative Hamming distance (or, simply, relative distance) between $x$ and $y$ is the fraction of coordinates on which $x$ and $y$ differ, and is denoted by $\text{dist}(x, y) \overset{\text{def}}{=} |\{i \in [n] : x_i \neq y_i\}|/n$. We have the following useful approximation.

Fact 2.1. For every $x, y \in \mathbb{R}$ such that $0 \leq x \cdot y \leq 1$, it holds that

$$(1 - x)^y \leq 1 - \frac{1}{4} \cdot x \cdot y.$$

Proof. It holds that

$$(1 - x)^y \leq e^{-x \cdot y} \leq 1 - \frac{1}{4} \cdot x \cdot y.$$

This conversion will lead to an LDC with the same query complexity, but the running time of the local decoder will be small only if the systematic encoding map can be computed efficiently.
The second inequality relies on the fact that \( 1 - \frac{1}{4} \cdot x \geq e^{-x} \) for every \( x \in (0, 1) \), which can be proved by noting that \( 1 - \frac{1}{4} \cdot x = e^{-x} \) at \( x = 0 \), and that the derivative of \( e^{-x} \) is smaller than that of \( 1 - \frac{1}{4} \cdot x \) for every \( x \in (0, 1) \). The first inequality relies on the fact that \( 1 - x \leq e^{-x} \) for every \( x \in \mathbb{R} \), which can be proved using similar considerations.

\[ \text{Definition 2.3.} \]

\[ \text{Fact 2.2} \quad \text{(Reed-Solomon Codes \cite{rs60}).} \quad \text{For every } k,n \in \mathbb{N} \text{ such that } n \geq k, \text{ and for every finite field } F \text{ such that } |F| \geq n, \text{ there exists an } F\text{-linear code } RS_{k,n} \subseteq F^n \text{ with rate } r = k/n, \text{ and relative distance at least } 1 - \frac{k-1}{n} > 1 - r. \text{ Furthermore, } RS_{k,n} \text{ has an encoding map } E : F^k \to RS_{k,n} \text{ which can be computed in time } \text{poly}(n, \log |F|), \text{ and can be decoded from up to } (1 - \frac{k-1}{n})/2 \text{ fraction of errors in } \text{time } \text{poly}(n, \log |F|). \]

### 2.2 Locally-correctable codes

Intuitively, a code is said to be locally correctable \cite{bfls91, stv01, kt00} if, given a codeword \( c \in C \) that has been corrupted by some errors, it is possible to decode any coordinate of \( c \) by reading only a small part of the corrupted version of \( c \). Formally, it is defined as follows.

\[ \text{Definition 2.3.} \quad \text{We say that a code } C \subseteq \Sigma^n \text{ is locally correctable from } \tau \text{-fraction of errors with query complexity } q \text{ if there exists a randomized algorithm } A \text{ that satisfies the following requirements:} \]

- **Input:** \( A \) takes as input a coordinate \( i \in [n] \) and also gets oracle access to a string \( z \in \Sigma^n \) that is \( \tau \)-close to a codeword \( c \in C \).

- **Output:** \( A \) outputs \( c_i \) with probability at least \( \frac{2}{3} \).

- **Query complexity:** \( A \) makes at most \( q \) queries to the oracle \( z \).

We say that the algorithm \( A \) is a local corrector of \( C \). Given an infinite family of LCCs \( \{C_n\}_n \), a uniform local corrector for the family is a randomized oracle algorithm that given \( n \), computes the
local corrector of $C_n$. We will often be also interested in the running time of the uniform local corrector.

Remark 2.4. The above success probability of $\frac{2}{3}$ can be amplified using sequential repetition, at the cost of increasing the query complexity. Specifically, amplifying the success probability to $1 - e^{-t}$ requires increasing the query complexity by a factor of $O(t)$.

2.3 Locally-testable codes

Intuitively, a code is said to be locally testable [FS95, RS96, GS00] if, given a string $z \in \Sigma^n$, it is possible to determine whether $z$ is a codeword of $C$, or rather far from $C$, by reading only a small part of $z$. There are two variants of LTCs in the literature, “weak” LTCs and “strong” LTCs. From now on, we will work exclusively with strong LTCs, since it is a simpler notion and allows us to state a stronger result.

Definition 2.5. We say that a code $C \subseteq \Sigma^n$ is (strongly) locally testable with query complexity $q$ if there exists a randomized algorithm $A$ that satisfies the following requirements:

- **Input:** $A$ gets oracle access to a string $z \in \Sigma^n$.
- **Completeness:** If $z$ is a codeword of $C$, then $A$ accepts with probability 1.
- **Soundness:** If $z$ is not a codeword of $C$, then $A$ rejects with probability at least $\text{dist}(z, C)$.
- **Query complexity:** $A$ makes at most $q$ non-adaptive queries to the oracle $z$.

We say that the algorithm $A$ is a local tester of $C$. Given an infinite family of LTCs $\{C_n\}_n$, a uniform local tester for the family is a randomized oracle algorithm that given $n$, computes the local tester of $C_n$. Again, we will often also be interested in the running time of the uniform local tester.

A remark on amplifying the rejection probability. It is common to define strong LTCs with an additional parameter $\rho$, and have the following soundness requirement:

- If $z$ is not a codeword of $C$, then $A$ rejects with probability at least $\rho \cdot \text{dist}(z, C)$.

Our definition corresponds to the special case where $\rho = 1$. However, given an LTC with $\rho < 1$, it is possible to amplify $\rho$ up to 1 at the cost of increasing the query complexity. Hence, we chose to fix $\rho$ to 1 in our definition, which somewhat simplifies the presentation.

The amplification of $\rho$ is performed as follows: The amplified tester invokes the original tester $A$ for $\left\lceil \frac{1}{\rho} \right\rceil$ times, and accepts only if all invocations of $A$ accept. Clearly, this increases the query complexity by a factor of $\frac{1}{\rho}$ and preserves the completeness property. To analyze the rejection probability, let $z$ be a string that is not a codeword of $C$, and observe that the amplified tester rejects it with probability at least

$$1 - (1 - \rho \cdot \text{dist}(z, C))^{\left\lceil \frac{1}{\rho} \right\rceil} \geq 1 - \left(1 - \frac{4}{\rho} \cdot \text{dist}(z, C)\right) \quad \text{(Fact 2.1)}$$

$$= \text{dist}(z, C),$$

as required.
2.4 Expander graphs

Expander graphs are graphs with certain pseudorandom connectivity properties. Below, we state the construction and properties that we need. The reader is referred to [HLW06] for a survey. For a graph \( G \), a vertex \( s \) and a set of vertices \( T \), let \( E(s, T) \) denote the set of edges that go from \( s \) into \( T \).

**Definition 2.6.** Let \( G = (U \cup V, E) \) be a bipartite \( d \)-regular graph with \( |U| = |V| = n \). We say that \( G \) is an \((\alpha, \gamma)\)-sampler if the following holds for every \( T \subseteq V \): For at least \( 1 - \alpha \) fraction of the vertices \( s \in U \) it holds that

\[
\frac{|E(s, T)|}{d} - \frac{|T|}{n} \leq \gamma.
\]

**Lemma 2.7.** For every \( \alpha, \gamma > 0 \) and every sufficiently large \( n \in \mathbb{N} \) there exists a bipartite \( d \)-regular graph \( G_{n,\alpha,\gamma} = (U \cup V, E) \) with \( |U| = |V| = n \) and \( d = \text{poly}\left(\frac{1}{\alpha \cdot \gamma}\right) \) such that \( G_{n,\alpha,\gamma} \) is an \((\alpha, \gamma)\)-sampler. Furthermore, there exists an algorithm that takes as inputs \( n, \alpha, \gamma \), and a vertex \( w \) of \( G_{n,\alpha,\gamma} \), and computes the list of the neighbors of \( w \) in \( G_{n,\alpha,\gamma} \) in time \( \text{poly}(\log n) \).

