

# Locally Correctable and Testable Codes Approaching the Singleton Bound

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## Abstract

Locally-correctable codes (LCCs) and locally-testable codes (LTCs) are codes that admit local algorithms for decoding and testing respectively. The local algorithms are randomized algorithms that make only a small number of queries to their input. LCCs and LTCs are both interesting in their own right, and have important applications in complexity theory.

It is a well-known question what are the best rate and distance that such LCCs and LTCs can achieve. When discussing LCCs and LTCs that use a constant number of queries (which is the most common setting), it is known that LCCs can not achieve a constant rate, and it is believed that the same is true for LTCs. However, it has recently been discovered that the situation is radically different when using  $n^\beta$  queries ( $\beta > 0$ ): it turns out that there are both LCCs and LTCs that achieve *any* constant rate, while using  $n^\beta$  queries.

In this work, we observe that in fact, LCCs and LTCs with  $n^\beta$  queries can, for any rate, approach the best possible relative distance. More specifically, recall that, by the Singleton bound, an error-correcting code of rate  $r$  can have relative distance of at most  $1 - r$ . We construct LCCs and LTCs that, for every  $r > 0$  and  $\varepsilon > 0$ , have rate  $r$  and relative distance  $1 - r - \varepsilon$ , where the alphabet size is a constant that depends on  $\varepsilon$ . By applying concatenation to those codes, we obtain binary LCCs and LTCs with  $n^\beta$  queries that achieve the Zyablov bound, which constitutes the best known parameters for (explicit) binary codes.

## 1 Introduction

Locally-correctable codes [BFLS91, STV01, KT00] and locally-testable codes [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 an 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 an algorithm that, when given a string  $z$ , decides whether  $w$  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 (e.g.

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[STV01]), and probabilistically checkable proofs [AS98, ALM<sup>+</sup>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 high rate. In particular, in the setting of constant query complexity, it is known that linear LCCs can not have constant rate [KT00]<sup>1</sup>, and that LTCs with certain restrictions can not have constant rate [DK11, BSV12]. On the other hand, the best known LCCs have exponential length<sup>2</sup>, and the best known LTCs have quasi-linear length [BS08, Din07, Vid13].

It turns out the picture is completely different when considering query complexity of  $n^\beta$  (for any constant  $\beta > 0$ ). In this setting, it has long been known that LTCs can achieve a constant rate [PS94, BS06]. More recently, it has been discovered that both LCCs [KSY11, GKS13, HOW13] and LTCs [Vid11] can achieve rates that are arbitrarily close to 1, where the relative distance is some small constant that depends on the rate sub-optimally. This discovery was quite surprising, considering the state of affairs in the setting of constant query complexity, and the general belief that local correctability and testability require much redundancy.

In this work, we show that LCCs and LTCs with  $n^\beta$  queries can, for any rate, approach the best possible relative distance. This means that, surprisingly, local correctability and local testability with  $n^\beta$  queries do not require “paying” anything in terms of rate and relative distance. In other words, LCCs and LTCs with  $n^\beta$  queries match the best error-correcting codes.

More specifically, recall that, by the Singleton bound [Kom53, Sin64], an error-correcting code of rate  $r$  can have relative distance of at most<sup>3</sup>  $1 - r$ . Our two main results are the following (we refer the reader to Section 2 for the required preliminaries).

**Theorem 1.1** (LCCs approaching the Singleton bound). *For every  $0 < r < 1$  and  $\beta, \varepsilon > 0$  there exists a finite field  $\mathbb{H} = \mathbb{F}_{2^p}$  such that the following holds: There exists an infinite family of ( $\mathbb{F}_2$ -linear) codes  $\{C_k\}_k$  such that the code  $C_k : \mathbb{F}_2^k \rightarrow \mathbb{H}^n$  is an LCC with message length  $k$ , rate at least  $r$ , relative distance at least  $1 - r - \varepsilon$ , and query complexity  $O(k^\beta)$ .*

**Theorem 1.2** (Strong LTCs approaching the Singleton bound). *For every  $0 < r < 1$  and  $\beta, \varepsilon > 0$  there exists a finite field  $\mathbb{H} = \mathbb{F}_{2^p}$  such that the following holds: There exists an infinite family of ( $\mathbb{F}_2$ -linear) codes  $\{C_k\}_k$  such that the code  $C_k : \mathbb{F}_2^k \rightarrow \mathbb{H}^n$  is a strong LTC with message length  $k$ , rate at least  $r$ , relative distance at least  $1 - r - \varepsilon$ , and query complexity  $O(k^\beta)$ .*

By concatenating the codes of Theorems 1.1 and 1.2 with Gilbert-Varshamov codes [Gil52, Var57] of constant length, we immediately obtain binary LCCs and the LTCs that approach the Zyablov bound, which represents the best known explicit construction of binary codes. More specifically, the Zyablov bound [Zya71] provides, for every  $\varepsilon > 0$  and any rate  $0 < r < 1$ , binary codes with rate  $r$ , and relative distance

$$Z_\varepsilon(r) \stackrel{\text{def}}{=} \max_{r < R < 1} \left\{ (1 - R - \varepsilon) \cdot H^{-1}\left(1 - \frac{r}{R}\right) \right\}$$

where  $H^{-1}$  is the inverse of the binary entropy function in the domain  $(0, \frac{1}{2})$ . We have the following results.

<sup>1</sup>[KT00] prove a lower bound for the related notion of LDCs (locally decodable codes). Since every linear LCC is also an LDC, their lower bound applies to linear LCCs.

<sup>2</sup>For example, a constant-degree Reed-Muller code is such an LCC.

<sup>3</sup>In fact, the relative distance may be slightly larger, i.e.,  $1 - r + \frac{1}{n}$  where  $n$  is the block length of the code.

