Mutually Uncorrelated Codes for DNA Storage

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Research Thesis

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Contents

Abstract 1

Abbreviations and Notations 3

1 Introduction 4
   1.1 Motivation ............................................. 4
   1.2 Our Work ............................................. 5
   1.3 Outline of the Thesis ................................. 6

2 Background and Previous Results 7
   2.1 DNA Basics .......................................... 7
   2.2 DNA Storage Systems ................................. 8
   2.3 Random Access in DNA Storage ....................... 9
   2.4 Definitions and Previous work ...................... 10

3 Mutually Uncorrelated Codes 13
   3.1 Run Length Limited Constraint ................... 13
       3.1.1 Cardinality Analysis ......................... 13
       3.1.2 Efficient Encoding and Decoding Algorithms . 23
   3.2 Cardinality of MU Code Construction .............. 26

4 \((d_h, d_m)\)-Mutually Uncorrelated Codes 33
   4.1 Window Weight Limited Constraint ................. 33
   4.2 \((d_h, d_m)\)-MU Codes Cardinality Analysis ....... 37
       4.2.1 Upper Bound .................................. 37
       4.2.2 Construction .................................. 38
   4.3 Related Families of Codes ......................... 41
List of Figures

3.1 Comparison between the construction’s cardinality according to Lemma 3.11, multiplied by $\frac{n}{q^n}$ for different values of $z$, and the bounds (2.2) and (2.3) from [11, 16, 25] . . . . . . . . . . . . . . 31

3.2 A comparison between the construction’s maximal cardinality according to Theorem 3.12 and the upper bound of the cardinality of MU codes from [28], mentioned in (2.1). . . . . . . . . . . . . . . 32
List of Tables

4.1 Redundancy summary of $k$-RLL and $(d, k)$-WWL constraints . . . . 36
6.1 Redundancy summary for binary MU codes . . . . . . . . . . . . . . 54
Abstract

Mutually Uncorrelated (MU) codes are a class of codes in which no proper prefix of one codeword is a suffix of another codeword. These codes were originally studied for synchronization purposes and recently, Yazdi et al. showed their applicability to enable random access in DNA storage.

In the context of storing data, we refer to a DNA strand as a linear sequence over the alphabet \{A,C,G,T\}. In 2012 and 2013 two pioneering DNA based storage systems were implemented by Church et al. and Goldman et al. respectively. Those systems paved the way for multiple studies in which synthetic DNA molecules are used as a volume for storing data. The different systems share similar architecture principles. The writing channel is a chemical process named synthesis that conducts artificial DNA strands. The reading is done using a technology called DNA sequencing. The length of the strands is limited to few hundreds symbols hence the DNA pool consists multiple strands in order to store reasonable amount of data.

To allow random access, Yazdi et al. suggest to attach to each information strand an addressing part for the purpose of identifying the requested strand during the reading process. The constraints that the addresses set should satisfy are listed by Yazdi et al.

In this work we follow their research and study three of the listed constraints. Namely, we investigate MU codes along with their extensions to correct errors and balanced codes. We first review a well known construction of MU codes and study the asymptotic behavior of its cardinality. This task is accomplished by studying a special class of run-length limited codes that impose the longest run of zeros to be at most some function of the codeword length. We also present an efficient algorithm for this class of constrained codes and show how to use this analysis for MU codes. Next, we extend the results on the run-length limited codes in order to study \((d_h, d_m)\)-MU codes that impose a minimum Hamming distance of \(d_h\) between different codewords and \(d_m\) between prefixes and suffixes. In particular,
we show an efficient construction of these codes with nearly optimal redundancy. We also draw connections to the problems of comma-free and prefix synchronized codes. Lastly, we provide similar results for the edit distance and balanced MU codes.
## Abbreviations and Notations