**Proof sketch.** A full proof of Lemma 2.7 requires several definitions and lemmas that we have not stated, such as second eigenvalue, edge expansion, and the expander mixing lemma. Since this is not the focus of this paper, we only sketch the proof without stating those notions. The interested reader is referred to [HLW06].

Let \( \alpha, \gamma \) and \( n \) be as in the lemma. We sketch the construction of the graph \( G_{n,\alpha,\gamma} \) as defined above. First, observe that it suffices to construct a strongly-explicit non-bipartite graph \( G \) over \( n \) vertices (that is, a graph \( G \) in which the neighborhood of any given vertex is computable in time \( \text{poly}(\log n) \)) with the desired property. The reason is that each such graph \( G \) can be converted into a bipartite graph \( G' \) with the desired property, by taking two copies of the vertex set of \( G \) and connecting the two copies according to the edges in \( G \). The existence of the algorithm stated in the lemma follows from the fact that \( G' \) is strongly-explicit.

We thus focus on constructing the graph \( G' \). This is done in two steps: first, we show how to construct a strongly-explicit expander \( G'' \) over \( n \) vertices – this requires a bit of work, since \( n \) can be an arbitrary number, and expanders are usually constructed for special values of \( n \). In the second step, we amplify the spectral gap of \( G'' \) by powering, and set \( G' \) to be the powered graph. We then prove that \( G' \) has the desired sampling property.

**The first step.** The work of [GG81] gives a strongly-explicit expander with constant degree and constant edge expansion for every \( n \) that is a square, so we only need to deal with the case in which \( n \) is not a square. Suppose that \( n = m^2 - k \), where \( m^2 \) is the minimal square larger than \( n \), and observe that \( k \leq 2m - 1 \), which is at most \( \frac{1}{2} \cdot m^2 \) for sufficiently large \( m \). Now, we construct an expander over \( m^2 \) vertices using [GG81], and then merge \( k \) pairs of vertices. In order to maintain the regularity, we add self-loops to all the vertices that were not merged. We set \( G'' \) to be the resulting graph.

It is easy to see that \( G'' \) is a regular graph over \( n \) vertices. Since the merge and the addition of self-loops maintain the degree and the edge expansion of the original expander up to a constant factor, it follows that \( G'' \) is an expander with constant degree and constant edge expansion. Furthermore, it is not hard to see that \( G'' \) is strongly-explicit.

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The second step. Since $G''$ is an expander, and in particular has constant edge expansion, it follows from the Cheeger inequality \cite{Dod84, AM85} that its second-largest normalized eigenvalue (in absolute value) is some constant smaller than 1. Let us denote this normalized eigenvalue by $\lambda$. We note that the degree and the edge expansion of $G''$, as well as $\lambda$, are independent of $n$.

We now construct the graph $G'$ by raising $G''$ to the power $\log \lambda (\sqrt{\alpha \cdot \gamma})$. Observe that $G'$ is a graph over $n$ vertices with degree $d \overset{\text{def}}{=} \text{poly} \left( \frac{1}{\alpha \cdot \gamma} \right)$ and normalized second eigenvalue $\sqrt{\alpha \cdot \gamma}$. It is not hard to see that $G'$ is strongly-explicit.

The sampling property. We prove that $G'$ has the desired sampling property. Let $T$ be a subset of vertices of $G'$. We show that for at least $(1 - \alpha)$ fraction of the vertices $s$ of $G'$ it holds that

$$\frac{|E(s, T)|}{d} - \frac{|T|}{n} \leq \gamma.$$ 

To this end, let

$$S \overset{\text{def}}{=} \left\{ s \in U \middle| \frac{|E(s, T)|}{d} - \frac{|T|}{n} > \gamma \right\}.$$ 

Clearly, it holds that

$$\frac{|E(S, T)|}{d \cdot |S|} - \frac{|T|}{n} > \gamma.$$ 

On the other hand, the expander mixing lemma \cite{AC88} implies that

$$\frac{|E(S, T)|}{d \cdot |S|} - \frac{|T|}{n} \leq \sqrt{\alpha \cdot \gamma} \cdot \sqrt{|T|}.$$ 

By combining the above pair of inequalities, we get

$$\gamma < \sqrt{\alpha \cdot \gamma} \cdot \sqrt{|T|}$$

$$|S| < \alpha \cdot |T| \leq \alpha \cdot n,$$

as required. \hfill \blacksquare

3 \quad \text{LCCs with sub-polynomial query complexity}

In this section, we prove the following theorem on LCCs, which immediately implies Theorem \ref{thm:LCC} from the introduction.

**Theorem 3.1** (Main LCC theorem). For every $r \in (0, 1)$, there exists an explicit infinite family of $\mathbb{F}_2$-linear codes $\{C_n\}$ satisfying:

1. $C_n$ has block length $n$, rate at least $r$, and relative distance at least $1 - r - o(1)$.

2. $C_n$ is locally correctable from $\frac{1-r-o(1)}{2}$ fraction of errors with query complexity $\exp(\sqrt{\log n \cdot \log \log n})$.

3. The alphabet of $C_n$ is a vector space $\Sigma_n$ over $\mathbb{F}_2$, such that $|\Sigma_n| \leq \exp \left( \exp(\sqrt{\log n \cdot \log \log n}) \right)$.
Furthermore, the family \( \{ C_n \} \) has a uniform local corrector that runs in time \( \exp(\sqrt{\log n \cdot \log \log n}) \).

We note that the existence of binary LCCs (Theorem 1.1) also follows from Theorem 3.1. In order to construct the binary LCCs, we concatenate the codes of Theorem 3.1 with any asymptotically good inner binary code that has efficient encoding and decoding algorithms. The local corrector of the binary LCCs will emulate the original local corrector, and whenever the latter queries a symbol, the binary local corrector will emulate this query by decoding the corresponding codeword of the inner code. Since such constructions are standard (see [KSY14]), we do not provide the full details.

The proof of Theorem 3.1 has two steps. In the first step, we give a transformation that amplifies the fraction of errors from which an LCC can be corrected – this step follows the distance amplification of [AL96]. In the second step, we construct a locally-correctable code \( W \) with the desired query complexity but that can only be corrected from a sub-constant fraction of errors. Finally, we construct the code \( C \) by applying the distance amplification to \( W \). Those two steps are formalized in the following pair of lemmas, which are proved in Sections 3.1 and 3.2 respectively.

**Lemma 3.2.** Suppose that there exists a code \( W \) that is locally correctable from \( \tau_W \) fraction of errors with query complexity \( q \), such that:

- \( W \) has rate \( r_W \).
- \( W \) is \( \mathbb{F}_2 \)-linear.

Then, for every \( 0 < \tau < \frac{1}{2} \) and \( 0 < \varepsilon < 1 \), there exists a code \( C \) that is locally correctable from \( \tau \) fraction of errors with query complexity \( q \cdot \text{poly}(1/(\varepsilon \cdot \tau_W)) \), such that:

- \( |C| = |W| \).
- \( C \) has relative distance at least \( 2 \cdot \tau \), and rate at least \( r_W \cdot (1 - 2 \cdot \tau - \varepsilon) \).
- Let \( \Lambda \) denote the alphabet of \( W \). Then, the alphabet of \( C \) is \( \Sigma \defeq \Lambda^p \) for some \( p = \text{poly}(1/(\varepsilon \cdot \tau_W)) \).
- \( C \) is \( \mathbb{F}_2 \)-linear.

Furthermore,

- There is a polynomial time algorithm that computes a bijection from every code \( W \) to the corresponding code \( C \), given \( r_W, \tau_W, r, \varepsilon \) and \( \Lambda \).