**Corollary 1.3** (LCCs approaching the Zyablov bound). *For every  $\beta, r, \varepsilon > 0$  there exists an infinite family of linear binary codes  $\{C_k\}_k$ , such that the code  $C_k$  is an LCC with message length  $k$ , rate at least  $r$ , relative distance at least  $Z_\varepsilon(r)$ , and query complexity  $O(k^\beta)$ .*

**Corollary 1.4** (Strong LTCs approaching the Zyablov bound). *For every  $\beta, r, \varepsilon > 0$  there exists an infinite family of linear binary codes  $\{C_k\}_k$ , such that the code  $C_k$  is a strong LTC with message length  $k$ , rate at least  $r$ , relative distance at least  $Z_\varepsilon(r)$ , and query complexity  $O(k^\beta)$ .*

To sum-up, we construct LCCs and LTCs that achieve almost optimal trade-off between rate and relative distance over non-binary alphabets of constant size, and binary LCCs and LTCs that achieve the best known trade-off for binary codes.

**Previous work.** Our constructions are based on a technique of [AL96], who constructed codes that approach the Singleton bound and can be encoded and decoded from erasures in linear time.

Their construction has two stages: In the first stage, they construct “weak” codes with linear-time encoding and decoding. In the second stage, they use the latter codes to construct codes that approach the Singleton bound. The basic idea that underlies their second stage is the following: the codeword of the “weak” code is broken into blocks of constant size, and each of those blocks is encoded by Reed-Solomon. The decoding procedure works by decoding each of the blocks separately, which can corrects the majority of the erasures. Then, the decoding procedure of the “weak” code is applied to correct the remaining few erasures. Note that both steps can be done in linear time. We mention that the foregoing description is not complete, and that an additional modification needs to be made to make this idea work.

Before proceeding, we mention that the technique of [AL96] was used by [GI05] to construct similar codes that can be decoded from *errors* (rather than *erasures*). [GI05] followed the above scheme of [AL96], and their key step was to replace the codes of the first stage with codes that can be decoded from errors (but otherwise have the same properties). After this replacement was done, the second stage was more or less the same. A similar idea was used by [GI02, GR08] to construct capacity-achieving list-decodable codes with constant alphabet, where again the main step was replacing the codes of the first stage with appropriate codes.

**Our observation.** The key observation of this paper is that the same scheme works as well for LCCs and LTCs. Specifically, we replace the codes of the first stage of [AL96] with the LCCs of [KSY11] (for Theorem 1.1) or with the LTCs of [Vid11] (for Theorem 1.2). Then, we observe that the second stage of [AL96] works just as before. The reason is that decoding the Reed-Solomon blocks can be done locally, since those blocks are of constant size.

More generally, we wish to draw attention to the technique of [AL96]. We believe that it should be viewed as a general scheme for constructing codes that approach the Singleton bound. A “dream version” of such a scheme could be stated as follows.

**Claim 1.5** (“Dream version”). *For any property  $\mathcal{P}$  of codes the following holds. Suppose that for every constant  $r_W < 1$ , we can construct a “weak” code  $W$  such that:*

1.  $W$  has the property  $\mathcal{P}$ .
2.  $W$  has rate  $r_W$ .
3.  $W$  has constant relative distance  $\delta_W > 0$  which depends on  $r_W$  in an arbitrary way.

*Then, we can construct a code  $C$  that has the property  $\mathcal{P}$  and approaches the Singleton bound (in the sense of Theorems 1.1 and 1.2).*

Unfortunately, the technique of [AL96] does not imply such a clean and powerful claim. For example, such a claim obviously does not hold for the property  $\mathcal{P}$  of “not approaching the Singleton bound”. Worse yet, we do not have any characterization of the properties  $\mathcal{P}$  for which the scheme of [AL96] works. Nevertheless, the scheme of [AL96] does seem to work for many properties  $\mathcal{P}$ , and we believe that this is a good “take-home message” from this work.

**Organization of this paper.** We review the required preliminaries in Section 2, construct our LCCs in Section 3, and construct our LTCs in 4.

## 2 Preliminaries

All logarithms in this paper are in base 2. For any  $n \in \mathbb{N}$  we denote  $[n] \stackrel{\text{def}}{=} \{1 \dots, n\}$ . For any finite alphabet  $\Sigma$  and any two 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  $\text{dist}(x, y) \stackrel{\text{def}}{=} |\{x_i \neq y_i : i \in [n]\}| / n$ .

We denote by  $\mathbb{F}_{2^p}$  the finite field of  $2^p$  elements. If  $\mathbb{F}$  and  $\mathbb{H}$  are finite fields such that  $\mathbb{H}$  is an extension of  $\mathbb{F}$ , then we say that a function  $f : \mathbb{F}^k \rightarrow \mathbb{H}^n$  is  $\mathbb{F}$ -linear if for every  $\alpha, \beta \in \mathbb{F}$  and  $x, y \in \mathbb{F}^k$  it holds that

$$f(\alpha \cdot x + \beta \cdot y) = \alpha \cdot f(x) + \beta \cdot f(y).$$

### 2.1 Error correcting codes

Let  $\mathbb{F}$  and  $\mathbb{H}$  be finite fields such that  $\mathbb{H}$  is an extension field of  $\mathbb{F}$ . We say that  $C : \mathbb{F}^k \rightarrow \mathbb{H}^n$  is an  $\mathbb{F}$ -linear code  $C$  with **message length**  $k$  and **block length**  $n$  if it is an injective  $\mathbb{F}$ -linear function. If  $\mathbb{F} = \mathbb{H}$ , we say that  $C$  is a **linear code over  $\mathbb{F}$** , and if  $\mathbb{F} = \mathbb{F}_2$  then we say that  $C$  is a **binary linear code**. The **rate**  $r_C$  of the code  $C$  is the ratio  $\frac{k \cdot \log|\mathbb{F}|}{n \cdot \log|\mathbb{H}|}$ .