| *for i, j ∈ ℤ, [i, j]* | \{z | z ∈ ℤ, i ≤ z ≤ j\} |
| *for k ∈ ℤ[k]* | \{1, k\} |
| \( \Sigma \) | binary alphabet, \{0, 1\} |
| \( \Sigma_q \) | \( q \)-ary alphabet, \{0, \ldots, q – 1\} |
| \( ⟨n⟩_k \) | \( n \) modulo \( k \) |
| \( a ∈ \Sigma_q^n \) | \( (a_1, \ldots, a_n), a_i ∈ Σ_q \) |
| *for i, j ∈ [n], i ≤ j : ai* | subvector of \( a \), \( (a_i, \ldots, a_j) \) |
| *for i, j ∈ [n], i > j : ai* | empty word |
| \( w_H(a) \) | the Hamming weight of a vector \( a \) |
| \( d_H(a, b) \) | the Hamming distance between \( a \) and \( b \) |
| \( a^i \) | concatenation of the vector \( a \) \( i \) times |
| \( ab \) | concatenation of the two vectors \( a \) and \( b \) |
| \( A^i \) | concatenation of the set \( A \) \( i \) times |
| \( AB \) | concatenation of the two sets \( A \) and \( B \) |
| \( f(n) \preceq g(n) \) | \( \lim_{n→∞} \frac{f(n)}{g(n)} \leq 1 \) |
| \( f(n) \approx g(n) \) | \( \lim_{n→∞} \frac{f(n)}{g(n)} = 1 \) |
| \( \text{red}(A) \) | redundancy of a set \( A ⊆ Σ_q^n \), \( n − \log_q |A| \) |
| \( \log x \) | \( \log_2 x \) |
Chapter 1

Introduction

1.1 Motivation

Mutually Uncorrelated (MU) codes satisfy the constraint in which the prefixes set and suffixes set of all codewords are disjoint. This class of codes was first studied by Levenshtein [25] for the purpose of synchronization, and has received attention recently due to its relevance and applicability for DNA storage [39]. Namely, these codes offer random access of DNA blocks in synthetic DNA storage.

The potential of DNA molecules as a volume for storing data was recognized due to its unique qualities of density and durability. The first large scale DNA storage system was designed by Church et al. [12] in 2012. Since then a few similar systems were implemented for archival applications as they did not support random access to the memory. Recently, in [9,39] two random access DNA storage systems were proposed. To enable random access, the authors suggested to equip the DNA information blocks with unique addresses, also known as primers. The constraints that a good primers set should satisfy are listed by Yazdi et al. in [39]. In this paper we focus on three of the four listed constraints, described as follows: 1) the MU constraint is imposed as it avoids overlaps in two primers, which are likely to cause erroneous identification of DNA blocks, 2) both the writing and reading channels of DNA introduce substitution errors, therefore we are interested in large mutual Hamming distance, and 3) we require the primers to be balanced since balanced DNA sequences increase the chances of successful reads. In addition, the authors of [37] mentioned that deletion errors are introduced during synthesis. We therefore also extend our study to MU codes with edit distance.
1.2 Our Work

MU codes were rigorously studied in the literature under different names such as codes without overlaps \([28, 29]\), non-overlapping codes \([7]\) and cross-bifix-free codes \([2, 6, 11]\). However, the basic problem of finding the largest MU code is still not fully solved. Let us define by \(A_{\text{MU}}(n, q)\) the size of the largest MU code over a field of size \(q\). The best upper bound, found by Levenshtein \([28]\), states that \(A_{\text{MU}}(n, q) < q^n / (e(n - 1))\), while the best known constructive lower bound, given independently in \([11, 16, 25]\), states that \(A_{\text{MU}}(n, q) \gtrsim q^{n - 1} / q^n\). Hence, for the binary case there is still a gap of factor 2 between the lower and upper bounds. The construction of MU codes is explicit. Given some \(k < n\), one fixes the first \(k\) bits to be zero, followed by a single non-zero symbol, and the last symbol is non-zero as well. The sequence of the remaining \(n - k - 2\) symbols needs to satisfy the constraint that it does not have a zeros run of length \(k\). Previous results claimed that \(q^{n - 1} / q^n\) is a lower bound on the construction’s code size, when \(n = (q^i - 1) / (q - 1)\), however it was not known whether it is possible to achieve codes with larger cardinality. We give an explicit expression of the asymptotic cardinality of these codes for any value of \(n\) and show that the lower bound \(q^{n - 1} / q^n\) is indeed tight. This result is accomplished by studying a special family of run-length limited constrained codes, called \(k\)-run length limited (RLL) codes, which impose the longest run of zeros to be of size at most \(k - 1\), where \(k\) is a function of the word’s length. We also present an efficient encoding and decoding algorithm for \(k\)-RLL codes with linear complexity which results in efficient binary MU codes with \(\lceil \log(n) \rceil + 4\) redundancy bits.