- There is an oracle algorithm that when given black box access to the local corrector of any code \( W \), and given also \( r_W, \tau_W, r, \varepsilon, \Lambda \), computes the local corrector of the corresponding code \( C \). The resulting local corrector of \( C \) runs in time that is polynomial in the running time of the local corrector of \( W \) and in \( 1/\tau_W, 1/\varepsilon \) and \( \log(n_W) \) where \( n_W \) is the block length of \( W \).

**Lemma 3.3.** There exists an explicit infinite family of \( \mathbb{F}_2 \)-linear codes \( \{ W_n \} \) satisfying:

1. \( W_n \) has block length \( n \), rate at least \( 1 - \frac{1}{\log n} \), and relative distance at least \( \Omega \left( \sqrt{\frac{\log \log n}{\log^3 n}} \right) \).

2. \( W_n \) is locally correctable from \( \Omega \left( \sqrt{\frac{\log \log n}{\log^3 n}} \right) \) fraction of errors with query complexity \( \exp(\sqrt{\log n \cdot \log \log n}) \).
3. The alphabet of \( W_n \) is a vector space \( \Lambda_n \) over \( \mathbb{F}_2 \), such that \(|\Lambda_n| \leq \exp\left(\exp(\sqrt{\log n \cdot \log \log n})\right)\). Furthermore, the family \( \{W_n\}_n \) has a uniform local corrector that runs in time \( \exp(\sqrt{\log n \cdot \log \log n}) \).

**Proof of Theorem 3.1.** We construct the family \( \{C_n\}_n \) by applying Lemma 3.2 to the family \( \{W_n\}_n \) of Lemma 3.3 with \( \tau_w = \Omega\left(\sqrt{\frac{\log \log n}{\log^3 n}}\right) \), \( \varepsilon = \frac{1}{\log n} \), and

\[
\tau = \frac{1}{2} \cdot \left(1 - \frac{r}{\log n} - \varepsilon\right) = \frac{1}{2} \cdot \left(1 - r - O\left(\frac{1}{\log n}\right)\right).
\]

It is easy to see that \( C_n \) has the required rate, relative distance and alphabet size, and that it can be locally corrected from the required fraction of errors with the required query complexity. The family \( \{C_n\}_n \) is explicit with the required running time due to the first item in the “furthermore” part of Lemma 3.2 and has a uniform local corrector due to the second item of that part. ■

**Remark 3.4.** In Lemma 3.2 above, we chose to assume that \( W \) is \( \mathbb{F}_2 \)-linear for simplicity. More generally, if \( W \) is \( \mathbb{F} \)-linear for any finite field \( \mathbb{F} \), then \( C \) is \( \mathbb{F} \)-linear as well. Furthermore, the lemma also works if \( W \) is not \( \mathbb{F} \)-linear for any field \( \mathbb{F} \), in which case \( C \) is not guaranteed to be \( \mathbb{F} \)-linear for any field \( \mathbb{F} \).

### 3.1 Proof of Lemma 3.2

#### 3.1.1 Overview

Let \( 0 < \tau < \frac{1}{2} \). Our goal is to construct a code \( C \) that can be locally corrected from a fraction of errors at most \( \tau \). The idea of the construction is to combine the LCC \( W \) with a Reed-Solomon code to obtain a code \( C \) that enjoys “the best of both worlds”: both the local correctability of \( W \) and the good error correction capability of Reed-Solomon. We do it in two steps: first, we construct a code \( C' \) which can be corrected from \( \tau \) fraction of random errors. Then, we augment \( C' \) to obtain a code \( C \) that can be corrected from \( \tau \) fraction of adversarial errors.

We first describe the construction of \( C' \). To this end, we describe a bijection from \( W \) to \( C' \). Let \( w \) be a codeword of \( W \). To obtain the codeword \( c' \in C' \) that corresponds to \( w \), we partition \( w \) into blocks of length \( b \) (to be determined later), and encode each block with a Reed-Solomon code \( RS_{b,d} \). We choose the relative distance of \( RS_{b,d} \) to be \( 2 \cdot \tau + \varepsilon \), so its rate is \( 1 - 2 \cdot \tau - \varepsilon \) and the rate of \( C' \) is indeed \( r_w \cdot (1 - 2 \cdot \tau - \varepsilon) \), as required.

We now claim that if one applies to a codeword \( c' \in C' \) a noise that corrupts each coordinate with probability \( \tau \), then the codeword \( c' \) can be recovered from its corrupted version with high probability. To see it, first observe that with high probability, almost all the blocks of \( c' \) have at most \( \tau + \frac{\varepsilon}{2} \) fraction of corrupted coordinates. Let us call those blocks “good blocks”, and observe that the good blocks can be corrected by decoding them to the nearest codeword of \( RS_{b,d} \) (since \( \tau + \frac{\varepsilon}{2} \) is half the relative distance of \( RS_{b,d} \)). Next, observe that if \( b \) is sufficiently large, the fraction of “good blocks” is at least \( 1 - \tau_w \), and hence we can correct the remaining \( \tau_w \) fraction of errors using the decoding algorithm of \( W \). It follows that \( C' \) can be corrected from \( \tau \) fraction of random errors, as we wanted.

Next, we show how to augment \( C' \) to obtain a code \( C \) that is correctable from adversarial errors. This requires two additional ideas. The first idea to apply a permutation that is “pseudorandom” in some sense to the coordinates of \( C' \). The “pseudorandom” permutation is determined by the
edges of an expander graph (see Section 2.4). This step is motivated by the hope that, after the adversary decided which coordinates to corrupt, applying the permutation to the coordinates will make the errors behave pseudorandomly. This will allow the above analysis for the case of random errors to go through.

Of course, on its own, this idea is doomed to fail, since the adversary can take the permutation into account when he chooses where to place the errors. Here the second idea comes into play: after applying the permutation to the coordinates of \( C \)'s, we will increase the alphabet size of the code, packing each block of symbols into a new big symbol. The motivation for this step is that increasing the alphabet size restricts the freedom of the adversary in choosing the pattern of errors. Indeed, we will show that after the alphabet size is increased, applying the permutation to the coordinates of the code makes the errors behave pseudorandomly. This allows us to prove that the code can be decoded from \( \tau \) fraction of errors, as we wanted.

### 3.1.2 The construction of \( C \)

**Choosing the parameters.** Let \( W, r_W, \tau_W, r, \varepsilon, \) and \( \Lambda \) be as in Lemma 3.2. Let \( \{G_n\}_n \) be an infinite family of \((\tau_W, \frac{1}{2} \cdot \varepsilon)\)-samplers as in Theorem 2.7, and let \( d \) be their degree.

Recall that we assumed that \( W \) is \( \mathbb{F}_2 \)-linear, so \( |\Lambda| \) is a power of 2. Let \( \mathbb{F} \) be an extension field of \( \mathbb{F}_2 \), whose size is the minimal power of \( |\Lambda| \) that is at least \( d \). Let \( RS_{b,d} \) be a Reed-Solomon code over \( \mathbb{F} \) with relative distance \( 2 \cdot \tau + \varepsilon \), rate \( 1 - 2 \cdot \tau - \varepsilon \), and block length \( d \).

Let \( n_W \) be the block length of \( W \), and let \( t \) be such that \( |\mathbb{F}| = |\Lambda|^t \). The block length of \( C \) will be \( n = \frac{n_W}{b \cdot t} \), and its alphabet will be \( \Sigma = \mathbb{F}^d \). Here, we assume that \( n_W \) is divisible by \( b \cdot t \). If \( n_W \) is not divisible by \( b \cdot t \), we consider two cases:

- if \( n_W > b \cdot t / \varepsilon \), we increase \( n_W \) to the next multiple of \( b \cdot t \) by padding the codewords of \( W \) with additional zero coordinates. This decreases the rate of \( W \) by at most \( \varepsilon \), which essentially does not affect our results.