We will sometimes identify  $C$  with its image  $C(\mathbb{F}^k)$ . Specifically, we will write  $c \in C$  to indicate the fact that there exists  $x \in \mathbb{F}^k$  such that  $c = C(x)$ . In such case, we also say that  $c$  is a **codeword** of  $C$ . We say that  $C$  has **relative distance**  $\delta$  if for every two distinct codewords  $c_1, c_2 \in C$  it holds that  $\text{dist}(c_1, c_2) \geq \delta$ . We will use the notation  $\text{dist}(w, C)$  to denote the relative distance of a string  $w \in \mathbb{H}^n$  from  $C$ , and say that  $w$  is  $\varepsilon$ -close (respectively,  $\varepsilon$ -far) from  $C$  if  $\text{dist}(w, C) < \varepsilon$  (respectively, if  $\text{dist}(w, C) \geq \varepsilon$ ).

**Reed-Solomon codes.** We use the following fact, which states the existence of Reed-Solomon codes and their relevant properties.

**Fact 2.1** (Reed-Solomon Codes [RS60]). *For every  $k, n \in \mathbb{N}$  such that  $n \geq k$ , and a finite field  $\mathbb{F}$  such that  $|\mathbb{F}| \geq n$ , there exists a linear code  $RS_{k,n}$  over  $\mathbb{F}$  with message length  $k$ , block length  $n$ , rate  $r = k/n$ , and relative distance  $1 - \frac{k-1}{n} > 1 - r$ .*

We mention that Reed-Solomon codes meet the Singleton bound.

**Infinite families of codes.** An infinite family of codes  $C = \{C_k\}$  is a sequence of codes such that the code  $C_k$  has message length  $k$ . The block length  $n(k)$ , rate  $R(k)$  and relative distance  $\delta(k)$  of such a family are functions of  $k$  such that  $C_k$  has block length  $n(k)$ , rate  $R(k)$  and relative distance  $\delta(k)$ . We say that the family has **constant rate** (resp., **constant relative distance**) if  $R(k)$  (resp.,  $\delta(k)$ ) is a constant that is independent of  $k$ .

Throughout this paper we will often work with infinite families of codes, and refer to them simply as “codes”. For example, we will say that a code  $C$  has constant rate, and mean that the family  $C$  has constant rate. We will say that a string  $c$  is a codeword of  $C$  if it is a codeword of one of the codes  $C_k$ , and will say that a string  $w \in \mathbb{H}^n$  is  $\varepsilon$ -close to  $C$  if it is  $\varepsilon$ -close to one of the codes  $C_k$ .

## 2.2 Locally-correctable codes

Intuitively, a code is said to be locally correctable [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.

**Definition 2.2.** Let  $C = \{C_k\}_k$  be an infinite family of codes with block length  $n = n(k)$  and relative distance  $\delta = \delta(k)$ , and whose codewords belong to  $\mathbb{H}^{n(k)}$  for some finite field  $\mathbb{H}$ . Let  $q : \mathbb{N} \rightarrow \mathbb{N}$ . We say that  $C$  is **locally correctable** with **query complexity**  $q(k)$  if there exists a randomized algorithm  $A$  that satisfies the following requirements:

- **Input:**  $A$  takes as input a message length  $k \in \mathbb{N}$ , and a coordinate  $i \in [n]$  for  $n = n(k)$ , and also gets oracle access to a string  $z \in \mathbb{H}^n$  that is  $\frac{\delta(k)}{2}$ -close to a codeword  $c \in C_k$ .
- **Output:**  $A$  outputs  $c_i$  with probability at least  $\frac{2}{3}$ .
- **Query complexity:**  $A$  makes at most  $q(k)$  queries to the oracle  $z$ .

We say that the algorithm  $A$  is a **local corrector** of  $C$ .

**Remark 2.3.** The common definition of LCCs includes an additional parameter  $\tau \leq \frac{\delta}{2}$ , and only requires that  $A$  works when given a string  $u$  that is  $\tau$ -close to  $C$ . This definition can be “simulated” by our definition, by pretending that  $C$  has smaller relative distance than it actually has.

## 2.3 Locally-testable codes

Intuitively, a code is said to be locally testable [FS95, RS96, GS00] if, given a string  $z \in \mathbb{H}^n$ , 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. Below, we only give the definition of strong LTCs, since it is simpler and allows us to state a stronger result.

**Definition 2.4.** Let  $C = \{C_k\}_k$  be an infinite family of codes with block length  $n = n(k)$ , whose codewords belong to  $\mathbb{H}^{n(k)}$  for some finite field  $\mathbb{H}$ . Let  $q : \mathbb{N} \rightarrow \mathbb{N}$ . We say that  $C$  is **(strongly) locally testable** with **query complexity**  $q(k)$  if there exists a randomized algorithm  $A$  that satisfies the following requirements:

- **Input:**  $A$  takes as input a message length  $k \in \mathbb{N}$ , and also gets oracle access to a string  $z \in \mathbb{H}^{n(k)}$ .
- **Completeness:** If  $z$  is a codeword of  $C_k$ , then  $A$  accepts with probability 1.
- **Soundness:**  $A$  rejects with probability at least  $\text{dist}(z, C_k)/2$ .
- **Query complexity:**  $A$  makes at most  $q(k)$  non-adaptive queries to the oracle  $z$ .

We say that the algorithm  $A$  is a **local tester** of  $C$ .

**Remark 2.5.** The common definition of strong LTCs also includes an additional parameter  $\rho < 1$ , and requires that  $A$  rejects with probability  $\rho \cdot \text{dist}(u, C_k)$ . For simplicity, we chose to fix  $\rho$  to  $\frac{1}{2}$ . If  $\rho < \frac{1}{2}$ , one can amplify it to  $\frac{1}{2}$  by executing  $A$  multiple times.

## 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 it holds that

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

**Theorem 2.7.** For every  $\alpha, \gamma > 0$  there exists a constant  $d \in \mathbb{N}$  and an infinite family of graphs  $\{G_n\}_{n \in \mathbb{N}}$  such that the following holds for each graph  $G = G_n$ :

- $G = (U \cup V, E)$  is a bipartite  $d$ -regular graph with  $|U| = |V| = n$ .
- $G$  is an  $(\alpha, \gamma)$ -sampler.

Furthermore, there exists an algorithm that on input  $n$  outputs  $G_n$  in time  $\text{poly}(n)$ .

**Proof sketch.** We give a brief sketch of the proof, which uses the notions of edge expansion and second eigenvalue. We do not define these notions because they will not be used in the rest of the paper. The interested reader is referred to [HLW06]. In addition, we made no effort to optimize the dependence of  $d$  on  $\alpha$  and  $\gamma$ .

First, observe that it suffices to prove that there exists a family  $\{G'_n\}_{n \in \mathbb{N}}$  of graphs such that  $G'_n$  is a *non-bipartite*  $d$ -regular graph over  $n$  vertices, which has the required sampling property. The reason is that each such graph  $G'_n$  can be converted into a bipartite graph  $G_n$  with the sampling property, by taking two copies of the vertex set of  $G'_n$  and connecting the two copies according to the edges in  $G'_n$ .

We start constructing the family  $\{G'_n\}_{n \in \mathbb{N}}$  by constructing a family  $\{G''_n\}_{n \in \mathbb{N}}$  of graphs which has a constant edge expansion. As noted by [Din07], this can be done as follows: Known constructions of expanders (e.g. [RVW00]) give a graph  $G''_n$  with constant edge expansion for every  $n$  that is a power of 2. In order to deal with values of  $n$  of the form  $2^m - k$ , we construct an expander over  $2^m$  vertices and merge  $k$  pairs of vertices. Then, we maintain the regularity by adding self-loops. The resulting graph has  $2^m - k$  vertices, and maintains the edge expansion of the original graph over  $2^m$  vertices.

Next, we note that by the Cheeger inequality for expanders [Dod84, AM85], the family  $\{G''_n\}_{n \in \mathbb{N}}$  has second-largest normalized eigenvalue (in absolute value) that is bounded away from 1. This gives the sampling property for some fixed choices of  $\alpha$  and  $\gamma$  by the expander mixing lemma [AC88].

Finally, in order to improve  $\alpha$  and  $\gamma$  to the required level, we raise the graphs in  $\{G''_n\}_{n \in \mathbb{N}}$  to some constant power. We define the family  $\{G'_n\}_{n \in \mathbb{N}}$  to be the family of the resulting graphs. ■

## 3 LCCs approaching the Singleton bound

In this section, we prove Theorem 1.1, restated next.

**Theorem 1.1.** For every  $0 < r < 1$  and  $\beta, \varepsilon > 0$  there exists a finite field  $\mathbb{H} = \mathbb{F}_{2^p}$  such that the following holds: There exists an infinite family of ( $\mathbb{F}_2$ -linear) codes  $\{C_k\}_k$  such that the code  $C_k : \mathbb{F}_2^k \rightarrow \mathbb{H}^n$  is an LCC with message length  $k$ , rate at least  $r$ , relative distance at least  $1 - r - \varepsilon$ , and query complexity  $O(k^\beta)$ .

To this end, we use the following construction of LCCs with high rate and  $k^\beta$  queries, which follows from Theorem 4 of [KSY11]:

**Theorem 3.1** ([KSY11]). *For every  $0 < \beta, r_W < 1$  there exists  $\delta_W > 0$  such that the following holds. Let  $\mathbb{F}$  be a finite field. There exists an infinite family of codes  $\{W_k\}_k$  over  $\mathbb{F}$  such that  $W_k$  has message length  $k$ , rate at least  $r_W$ , relative distance at least  $\delta_W$ , and locally correctable with query complexity  $O(k^\beta)$ .*

The rest of this section is organized as follows. In Section 3.1, we give an overview of the proof of Theorem 1.1. Then, in Section 3.2, we provide a rigorous construction of the codes of Theorem 1.1 and Corollary 1.3. Finally, in Section 3.3, we prove that those codes are locally correctable.

### 3.1 Overview

We start with an overview of the construction of the code  $C = C_k$ . The following construction is due to [AL96]. Fix a rate  $r$  and a constant  $\varepsilon > 0$ . Our goal is to construct a code  $C$  with rate  $r$  and relative distance  $1 - r - \varepsilon$ .

Our construction will use two basic ingredients:

- The LCC  $W = W_k$  from Theorem 3.1 of rate  $\frac{1}{1+\varepsilon/2}$  and relative distance  $\delta_W$ , where  $\delta_W$  is some small constant.
- A Reed-Solomon code  $RS = RS_{b,d}$  from Fact 2.1, with rate  $r \cdot (1 + \frac{\varepsilon}{2})$  and relative distance at least  $1 - r - \frac{\varepsilon}{2}$ .