- Otherwise, we set \( C \) to be any Reed-Solomon code with blocklength \( n_W \), relative distance \( 2 \cdot \tau \), and rate \( 1 - 2 \cdot \tau \). Observe that such a Reed-Solomon is locally correctable from \( \tau \) fraction of errors with query complexity

\[
  n_W \leq b \cdot t / \varepsilon = \text{poly}(1/(\varepsilon \cdot \tau_W)),
\]

which satisfies our requirements.

**A bijection from \( W \) to \( C \).** We construct the code \( C \) by describing a bijection from \( W \) to \( C \).

Given a codeword \( w \in W \), one obtains the corresponding codeword \( c \in C \) as follows:

- Partition \( w \) into \( n = \frac{n_W}{b \cdot t} \) blocks of length \( b \cdot t \). We view each of those blocks as a vector in \( \mathbb{F}^b \), and encode it via the code \( RS_{b,d} \). Let us denote the resulting string by \( c' \in \mathbb{F}^{n-d} \) and the resulting codewords of \( RS_{b,d} \) by \( B_1, \ldots, B_n \in \mathbb{F}^d \).

- Next, we apply a “pseudorandom” permutation to the coordinates of \( c' \) as follows: Let \( G_n \) be the graph from the infinite family above and let \( U = \{u_1, \ldots, u_n\} \) and \( V = \{v_1, \ldots, v_n\} \) be the left and right vertices of \( G_n \) respectively. For each \( i \in [n] \) and \( j \in [d] \), we write the \( j \)-th symbol of \( B_i \) on the \( j \)-th edge of \( u_i \). Then, we construct new blocks \( S_1, \ldots, S_n \in \mathbb{F}^d \), by setting the \( j \)-th symbol of \( S_i \) to be the symbol written on the \( j \)-th edge of \( v_i \).
Finally, we define the codeword $c$ of $C \subseteq \Sigma^n$ as follows: the $i$-th coordinate $c_i$ is the block $S_i$, reinterpreted as a symbol of the alphabet $\Sigma \eqdef \mathbb{F}_d$. We choose $c$ to be the codeword in $C$ that corresponds to the codeword $w$ in $W$.

This concludes the definition of the bijection. It is not hard to see that this bijection can be computed in polynomial time, and that the code $C$ is $\mathbb{F}_2$-linear. Furthermore, $\Sigma = \mathbb{F}_d = \Lambda^d$ where $d \cdot t \leq d \log d = \text{poly}(1/(\varepsilon \cdot \tau_W))$. The rate of $C$ is

$$\frac{\log |C|}{n \cdot \log |\Sigma|} = \frac{\log |W|}{n \cdot d \cdot \log |\mathbb{F}|} = \frac{r_w \cdot \log |\Lambda^{n_W}|}{n \cdot d \cdot \log |\mathbb{F}|} = \frac{r_w \cdot (b \cdot t) \cdot \frac{1}{b} - 2 \cdot \tau \cdot \varepsilon}{b} \cdot \frac{1}{t}$$

as required. The relative distance of $C$ is at least $2 \cdot \tau$ – although this could be proved directly, it also follows immediately from the fact that $C$ is locally correctable from $\tau$ fraction of errors, which is proved in the next section.

### 3.1.3 Local correctability

In this section, we complete the proof of Lemma 3.2 by proving that $C$ is locally correctable from $\tau$ fraction of errors with query complexity $\text{poly}(d) \cdot q$. To this end, we describe a local corrector $A$. The algorithm $A$ is based on the following algorithm $A_0$, which locally corrects coordinates of $W$ from a corrupted codeword of $C$.

**Lemma 3.5.** There exists an algorithm $A_0$ that satisfies the following requirements:

- **Input:** $A_0$ takes as input a coordinate $i \in [n_W]$, and also gets oracle access to a string $z \in \Sigma^n$ that is $\tau$-close to a codeword $c \in C$.

- **Output:** Let $w^c$ be the codeword of $W$ from which $c$ was generated. Then, $A_0$ outputs $w^z_i$ with probability at least $1 - \frac{1}{3 \cdot b \cdot c}$.

- **Query complexity:** $A_0$ makes $\text{poly}(d) \cdot q$ queries to the oracle $z$.

Before proving Lemma 3.5, we show how to construct the algorithm $A$ given the algorithm $A_0$. Suppose that the algorithm $A$ is given oracle access to a string $z$ that is $\tau$-close to a codeword $c \in C$, and a coordinate $i \in [n]$. The algorithm is required to decode $c_i$. Let $w^c \in \Lambda^{n_W}$ be the codeword of $W$ from which $c$ was generated, and let $B^c_{i1}, \ldots, B^c_{ii}$ and $S^c_{i1}, \ldots, S^c_{in}$ be the corresponding blocks.

In order to decode $c_i$, the algorithm $A$ should decode each of the symbols in the block $S^c_i \in \mathbb{F}_d$. Let $u_{i1}, \ldots, u_{i4}$ be the neighbors of $v_i$ in the graph $G$. Each symbol of the block $S^c_i$ belongs to one of the blocks $B^c_{i1}, \ldots, B^c_{i4}$, and therefore it suffices to retrieve the latter blocks. Now, each block $B^c_{ij}$ is the encoding via $RS_{b,d}$ of $b \cdot t$ symbols of $w^c$ (in the alphabet $\Lambda$). The algorithm $A$ invokes the algorithm $A_0$ to decode each of those $b \cdot t$ symbols of $w^c$, for each of the blocks $B^c_{ij}, \ldots, B^c_{i4}$. By the
union bound, the algorithm \( A_0 \) decodes all those \( b \cdot t \cdot d \) symbols of \( w^c \) correctly with probability at least \( 1 - b \cdot t \cdot d \cdot \frac{1}{3b \cdot t \cdot d} = \frac{2}{3} \). Whenever that happens, the algorithm \( A \) retrieves the blocks \( B_{j_1}^c, \ldots, B_{j_d}^c \) correctly, and therefore computes the block \( S_i^c \) correctly. This concludes the construction of the algorithm \( A \). Note that the query complexity of \( A \) is larger than that of \( A_0 \) by a factor of at most \( b \cdot t \cdot d \), and hence it is at most \( \text{poly}(d) \cdot q \). It remains to prove Lemma 8.5.

**Proof of Lemma 8.5.** Let \( A_w \) be the local corrector of the code \( W \). By amplification, we may assume that \( A_w \) errs with probability at most \( \frac{1}{3b \cdot t \cdot d} \), and this incurs a factor of at most \( \text{poly}(d) \) to its query complexity.

Suppose that the algorithm \( A_0 \) is invoked on a string \( z \in \Sigma^n \) and a coordinate \( i \in [n_w] \). The algorithm \( A_0 \) invokes the algorithm \( A_w \) to retrieve the coordinate \( i \), and emulates \( A_w \) in the natural way: Recall that \( A_w \) expects to be given access to a corrupted codeword of \( W \), and makes queries to it. Whenever \( A_w \) makes a query to a coordinate \( i_w \in [n_w] \), the algorithm \( A_0 \) performs the following steps:

1. \( A_0 \) finds the block \( B_l \) to which the coordinate \( i_w \) belongs. Formally, \( l \defeq \lceil i_w / (b \cdot t) \rceil \).
2. \( A_0 \) finds the neighbors of the vertex \( u_l \) in \( G_n \). Let us denote those vertices by \( v_{j_1}, \ldots, v_{j_d} \).
3. \( A_0 \) queries the coordinates \( j_1, \ldots, j_d \), thus obtaining the blocks \( S_{j_1}, \ldots, S_{j_d} \).
4. \( A_0 \) reconstructs the block \( B_l \) by reversing the permutation of \( G_n \) on \( S_{j_1}, \ldots, S_{j_d} \).
5. \( A_0 \) attempts to decode \( B_l \) by applying an efficient decoding algorithm of Reed-Solomon.
6. Suppose that the decoding succeeded and returned a codeword of \( RS_{b,d} \) that is \((\tau + \frac{\varepsilon}{2})\)-close to \( B_l \). Then, \( A_0 \) retrieves the value of the \( i_w \)-th coordinate of \( w^c \) from the latter codeword, and feeds it to \( A_w \) as an answer to its query.
7. Otherwise, \( A_0 \) feeds 0 as an answer to the query of \( A_w \).