The idea of the construction is to combine the LCC  $W$  and the Reed-Solomon code  $RS$  to obtain a code  $C$  that enjoys “the best of the all worlds”: both the local correctability of  $W$  and the good parameters of  $RS$ . We do it in two steps: first, we construct a code  $C'$  which can be corrected from  $\frac{1-r-\varepsilon}{2}$  fraction of *random* errors. Then, we augment  $C'$  to obtain a code  $C$  that can be corrected from  $\frac{1-r-\varepsilon}{2}$  fraction of *adversarial* errors, and hence has relative distance  $1 - r - \varepsilon$ .

We first describe the construction of  $C'$ . The code  $C'$  encodes a message  $x$  as follows: the code  $C'$  first encodes  $x$  via  $W$ , thus obtaining a codeword  $w \in W$ . Then,  $C'$  partitions  $w$  into blocks of constant length  $b$  (to be determined later), and encodes each block with the code  $RS$ . The resulting string  $c'$  is defined to be the encoding of  $x$  via  $C'$ .

It is easy to see that  $C'$  has rate  $r$ , as required. We claim that if one applies to a codeword  $c' \in C'$  a noise that corrupts each coordinate with probability  $\frac{1-r-\varepsilon}{2}$ , 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  $\frac{1-r-\varepsilon/2}{2}$  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  $RS$ . Next, observe that if  $b$  is a sufficiently large constant, the fraction of “good blocks” is at least  $1 - \delta_W/2$ , and hence we can correct the remaining  $\delta_W/2$  fraction of errors using the decoding algorithm of  $W$ . It follows that  $C'$  can be corrected from  $\frac{1-r-\varepsilon}{2}$  fraction of random errors.

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 is to apply a pseudorandom permutation 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, since the adversary can take in the permutation into account when it chooses where to place the errors. Here the second idea comes into play: after applying the permutation to the coordinates of  $C'$ , 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 permutation to the coordinates of the code makes the errors behave pseudorandomly. This allows us to prove that the code can be decoded from  $\frac{1-r-\varepsilon}{2}$  errors, as we wanted.

### 3.2 Construction

Let  $\beta, r, \varepsilon > 0$  be as in Theorem 1.1. Let  $r_W \stackrel{\text{def}}{=} \frac{1}{1+\varepsilon/2}$ , and let  $\delta_W$  be the constant relative distance guaranteed by Theorem 3.1. Let  $d \in \mathbb{N}$  be a constant that is sufficiently large such that

- There exists a family of  $d$ -regular  $(\frac{1}{2} \cdot \delta_W, \frac{1}{4} \cdot \varepsilon)$ -samplers  $\{G_n\}_n$ , as in Theorem 2.7.
- There exists a Reed-Solomon code  $RS = RS_{b,d}$  of rate at least  $r \cdot (1 + \varepsilon/2)$ , relative distance at least  $1 - r - \frac{1}{2} \cdot \varepsilon$ , and block length  $d$ .

We choose the latter code  $RS_{b,d}$  to be over a finite field  $\mathbb{F}$  that is an extension field of  $\mathbb{F}_2$ . Let  $\mathbb{H}$  be an extension of  $\mathbb{F}$  of dimension  $d$ , so  $|\mathbb{H}| = |\mathbb{F}|^d$ . We show how to construct a family of  $\mathbb{F}$ -linear locally-correctable codes  $\{C_k : \mathbb{F}^k \rightarrow \mathbb{H}^{n(k)}\}_k$ .

**Remark 3.2.** Recall that Theorem 1.1 requires us to construct codes  $C$  whose messages are taken from  $\mathbb{F}_2^k$  rather than  $\mathbb{F}^k$ . However, every message in  $\mathbb{F}^k$  can be interpreted as a message in  $\mathbb{F}^{k'}$  (for some  $k' = O(k)$ ), by “unpacking” each symbol in  $\mathbb{F}$  into bits. Note that when the codes are reinterpreted in this way, they are  $\mathbb{F}_2$ -linear.

Fix a message length  $k \in \mathbb{N}$ . We explain how to construct a code  $C = C_k$  in the family. We use the following ingredients:

- The Reed-Solomon code  $RS_{b,d}$  from above, over the field  $\mathbb{F}$ , with message length  $b$  and block length  $d$ .
- A code  $W = W_k$  from Theorem 3.1, where  $W : \mathbb{F}^k \rightarrow \mathbb{F}^{n_W}$  is a code over  $\mathbb{F}$  with rate  $r_W$  and relative distance  $\delta_W$ , and is locally correctable with query complexity  $O(k^\beta)$ . For simplicity, we assume that  $b$  divides  $n_W$ , so  $n_W = n \cdot b$  for some  $n \in \mathbb{N}$ .
- A  $d$ -regular  $(\frac{1}{2} \cdot \delta_W, \frac{1}{4} \cdot \varepsilon)$ -sampler  $G = G_n = (U \cup V, E)$  with  $|U| = |V| = n$ , as in Theorem 2.7 above.

The code  $C$  encodes a message  $x \in \mathbb{F}^k$  as follows:

- $C$  first encodes  $x$  via the code  $W$ . This step yields a codeword  $w \in W$  of length  $n_W$ .
- Next,  $C$  partitions the string  $w$  into  $n$  blocks of length  $b$ , and encodes each block via the code  $RS_{b,d}$ . Let us denote the resulting string by  $c' \in \mathbb{F}^{n \cdot d}$  and the resulting codewords of  $RS_{b,d}$  by  $B_1, \dots, B_n \in \mathbb{F}^d$ .
- Now,  $C$  applies a “pseudorandom” permutation to the coordinates of  $c'$  as follows: Let  $U = \{u_1, \dots, u_n\}$  and  $V = \{v_1, \dots, v_n\}$  be the left and right vertices of  $G$  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  $S_1, \dots, S_n \in \mathbb{F}^d$ , by setting the  $j$ -th symbol of  $B'_i$  to be the symbol written on the  $j$ -th edge of  $v_i$ .



- Finally, we define a string  $c \in \mathbb{H}^n$  as follows: the  $i$ -th coordinate  $c_i$  is the block  $S_i \in \mathbb{F}^d$ , reinterpreted as a symbol of  $\mathbb{H}$ . We choose  $c$  to be the encoding of  $x$  via the code  $C$ .

This concludes the definition of the code  $C$ . It is not hard to see that  $C$  is  $\mathbb{F}$ -linear. The rate of  $C$  is

$$\begin{aligned}
\frac{k \cdot \log |\mathbb{F}|}{n \cdot \log |\mathbb{H}|} &= \frac{k \cdot \log |\mathbb{F}|}{n \cdot d \cdot \log |\mathbb{F}|} \\
&= \frac{k}{n \cdot d} \\
&= \frac{k}{n \cdot b} \cdot \frac{b}{d} \\
&= \frac{k}{n_W} \cdot \frac{b}{d} \\
&= r_W \cdot r \cdot \left(1 + \frac{\varepsilon}{3}\right) \\
&= r,
\end{aligned}$$

where the last equality follows from the definition of  $r_W$ . In the next section, we prove that  $C$  is locally correctable, and that the local correction algorithm can correct up to  $\frac{1-r-\varepsilon}{2}$  fraction of errors. This will imply that  $C$  has the required relative distance  $1 - r - \varepsilon$ .

### 3.3 Local correctability

In this section, we complete the proof of Theorem 1.1, by proving that the family  $\{C_k\}_k$  is locally correctable from  $\frac{1-r-\varepsilon}{2}$  fraction of errors using  $O(k^\beta)$  queries. To this end, we describe a local corrector  $A$ . The algorithm  $A$  is based on the following algorithm  $A_0$ , which locally decodes coordinates of  $W_k$  from a corrupted codeword of  $C_k$ .

**Lemma 3.3.** *There exists an algorithm  $A_0$  that satisfies the following requirements:*

- **Input:**  $A_0$  takes as input a message length  $k \in \mathbb{N}$ , a coordinate  $i \in [n_W]$ , and also gets oracle access to a string  $z \in \mathbb{H}^n$  that is  $\left(\frac{1-r-\varepsilon}{2}\right)$ -close to a codeword  $c \in C_k$ .
- **Output:** Let  $w^c$  be the codeword of  $W_k$  from which  $c$  was generated. Then,  $A_0$  outputs  $w_i^c$  with probability at least  $1 - \frac{1}{3 \cdot b \cdot d}$ .
- **Query complexity:**  $A$  makes at most  $O(k^\beta)$  queries to the oracle  $z$ .

Before proving Lemma 3.3, 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  $\left(\frac{1-r-\varepsilon}{2}\right)$ -close to a codeword  $c \in C_k$ , and a coordinate  $i \in [n]$ . The algorithm is required to decode  $c_i$ . Let  $w^c \in \mathbb{F}^{n_W}$  be the codeword of  $W$  from which  $c$  was generated, and let  $B_1^c, \dots, B_n^c$  and  $S_1^c, \dots, S_n^c$  be the corresponding blocks.

In order to decode  $c_i$ , the algorithm  $A$  should decode each of the symbols in the block  $S_i^c \in \mathbb{F}^d$ . Let  $B_{j_1}^c, \dots, B_{j_d}^c$  be the neighbors of  $S_i^c$  in the graph  $G_n$ . Each symbol of the block  $S_i^c$  belongs to one of the blocks  $B_{j_1}^c, \dots, B_{j_d}^c$ , and therefore it suffices to retrieve the latter blocks. Now, each block  $B_{j_h}^c$  is the encoding via  $RS_{b,d}$  of  $b$  symbols of  $w^c$ . The algorithm  $A$  invokes the algorithm  $A_0$  to decode each of those  $b$  symbols of  $w^c$ , for each of the blocks  $B_{j_1}^c, \dots, B_{j_d}^c$ . By the union bound, the algorithm  $A_0$  decodes all those  $b \cdot d$  symbols of  $w^c$  correctly with probability at least  $1 - \frac{1}{3 \cdot b \cdot d} \cdot b \cdot d = \frac{2}{3}$ . Whenever that happens, the algorithm  $A$  retrieves the blocks  $B_{j_1}^c, \dots, B_{j_d}^c$ , 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  $b \cdot d$ , and hence it is at most  $O(k^\beta)$ . It remains to prove Lemma 3.3.

**Proof of Lemma 3.3.** Let  $A_W$  be the local corrector of the code  $W$ . By amplification, we may assume that  $A_W$  errs with probability at most  $\frac{1}{3 \cdot b \cdot d}$ . Suppose that the algorithm  $A$  is invoked on a string  $z \in \mathbb{H}^n$  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  $q \in [n_W]$ , the algorithm  $A_0$  performs the follows steps.

1.  $A_0$  finds the block  $B_l$  to which the coordinate  $q$  belongs. Formally,  $l \stackrel{\text{def}}{=} \lceil q/b \rceil$ .
2.  $A_0$  finds the neighbors of the vertex  $u_l$  in  $G$ . Let us denote those vertices by  $v_{j_1}, \dots, v_{j_d}$ .
3.  $A_0$  queries the coordinates  $j_1, \dots, j_d$ , thus obtaining the blocks  $S_{j_1}, \dots, S_{j_d}$ .
4.  $A_0$  reconstructs the block  $B_l$  by reversing the permutation of  $G$  on  $S_{j_1}, \dots, S_{j_d}$ .
5.  $A_0$  decodes  $B_l$  to the nearest codeword of  $RS_{b,d}$ .
6.  $A_0$  retrieves the value of the  $q$ -th coordinate of  $w$  from the latter codeword of  $RS_{b,d}$ , and feeds it to  $A_W$  as an answer to its query.