When the algorithm \( A_w \) finishes running, the algorithm \( A_0 \) finishes and returns the output of \( A_w \). It is not hard to see that the query complexity of \( A_0 \) is at most \( d \) times the query complexity of \( A_w \), and hence it is at most \( \text{poly}(d) \cdot q \). It remains to show that \( A_0 \) succeeds in decoding from \( \tau \) fraction of errors with probability at least \( 1 - \frac{1}{3b \cdot t \cdot d} \).

Let \( z \in \Sigma^n \) be a string that is \( \tau \)-close to a codeword \( c \in C \). Let \( w^c \in \Lambda^{nw} \) be the codeword of \( W \) from which \( c \) was generated, and let \( B_{j_1}^c, \ldots, B_{j_d}^c \) and \( S_{j_1}^c, \ldots, S_{j_d}^c \) be the corresponding blocks. We also use the following definitions:

1. Let \( S_1^c, \ldots, S_n^c \in \mathbb{F}^d \) be the blocks that correspond to the symbols of \( z \).
2. Let \( B_1^c, \ldots, B_n^c \) be the blocks that are obtained from \( S_1^c, \ldots, S_n^c \) by reversing the permutation.
3. Define blocks \( B_1'^c, \ldots, B_n'^c \) as follows: if \( B_i^c \) is \((\tau + \frac{\varepsilon}{2})\)-close to \( RS_{b,d} \), then \( B_i'^c \) is the nearest codeword of \( RS_{b,d} \). Otherwise, \( B_i'^c \) is the all-zeroes block.
4. Let \( w^z \in \Lambda^{nw} \) be the string that is obtained by extracting the coordinates of \( w \) from each of the codewords \( B_1'^c, \ldots, B_n'^c \).
It is easy to see that $A_0$ emulates the action of $A_w$ on $w^z$. Therefore, if we prove that $w^z$ is $\tau_w$-close to $w^c$, we will be done. In order to do so, it suffices to prove that for at least $1 - \tau_w$ fraction of the blocks $B_i^f$, it holds that $B_i^c$ is $(\tau + \frac{c}{2})$-close to $B_i^c$.

To this end, let $J$ be the set of coordinates on which $z$ and $c$ differ. In other words, for every $j \in J$ it holds that $S_j^z \neq S_j^c$. By assumption, $|J| \leq \tau \cdot n$. Now, observe that since $G_n$ is a $(\tau_w, \frac{1}{2} \cdot \varepsilon)$-sampler, it holds that for at least $(1 - \tau_w)$ fraction of the vertices $u_i$ of $G_n$, there are at most $(\tau + \frac{c}{2}) \cdot d$ edges between $u_i$ and $J$. For each such $u_i$, it holds that $B_{u_i}^c$ is $(\tau + \frac{c}{2})$-close to $B_{u_i}^c$, and this concludes the proof. \hfill \Box

It can be verified that the local correctors $A_0$ and $A$ can be implemented efficiently with black box access to $A_w$, as required by the second item in the “furthermore” part of the lemma.

### 3.2 Proof of Lemma 3.3

In this section we prove Lemma 3.3, restated below.

**Lemma 3.3.** There exists an explicit infinite family of $\mathbb{F}_2$-linear codes $\{W_n\}_n$ satisfying:

1. $W_n$ has block length $n$, rate at least $1 - \frac{1}{\log n}$, and relative distance at least $\Omega\left(\sqrt{\log \log n / \log^2 n}\right)$.

2. $W_n$ is locally correctable from $\Omega\left(\sqrt{\log \log n / \log^2 n}\right)$ fraction of errors with query complexity $\exp(\sqrt{\log n \cdot \log \log n})$.

3. The alphabet of $W_n$ is a vector space $\Lambda_n$ over $\mathbb{F}_2$, such that $|\Lambda_n| \leq \exp(\sqrt{\log n \cdot \log \log n})$.

Furthermore, the family $\{W_n\}_n$ has a uniform local corrector that runs in time $\exp(\sqrt{\log n \cdot \log \log n})$.

For the proof of Lemma 3.3 we use the multiplicity codes of [KSY14], in a specialized subconstant relative distance regime.

**Lemma 3.6 ([KSY14] Lemma 3.5).** Let $\mathbb{F}$ be any finite field. Let $s, d, m$ be positive integers. Let $M$ be the multiplicity code of order $s$ evaluations of degree $d$ polynomials in $m$ variables over $\mathbb{F}$. Then $M$ has block length $|\mathbb{F}|^m$, relative distance at least $\delta = 1 - \frac{d}{s\cdot|\mathbb{F}|}$ and rate $\frac{(\frac{d+m}{m})}{\binom{m+s-1}{m} |\mathbb{F}|^m}$, which is at least

$$\left(\frac{s}{m+s}\right)^m \cdot \left(\frac{d}{s\cdot|\mathbb{F}|}\right)^m \geq \left(1 - \frac{m^2}{s}\right) \cdot (1 - \delta)^m.$$  

The alphabet of $C$ is $\mathbb{F}^{(m+s-1)}$, and $C$ is $\mathbb{F}$-linear. Furthermore, there is poly $\left(\mathbb{F}^m, (m+s-1)\right)$ time algorithm that computes an encoding map of $M$ given $s, d, m$, and $\mathbb{F}$.

**Lemma 3.7 ([KSY14] Lemma 3.6).** Let $M$ be the multiplicity code as above. Let $\delta = 1 - \frac{d}{s\cdot|\mathbb{F}|}$ be a lower bound for the relative distance of $M$. Suppose $|\mathbb{F}| \geq \max\{10 \cdot m, \frac{d+6}{s} \cdot (s+1)\}$. Then $M$ is locally correctable from $\delta/10$ fraction of errors with query complexity $O(s^m \cdot |\mathbb{F}|)$.

As discussed in Section 4.3 of [KSY14], this local corrector can be implemented to have running time $\text{poly}(|\mathbb{F}|, s^m)$ over fields of constant characteristic. In fact, [Kop14] shows that the query complexity and running time for local correcting multiplicity codes can be further reduced to $|\mathbb{F}| \cdot O\left(\left(\frac{1}{s}\right)^m\right)$ queries, but this does not lead to any noticeable improvement for our setting.

We now prove Lemma 3.3.
Proof. Let $n \in \mathbb{N}$ be a codeword length. We set the code $W_n$ to be a multiplicity code with the following parameters. We choose $F$ to be a field of size $2^{\sqrt{\log n \cdot \log \log n}}$, and choose $m = \sqrt{\frac{\log n}{\log \log n}}$. Note that indeed $|F|^m = n$. We choose $s = 2 \cdot m^2 \cdot \log n$. Let $\delta = \frac{1}{2m \cdot \log n}$ (this will be a lower bound on the relative distance of the code) and choose the degree of the polynomials to be $d = s \cdot |F| \cdot (1 - \delta)$.

It can be verified that the relative distance of the code is at least $\delta \geq \Omega \left( \sqrt{\frac{\log \log n}{\log 3n}} \right)$. The rate of the code is at least
\[
\left( 1 - \frac{m^2}{s} \right) \cdot (1 - \delta)^m \geq \left( 1 - \frac{1}{2 \cdot \log n} \right) \left( 1 - \frac{1}{2 \cdot m \cdot \log n} \right)^m \geq 1 - \frac{1}{\log n}.
\]
as required. The alphabet size is
\[
|F|^{m+1} \leq \exp \left( \sqrt{\log n \cdot \log \log n \cdot s^m} \right) = \exp \left( \sqrt{\log n \cdot \log \log n} \cdot \left( \log^2 n \right) \cdot \frac{\log n}{\log \log n} \right) = \exp \left( \exp \left( \sqrt{\log n \cdot \log \log n} \right) \right).
\]
Moreover, the alphabet is a vector space over $F$ and hence in particular over $F_2$ (since we chose the size of $F$ to be a power of 2). The code $W_n$ is $F$-linear and in particular $F_2$-linear.

By Lemma 3.7, $W_n$ is locally correctable from $\frac{1}{10} \cdot \delta \geq \Omega \left( \sqrt{\frac{\log \log n}{\log n}} \right)$ fraction of errors with query complexity
\[
O(s^m \cdot |F|) \leq O \left( \frac{\log^2 n}{\log \log n} \right)^{\frac{\log n}{\log \log n}} \cdot 2^{\sqrt{\log n \cdot \log \log n}} = 2 \cdot O \left( \sqrt{\log n \cdot \log \log n} \right),
\]
as required. Finally, the fact that the family $\{W_n\}_n$ is explicit follows from the “furthermore” part of Lemma 3.6 and the fact that it has an efficient uniform local corrector with the required running time follows from the discussion after Lemma 3.7.

3.3 LDCs

As remarked earlier, by choosing a systematic encoding map, linear LCCs automatically give LDCs with the same rate, relative distance, and query complexity. The running time of the local decoding algorithm will be essentially the same as the running time of the local correction algorithm, provided that the systematic encoding map can be computed efficiently. Using the fact that multiplicity codes have an efficiently computable systematic encoding map [Kop12], it is easy to check that the codes we construct above also have an efficiently computable systematic encoding map. Thus we get LDCs with the same parameters as our LCCs.

4 LTCs with sub-polynomial query complexity

In this section, we prove the following theorem on LTCs, which immediately implies Theorem [14] from the introduction.
Theorem 4.1 (Main LTC theorem). For every $r \in (0, 1)$, there exists an explicit infinite family of $\mathbb{F}_2$-linear codes $\{C_n\}_n$ satisfying:

1. $C_n$ has block length $n$, rate at least $r$, and relative distance at least $1 - r - o(1)$.
2. $C_n$ is locally testable with query complexity $\exp(\sqrt{\log n \cdot \log \log n})$.
3. The alphabet of $C_n$ is a vector space $\Sigma_n$ over $\mathbb{F}_2$, such that $|\Sigma_n| \leq \exp(\exp(\sqrt{\log n \cdot \log \log n}))$.

Furthermore, the family $\{C_n\}_n$ has a uniform local tester that runs in time $\exp(\sqrt{\log n \cdot \log \log n})$.

We note that the existence of binary LTCs (Theorem 1.2) also follows from Theorem 4.1: In order to construct the binary LTCs, we concatenate the codes of Theorem 4.1 with any asymptotically good inner binary code that has efficient encoding and decoding algorithms. The local tester of the binary LTCs will emulate the original local tester, and whenever the latter queries a symbol, the binary local tester will emulate this query by reading the corresponding codeword of the inner code. If this string is not a legal codeword, the binary tester will reject, and otherwise it will decode the symbol and feed it to the original tester. Since such constructions are standard, we do not provide the full details.

The proof of Theorem 4.1 has two steps. In the first step, we give a transformation that amplifies the relative distance of an LTC – this step follows the distance amplification of [AL96]. In the second step, we construct a locally-testable code $W_n$ with the desired query complexity but that has sub-constant relative distance. Finally, we construct the code $C_n$ by applying the distance amplification to $W_n$. Those two steps are formalized in the following pair of lemmas, which are proved in Sections 4.2 and 4.3 respectively.

Lemma 4.2. Suppose that there exists a code $W$ with relative distance $\delta_W$ that is locally testable with query complexity $q$ such that:

- $W$ has rate $r_W$.
- $W$ is $\mathbb{F}_2$-linear.

Then, for every $0 < \delta, \varepsilon < 1$, there exists a code $C$ with relative distance at least $\delta$ that is locally testable with query complexity $q \cdot \text{poly}(1/(\varepsilon \cdot \delta_W))$, such that:

- $|C| = |W|$.
- $C$ has rate at least $r_W \cdot (1 - \delta - \varepsilon)$.
- Let $\Lambda$ denote the alphabet of $W$. Then, the alphabet of $C$ is $\Sigma \overset{\text{def}}{=} \Lambda^p$ for some $p = \text{poly}(1/(\varepsilon \cdot \delta_W))$.
- $C$ is $\mathbb{F}_2$-linear.

Furthermore,

- There is a polynomial time algorithm that computes a bijection from every code $W$ to the corresponding code $C$, given $r_W$, $\delta_W$, $r$, $\varepsilon$ and $\Lambda$. 

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There is an oracle algorithm that when given black box access to the local tester of any code \( W \), and given also \( \tau_W, \delta_W, r, \varepsilon, \Lambda, \) and the block length of \( W \), computes the local tester of the corresponding code \( C \). The resulting local tester of \( C \) runs in time that is polynomial in the running time of the local tester of \( W \) and in \( 1/\delta_W, 1/\varepsilon \) and \( \log(\ell_W) \) where \( \ell_W \) is the block length of \( W \).

**Lemma 4.3.** There exists an explicit infinite family of \( \mathbb{F}_2 \)-linear codes \( \{ W_n \}_n \) satisfying:

1. \( W_n \) has block length \( n \), rate at least \( 1 - \frac{1}{\log n} \), and relative distance at least \( \exp(-\sqrt{\log n \cdot \log \log n}) \).

2. \( W_n \) is locally testable with query complexity \( \exp(\sqrt{\log n \cdot \log \log n}) \).

3. The alphabet of \( W_n \) is a vector space \( \Lambda_n \) over \( \mathbb{F}_2 \), such that \( |\Lambda_n| \leq \exp\left(\sqrt{\log n \cdot \log \log n}\right) \).

Furthermore, the family \( \{ W_n \}_n \) has a uniform local tester that runs in time \( \exp(\sqrt{\log n \cdot \log \log n}) \).

**Proof of Theorem 4.1.** We construct the family \( \{ C_n \}_n \) by applying Lemma 4.2 to the family \( \{ W_n \}_n \) of Lemma 4.3 with \( \delta_W = 2^{-O(\sqrt{\log n \cdot \log \log n})} \), \( \varepsilon = \frac{1}{\log n} \) and

\[
\delta = 1 - \frac{r}{1 - \frac{1}{\log n}} - \varepsilon = 1 - r - O\left(\frac{1}{\log n}\right).
\]

It is easy to see that \( C_n \) has the required rate, relative distance and alphabet size, and that it can be locally tested with the required query complexity. The family \( \{ C_n \}_n \) is explicit due to the first item in the “furthermore” part of Lemma 4.2 and has a uniform local corrector with the required running time due to the second item of that part. \( \blacksquare \)

**Remark 4.4.** In Lemma 4.2 above, as in Lemma 3.2, we chose to assume that \( W \) is \( \mathbb{F}_2 \)-linear for simplicity. More generally, if \( W \) is \( \mathbb{F} \)-linear for any finite field \( \mathbb{F} \), then \( C \) is \( \mathbb{F} \)-linear as well. Furthermore, the lemma also works if \( W \) is not \( \mathbb{F} \)-linear for any field \( \mathbb{F} \), in which case \( C \) is not guaranteed to be \( \mathbb{F} \)-linear for any field \( \mathbb{F} \).