When the algorithm  $A_W$  finishes running, the algorithm  $A_0$  finishes and outputs 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  $O(k^\beta)$ . It remains to show that  $A_0$  succeeds in decoding from  $\frac{1-r-\varepsilon}{2}$  fraction of errors.

Let  $z \in \mathbb{H}^n$  be a string that is  $(\frac{1-r-\varepsilon}{2})$ -close to a codeword  $c \in C$ . Let  $w^c \in \mathbb{F}^{n_W}$  be the codeword of  $W$  from which  $c$  was generated, and let  $B_1^c, \dots, B_n^c$  and  $S_1^c, \dots, S_n^c$  be the corresponding blocks. We also use the following definitions:

1. Let  $S_1^z, \dots, S_n^z \in \mathbb{F}^d$  be the blocks that correspond to the symbols of  $z$ .
2. Let  $B_1^z, \dots, B_n^z$  be the blocks that are obtained from  $S_1^z, \dots, S_n^z$  by reversing the permutation.
3. Let  $w^z \in \mathbb{F}^{n_W}$  be the string that is obtained by decoding each block  $B_l^z$  to the nearest codeword of  $RS_{b,d}$ , and extracting the coordinates of  $w$  from the resulting codewords.

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  $(\delta_W/2)$ -close to  $w^c$ , we will be done. In order to do so, it suffices to prove that for at least  $(1 - \frac{\delta_W}{2})$  fraction of the blocks  $B_l^z$ , it holds that  $B_l^c$  is the codeword of  $RS_{b,d}$  that is closest to  $B_l^z$ .

To this end, let  $j_1, \dots, j_t$  be the coordinates on which  $z$  and  $c$  differ. In other words, for every  $h \in [t]$  it holds that  $S_{j_h}^z \neq S_{j_h}^c$ . By assumption,  $t \leq (\frac{1-r-\varepsilon}{2}) \cdot n$ . Next, observe that since  $G_n$  is a  $(\frac{1}{2} \cdot \delta_W, \frac{1}{4} \cdot \varepsilon)$ -sampler, it holds that for at least  $(1 - \frac{\delta_W}{2})$  fraction of the vertices  $u_l$  of  $G$ , it holds that  $u_l$  has at most

$$\left( \frac{1-r-\varepsilon}{2} + \frac{\varepsilon}{4} \right) \cdot d = \frac{(1-r-\varepsilon/2)}{2} \cdot d$$

neighbors among  $j_1, \dots, j_t$ . Now, for each such vertex  $u_l$ , it holds that the block  $B_l^z$  is  $(\frac{1-r-\varepsilon/2}{2})$ -close to the block  $B_l^c$ . Since the code  $RS_{b,d}$  has relative distance  $1-r-\varepsilon/2$ , this implies that  $B_l^c$  is the codeword of  $RS_{b,d}$  that is closest to  $B_l^z$ , as required.  $\blacksquare$

**Obtaining the Zyablov bound.** As noted in the introduction, the codes that we constructed in this section can be concatenated with binary Gilbert-Varshamov codes, thus obtaining binary codes that achieve the Zyablov bound. We do not provide the details, since such constructions are standard in coding theory. Nevertheless, there is one subtle point that deserves attention: in order to decode the concatenated codes, we combine the local decoder  $A$  described above with the GMD decoder [Jr.66]. However, in order for the GMD decoder to work, the local corrector  $A$  needs to deal with both errors and erasures. While the algorithm  $A$  described above deals only with errors, it is not hard to modify it to deal with erasures as well. We refer the reader to [GI05] for an example of a similar construction.

## 4 LTCs approaching the Singleton bound

In this section, we prove Theorem 1.2, restated next.

**Theorem 1.2.** *For every  $0 < r < 1$  and  $\beta, \varepsilon > 0$  there exists a finite field  $\mathbb{H} = \mathbb{F}_{2^p}$  such that the following holds: There exists an infinite family of ( $\mathbb{F}_2$ -linear) codes  $\{C_k\}_k$  such that the code  $C_k : \mathbb{F}_2^k \rightarrow \mathbb{H}^n$  is a strong LTC with message length  $k$ , rate at least  $r$ , relative distance at least  $1 - r - \varepsilon$ , and query complexity  $O(k^\beta)$ .*

To this end, we use the following result, which follows from Theorem 3.1 of [Vid11].

**Theorem 4.1** ([Vid11]). *For every  $0 < \beta, r_W < 1$  there exists  $\delta_W > 0$  such that the following holds. Let  $\mathbb{F}$  be a finite field. There exists an infinite family of codes  $\{W_k\}_k$  over  $\mathbb{F}$  such that  $W_k$  has message length  $k$ , rate at least  $r_W$ , relative distance at least  $\delta_W$ , and locally testable with query complexity  $O(k^\beta)$ .*

Our construction of the LTCs  $\{C_k\}_k$  is the same as the construction of the LCCs of Section 3, with the only difference is that we use the LTCs of Theorem 4.1 instead of the LCCs of Theorem 3.1. Our LTCs have the required rate due to the same considerations as before. In order to show that our LTCs have the required relative distance, we use the same analysis of Section 3.3 to show that the codes can be corrected from  $\frac{1-r-\varepsilon}{2}$  fraction of errors, though here the correction is done by an algorithm that is not necessarily local or efficient.

It remains to show that those codes are locally testable. To this end, we describe a local tester  $A$ . In what follows, we use the notation of Section 3.2. Let  $A_W$  be the local tester of  $W = W_k$ .

When given oracle access to a purported codeword  $z \in \mathbb{H}^n$ , 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  $q \in [n_W]$ , the algorithm  $A$  performs the follows steps.