### 4.1 Proof of Lemma 4.2

Our construction of the LTC \( C \) is the same as the construction of the LCCs of Section 3.1 with \( \tau_W \) and \( \tau \) replaced by \( \delta_W/2 \) and \( \delta/2 \) respectively. Our LTCs have the required rate, relative distance and alphabet size due to the same considerations as before\(^5\).

It remains to prove that \( C \) is locally testable with query complexity \( q \cdot \text{poly}(1/(\varepsilon \cdot \delta_W)) \). To this end, we describe a local tester \( A \). In what follows, we use the notation of Section 3.1.2.

Let \( A_W \) be the local tester of \( W \). When given oracle access to a purported codeword \( z \in \Sigma^n \), the local tester \( A \) emulates the action of \( A_W \) in the natural way: Recall that \( A_W \) expects to be given access to a purported codeword of \( W \), and makes queries to it. Whenever \( A_W \) makes a query to a coordinate \( j \in [\ell_W] \), the algorithm \( A \) performs the following steps:

1. \( A \) finds the block \( B_t \) to which the coordinate \( j \) belongs. Formally, \( t \overset{\text{def}}{=} \lfloor j/(b \cdot t) \rfloor \).

\(^5\)In particular, the lower bound on the relative distance of our LTC \( C \) follows from the lower bound on the relative distance given in Lemma 3.2 using the fact that our LTC \( W \) has a (trivial, inefficient) \( \ell_W \) query local corrector from \( \delta_W/2 \) fraction errors. Again, this lower bound on the distance could have been argued directly, without talking about locality.
2. A finds the neighbors of the vertex \( u_l \) in \( G_n \). Let us denote those vertices by \( v_{j_1}, \ldots, v_{j_d} \).

3. A queries the coordinates \( j_1, \ldots, j_d \), thus obtaining the blocks \( S_{j_1}, \ldots, S_{j_d} \).

4. A reconstructs the block \( B_l \) by reversing the permutation of \( G_n \) on \( S_{j_1}, \ldots, S_{j_d} \).

5. If \( B_l \) is not a codeword of \( RS_{b,d} \), the local tester \( A \) rejects.

6. Otherwise, \( A \) retrieves the value of the \( j \)-th coordinate of \( w \) from \( B_l \), and feeds it to \( A_w \) as an answer to its query.

If \( A_w \) finishes running, then \( A \) accepts if and only if \( A_w \) accepts.

It is easy to see that the query complexity of \( A \) is \( d \cdot q \). It is also not hard to see that if \( z \) is a legal codeword of \( C \), then \( A \) accepts with probability 1. It remains to show that if \( z \) is not a codeword of \( C \) then \( A \) rejects with probability at least \( \text{dist}(z, C) \). To this end, it suffices to prove that \( A \) rejects with probability at least \( \frac{1}{\text{poly}(d)} \cdot \text{dist}(z, C) \) – as explained in Section 2.3, this rejection probability can be amplified to \( \text{dist}(z, C) \) while increasing the query complexity by a factor of \( \text{poly}(d) \), which is acceptable. We use the following definitions:

1. Let \( S_1^z, \ldots, S_n^z \in \mathbb{F}^d \) be the blocks that correspond to the symbols of \( z \).

2. Let \( B_1^z, \ldots, B_n^z \in \mathbb{F}^d \) be the blocks that are obtained from \( S_1^z, \ldots, S_n^z \) by reversing the permutation.

3. Let \( w^z \in (\Lambda \cup \{?\})^{nw} \) be the string that is obtained from the blocks \( B_1^z, \ldots, B_n^z \) as follows: for each block \( B_i^z \) that is a legal codeword of \( RS_{b,d} \), we extract from \( B_i^z \) the corresponding coordinates of \( w^z \) in the natural way. For each block \( B_i^z \) that is not a legal codeword of \( RS_{b,d} \), we set the corresponding coordinates of \( w^z \) to be “?”.

We would like to lower bound the probability that \( A \) rejects \( z \) in terms of the probability that \( A_w \) rejects \( w^z \). However, there is a small technical problem: \( A_w \) is defined as acting on strings in \( \Lambda^{nw} \), and not on strings in \( (\Lambda \cup \{?\})^{nw} \). To deal with this technicality, we define an algorithm \( A'_w \) that, when given access to a string \( y \in (\Lambda \cup \{?\})^{nw} \), emulates \( A_w \) on \( y \), but rejects whenever a query is answered with “?”.

We use the following proposition, whose proof we defer to Section 4.1.1.

**Proposition 4.5.** \( A'_w \) rejects a string \( y \in (\Lambda \cup \{?\})^{nw} \) with probability at least

\[
\frac{1}{2} \cdot \min \{ \text{dist}(y, W), \delta_w \}. 
\]

Now, it is not hard to see that when \( A \) is invoked on \( z \), it emulates the action of \( A'_w \) on \( w^z \). To finish the proof, note that since each coordinate in \( W \) has at most \( d \) coordinates of \( C \) that depend on it, it holds that

\[
\text{dist}(z, C) \cdot n \leq d \cdot \text{dist}(w^z, W) \cdot n_w
\]

and therefore

\[
\text{dist}(w^z, W) \geq \frac{n}{n_w} \cdot \frac{1}{d} \cdot \text{dist}(z, C) \geq \frac{1}{b \cdot t \cdot d} \cdot \text{dist}(z, C).
\]

It thus follows that \( A \) rejects \( z \) with probability at least

\[
\frac{1}{2} \cdot \min \{ \text{dist}(w^z, W), \delta_w \} \geq \frac{1}{\text{poly}(d)} \cdot \text{dist}(z, C),
\]

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as required.

It is not hard to see that the local tester $A$ can be implemented efficiently with black box access to $A_W$, as required by the second item in the “furthermore” part of the lemma.

### 4.1.1 Proof of Proposition 4.5

We use the following result.

**Claim 4.6.** Let $I \subseteq [n_W]$ be a set of coordinates. The algorithm $A_W$ queries some coordinate in $I$ with probability at least

$$\min \left\{ \frac{|I|}{n_W}, \frac{1}{2} \cdot \delta_W \right\}.$$ 

Note that this claim only makes sense since we assumed that $A_W$ makes non-adaptive queries (we assumed it in Definition 2.5). Without this assumption, the probability that $A_W$ queries some coordinate in $I$ would have depended on the tested string.

**Proof.** It suffices to prove that for every $I \subseteq [n_W]$ such that $\frac{|I|}{n_W} \leq \frac{1}{2} \cdot \delta_W$, the algorithm $A_W$ queries some coordinate in $I$ with probability at least $\frac{|I|}{n_W}$. Let $I$ be such a set, and let $s \in \Lambda^{n_W}$ be an arbitrary string that contains non-zero values inside $I$, and contains 0 everywhere outside $I$. Clearly,

$$\text{dist}(s, W) = \frac{|I|}{n_W},$$

and therefore $A_W$ rejects $s$ with probability at least $\frac{|I|}{n_W}$. On the other hand, $A_W$ can only reject $s$ if it queries some coordinate in $I$, since otherwise it cannot distinguish between $s$ and the all-zeros codeword. It follows that $A_W$ queries some coordinate in $I$ with probability at least $\frac{|I|}{n_W}$, as required.

We turn to proving Proposition 4.5. Let

$$E \overset{\text{def}}{=} \{ i : y_i = ? \}$$

be the set of erasures in $y$. We consider two cases:

- **$E$ is “large”:** Suppose that $\frac{|E|}{n_W} \geq \frac{1}{2} \cdot \text{dist}(y, W)$. In this case, it holds by Claim 4.6 that $A_W$ queries some coordinate in $E$ with probability at least

$$\frac{1}{2} \cdot \min \{ \text{dist}(y, W), \delta_W \}.$$ 

Since $A'_W$ rejects $y$ whenever $A_W$ queries some coordinate in $E$, the proposition follows.

- **$E$ is “small”:** Suppose that $\frac{|E|}{n_W} \leq \frac{1}{2} \cdot \text{dist}(y, W)$. Let $y_0 \in \Lambda^{n_W}$ be an arbitrary string that agrees with $y$ outside $E$. Clearly,

$$\text{dist}(y, W) \leq \text{dist}(y_0, W) + \frac{|E|}{n_W}.$$
so \( \text{dist}(y_0, W) \geq \frac{1}{2} \cdot \text{dist}(y, W) \). Let \( \mathcal{E} \) denote the event that \( A_w \) queries some coordinate in \( E \). We have that

\[
\Pr[A_w \text{ rejects } y] = \Pr[\mathcal{E}] \cdot \Pr[A_w \text{ rejects } y|\mathcal{E}] + \Pr[\neg \mathcal{E}] \cdot \Pr[A_w \text{ rejects } y|\neg \mathcal{E}]
\]

\[
= \Pr[\mathcal{E}] \cdot 1 + \Pr[\neg \mathcal{E}] \cdot \Pr[A_w \text{ rejects } y_0|\neg \mathcal{E}]
\]

\[
\geq \Pr[\mathcal{E}] \cdot \Pr[A_w \text{ rejects } y_0|\mathcal{E}] + \Pr[\neg \mathcal{E}] \cdot \Pr[A_w \text{ rejects } y_0|\neg \mathcal{E}]
\]

\[
= \Pr[A_w \text{ rejects } y_0]
\]

\[
\geq \text{dist}(y_0, W)
\]

\[
\geq \frac{1}{2} \cdot \text{dist}(y, W),
\]

as required.

This concludes the proof.

### 4.2 Proof of Lemma 4.3

In this section, we prove Lemma 4.3, restated below.

**Lemma 4.3.** There exists an explicit infinite family of \( \mathbb{F}_2 \)-linear codes \( \{W_n\}_n \) satisfying:

1. \( W_n \) has block length \( n \), rate at least \( 1-\frac{1}{\log n} \), and relative distance at least \( \exp(-\sqrt{\log n \cdot \log \log n}) \).

2. \( W_n \) is locally testable with query complexity \( \exp(\sqrt{\log n \cdot \log \log n}) \).

3. The alphabet of \( W_n \) is a vector space \( \Lambda_n \) over \( \mathbb{F}_2 \), such that \( |\Lambda_n| \leq \exp(\sqrt{\log n \cdot \log \log n}) \).

Furthermore, the family \( \{W_n\}_n \) has a uniform local tester that runs in time \( \exp(\sqrt{\log n \cdot \log \log n}) \).

For the proof of Lemma 4.3 we use the tensor product codes instantiated in the sub-constant relative distance regime. The use of tensor products to construct LTCs was initiated by [BS06], and was studied further in [Vid11]. Our construction is based on a result of [Vid11].

We start with some definitions. Let \( \mathbb{F} \) be a finite field. For a pair of vectors \( h_1 \in \mathbb{F}^\ell_1 \) and \( h_2 \in \mathbb{F}^\ell_2 \) their **tensor product** \( h_1 \otimes h_2 \) denotes the matrix \( M \in \mathbb{F}^{\ell_1 \times \ell_2} \) with entries

\[
M_{(i_1,i_2)} = (h_1)_{i_1} \cdot (h_2)_{i_2}
\]

for every \( i_1 \in [\ell_1] \) and \( i_2 \in [\ell_2] \). For a pair of linear codes \( H_1 \subseteq \mathbb{F}^\ell_1 \) and \( H_2 \subseteq \mathbb{F}^\ell_2 \) their **tensor product code** \( H_1 \otimes H_2 \) is defined to be the linear subspace spanned by all matrices of the form \( h_1 \otimes h_2 \) where \( h_1 \in H_1 \) and \( h_2 \in H_2 \). For a linear code \( H \), let \( H^1 = H \) and \( H^m = H^{m-1} \otimes H \).

The following are some useful facts regarding tensor product codes (see e.g. [DSW06]).

**Fact 4.7.** Let \( H_1 \subseteq \mathbb{F}^\ell_1 \) and \( H_2 \subseteq \mathbb{F}^\ell_2 \) be linear codes of rates \( r_1, r_2 \) and relative distances \( \delta_1, \delta_2 \) respectively. Then \( H_1 \otimes H_2 \subseteq \mathbb{F}^{\ell_1 \times \ell_2} \) is a linear code of rate \( r_1 \cdot r_2 \) and relative distance \( \delta_1 \cdot \delta_2 \). In particular, if \( H \subseteq \mathbb{F}^\ell \) is a linear code of rate \( r \) and relative distance \( \delta \) then \( H^m \subseteq \mathbb{F}^{\ell m} \) is a linear code of rate \( r^m \) and relative distance \( \delta^m \).

We use the following theorem that is given as Corollary 3.6 in [Vid11].

**Theorem 4.8** (Immediate corollary of [Vid11], Thm. 3.1). Let \( H \subseteq \mathbb{F}^\ell \) be a linear code with relative distance \( \delta \). Then for every \( m \geq 3 \), the code \( H^m \subseteq \mathbb{F}^{\ell m} \) is locally testable with query complexity

\[
\ell^2 \cdot \text{poly}(m)/\delta^{2m}.
\]
For the proof of Lemma 4.3, we instantiate Theorem 4.8 with the tensor product of Reed-Solomon codes.

**Proof of Lemma 4.3** Fix a codeword length \( n \in \mathbb{N} \). The code \( W_n \) is defined as follows. Let \( F \overset{\text{def}}{=} F_{2^{\sqrt{\log n \cdot \log \log n}}} \), and let \( m \overset{\text{def}}{=} \sqrt{\log n \cdot \log \log n} \). Let \( R \) be a Reed-Solomon code over \( F \) with block length \( n^{1/m} \), rate \( r \overset{\text{def}}{=} \left(1 - \frac{1}{\log n}\right)^{1/m} \) and relative distance \( 1 - r \). Note that indeed the block length is at most \( |F| \), which is required for the existence of such codes. Finally, let \( W_n = R^m \).

From the properties of tensor codes we have that \( W_n \) is a linear code over \( F \) with block length \( (n^{1/m})^m = n \), rate \( r^m = 1 - \frac{1}{\log n} \), and relative distance

\[
(1 - r)^m = \left(1 - \left(1 - \frac{1}{\log n}\right)^{1/m}\right)^m \\
\geq \left(1 - \left(1 - \frac{1}{4 \cdot m \cdot \log n}\right)\right)^m \\
= \left(\frac{4 \cdot m \cdot \log n}{1}ight)^m \\
= 2^{-O(m \cdot (\log m + \log \log n))} \\
= 2^{-O(\sqrt{\log n \cdot \log \log n})},
\]

as required. The fact that \( W_n \) can be encoded in time \( \text{poly}(n) \) follows from standard properties of tensor product codes (see e.g. [Sud01, Lecture 6]).

Finally, by Theorem 4.8 we have that \( W_n \) is locally testable with query complexity at most

\[
n^{2/m} \cdot \text{poly}(m) \cdot \left(\frac{1}{4 \cdot m \cdot \log n}\right)^{-2m} = 2^{O(\sqrt{\log n \cdot \log \log n})},
\]

as required. The fact that the family \( \{W_n\}_n \) has a uniform local tester with the required running time follows immediately from the proof of [Vid11]. ■

5 Open Questions

We conclude with some open questions.

- In this work we found that LCCs and LTCs with sub-constant relative distance can be useful. Are there better LCCs and LTCs in the sub-constant relative distance regime?
- LCCs and LTCs often come together with PCPs. Can we construct constant-rate PCPs with sub-polynomial query complexity?
- Are there applications of our LCCs and LTCs to complexity theory?

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\[\text{We chose Reed-Solomon codes for convenience, but any high-rate codes with reasonable distance will do.}\]
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