1.  $A$  finds the block  $B_l$  to which the coordinate  $q$  belongs. Formally,  $l \stackrel{\text{def}}{=} \lceil q/b \rceil$ .
2.  $A$  finds the neighbors of the vertex  $u_l$  in  $G$ . Let us denote those vertices by  $v_{j_1}, \dots, v_{j_d}$ .
3.  $A$  queries the coordinates  $j_1, \dots, j_d$ , and retrieves the blocks  $S_{j_1}, \dots, S_{j_d}$ .
4.  $A$  reconstructs the block  $B_l$  by reversing the permutation of  $G$  on  $S_{j_1}, \dots, S_{j_d}$ .
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  $q$ -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 at most  $d$  times the query complexity of  $A_W$ , and hence it is  $O(k^\beta)$ . 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  $A$  rejects with probability  $\text{dist}(z, C)/2$ . To this end, it suffices to prove that  $A$  rejects with probability at least  $\eta \cdot \text{dist}(z, C)$  for some constant  $\eta > 0$ , since  $\eta$  can be amplified to  $\frac{1}{2}$  by repetition. We use the following definitions:

1. Let  $S_1, \dots, S_n \in \mathbb{F}^d$  be the blocks that correspond to the symbols of  $z$ .
2. Let  $B_1, \dots, B_n$  be the blocks that are obtained from  $S_1, \dots, S_n$  by reversing the permutation.
3. Let  $w^z \in (\mathbb{F} \cup \{?\})^{nw}$  be the string that is obtained by from the blocks  $B_1, \dots, B_n$ : for each block  $B_l$  that is a legal codeword of  $RS_{b,d}$ , we extract from  $B_l$  the corresponding coordinates of  $w^z$  in the natural way. For each block  $B_l$  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  $\mathbb{F}^{nw}$ , and not on strings in  $(\mathbb{F} \cup \{?\})^{nw}$ . To deal with this technicality, we define an algorithm  $A'_W$  that, when given access to a string  $y \in (\mathbb{F} \cup \{?\})^{nw}$ , 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.

**Proposition 4.2.**  *$A'_W$  rejects a string  $y \in (\mathbb{F} \cup \{?\})^{nw}$  with probability at least*

$$\frac{1}{4} \cdot \min \{ \text{dist}(y, W), \delta_W \}.$$

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

$$\text{dist}(w^z, W) \geq \frac{1}{b \cdot d} \cdot \text{dist}(z, C),$$

since every coordinate of  $C$  is generated from at most  $b \cdot d$  coordinates of  $W$ . It thus follows that  $A$  rejects  $z$  with probability at least

$$\frac{1}{4} \cdot \min \{ \text{dist}(w^z, W), \delta_W \} \geq \min \left\{ \frac{1}{4 \cdot b \cdot d} \cdot \text{dist}(z, C), \frac{1}{4} \cdot \delta_W \right\}.$$

We conclude the proof by setting  $\eta = \min \left\{ \frac{1}{4 \cdot b \cdot d}, \frac{1}{4} \cdot \delta_W \right\}$ .

#### 4.1 Proof of Proposition 4.2

We use the following result.

**Claim 4.3.** *Let  $I \subseteq [n_W]$  be a set of coordinates. The algorithm  $A_W$  queries  $I$  with probability at least*

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

Note that this claim only makes sense since we assumed that  $A_W$  makes *non-adaptive* queries (we assumed it in Definition 2.4). Without this assumption, the probability that  $A_W$  queries  $I$  would have depended on the string that  $A_W$  gets oracle access to.

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

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

and therefore  $A_W$  rejects  $s$  with probability at least  $\frac{1}{2} \cdot \frac{|I|}{n}$ . On the other hand,  $A_W$  can only reject  $s$  if it queries  $I$ , since otherwise it can not distinguish between  $s$  and the all-zeroes codeword. It follows that  $A_W$  queries  $I$  with probability at least  $\frac{1}{2} \cdot \frac{|I|}{n}$ , as required. ■

We turn to proving Proposition 4.2. Let

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

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

- **$E$  is “large”:** Suppose that  $\frac{|E|}{n} \geq \frac{1}{2} \cdot \text{dist}(y, W)$ . In this case, it holds by Claim 4.3 that  $A_W$  queries  $E$  with probability at least

$$\frac{1}{4} \cdot \min \{ \text{dist}(y, W), \delta_W \}.$$

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

- **$E$  is “small”:** Suppose that  $\frac{|E|}{n} \leq \frac{1}{2} \cdot \text{dist}(y, W)$ . Let  $y_0 \in \mathbb{F}^{n_W}$  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  $E$ . We have that

$$\begin{aligned} \Pr [A'_W \text{ rejects } y] &= \Pr [\mathcal{E}] \cdot \Pr [A'_W \text{ rejects } y | \mathcal{E}] + \Pr [\neg \mathcal{E}] \cdot \Pr [A'_W \text{ rejects } y | \neg \mathcal{E}] \\ &= \Pr [\mathcal{E}] \cdot 1 + \Pr [\neg \mathcal{E}] \cdot \Pr [A_W \text{ rejects } y_0 | \neg \mathcal{E}] \\ &\geq \Pr [\mathcal{E}] \cdot \Pr [A_W \text{ rejects } y_0 | \mathcal{E}] + \Pr [\neg \mathcal{E}] \cdot \Pr [A_W \text{ rejects } y_0 | \neg \mathcal{E}] \\ &= \Pr [A_W \text{ rejects } y_0] \\ &\geq \frac{1}{2} \cdot \text{dist}(y_0, W) \\ &\geq \frac{1}{4} \cdot \text{dist}(y, W), \end{aligned}$$

as required.

This concludes the proof.

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