Next, we extend the study of \(k\)-RLL codes to the window weight limited constraint that imposes the Hamming weight of every length-\(k\) subsequence to be at least some prescribed number \(d\). Accordingly, we study MU codes with error-correction capabilities. A \((d_h, d_m)\)-MU code is an MU code with minimum Hamming distance \(d_h\), with the additional property that every prefix of length \(i\) differs by at least \(\min\{i, d_m\}\) symbols from all proper length suffixes. We show an upper bound on the size of \((d_h, d_m)\)-MU codes and give a construction of such codes with encoder and decoder of linear complexity. The redundancy of the construction is \(\left\lfloor \frac{d_h + 1}{2} \right\rfloor \log_q(n) + (d_m - 1) \log_q n + O(d_m \log d_m)\), which is nearly optimal with respect to our upper bound on the code size. A similar constraint is imposed when studying MU codes with edit distance. We give a general result of such codes and study MU codes that can correct a single deletion or insertion. For the latter
case we use a systematic encoder for the Varshamov Tenengolts codes [36]. Lastly, we study balanced MU codes, that is, codes which are both MU and balanced. We show that the achievable minimum redundancy of these codes is approximately $1.5 \log(n)$ and for an efficient construction we use Knuth’s algorithm while the redundancy is roughly $2 \log(n)$ bits.

1.3 Outline of the Thesis

The rest of the thesis is organized as follows. In Chapter 2, we introduce the basic terms related to DNA data storage and review some of the DNA storage systems presented recently to understand the motivation for this study. We then formally define the codes studied in this work and review related work. In Chapter 3, Section 3.1, we analyze the redundancy of $k$-RLL codes, when $k$ is a function of the word’s length, and propose efficient encoding and decoding algorithms for these codes. We use this analysis in Section 3.2 in order to study the asymptotic cardinality of a well known family of MU codes. We then define in Chapter 4 $(d_h, d_m)$-MU codes and extend the results on $k$-RLL codes to study the window weight limited constraint. Accordingly in Section 4.2, we extend the class of MU codes to $(d_h, d_m)$-MU codes construction and study its size. Furthermore, in Section 4.3 we draw connections to the problems of comma-free and prefix synchronized codes. We continue in Chapter 5 to explore other variations of MU codes, in Section 5.1 we study MU codes that can correct deletions and insertions, and in Section 5.2 we study balanced MU codes. Lastly, Chapter 6 concludes and summarizes the results of this work.
Chapter 2

Background and Previous Results

2.1 DNA Basics

*DeoxyriboNucleic Acid (DNA) strand* is a molecule constructed as a linear sequence, which consists of four types of *nucleotides*: Adenine (A), Cytonine(C), Guanine(G), and Thymine(T). A nucleotide in a strand is also referred to as a *base*. Motivated by the chemical bonds performed between the different nucleotides, A and T are said to be the complement bases of each other, and likewise for G and C. Two single stranded sequences can bind and form a *double strand*, if they are complementary, that is, each base in one strand aligns with the corresponding complementary base in the other strand. The process of combining two complementary single stranded sequences into one double stranded sequence is termed *hybridization*. Two strands do not have to be entirely complementary in order to bind to each other. In this case the binding process is referred to as *cross hybridization*.

*DNA Synthesis* is the process of creating DNA molecules. Arbitrary single stranded DNA sequences of length few hundreds bases, can be generated chemically. This process results in a DNA pool which consists of all the synthesized strands. In the context of storage, synthesis corresponds to the writing channel of storing data in the DNA storage, however as we shall see later, this process does not permit structured addressing of the written data (also known as random access).

*Sequencing* is the process of reading single stranded DNA sequences from the DNA pool. The most popular technique, called "sequencing by synthesis", involves an enzyme and free fluorescent nucleotides. The enzyme helps complementary nucleotides to form bonds and as a result, the original single strand in the pool...
binds, nucleotide after nucleotide, with the free fluorescent nucleotides to form a double stranded sequence. Since each type of fluorescent nucleotide emits a different color in the binding process, it is possible to read the complement strand optically.

The major progress in sequencing technologies in recent years has been achieved by a method termed next-generation sequencing (NGS) which allows generation of tremendously large number of sequencing reactions in parallel. Specifically, Illumina has played a significant role in advancing the sequencing industry. It is currently the leading company providing a variety of DNA sequencers to different applications.

2.2 DNA Storage Systems

The idea of using DNA molecules as a volume for storing data was first introduced in the late 1950s by Richard Feynman in his lecture "There’s plenty of room at the bottom". Feynman identified the unique qualities of DNA molecules, i.e. its density and durability, and understood the potential of DNA storage. However, at that time the technology was not mature enough in order to allow artificial manipulation of DNA molecules. Several decades later, in 1990, the human genome project was initiated with the objective of determining the DNA sequence of the entire human genome. A tremendous effort was invested, leading to a valuable progress in DNA sequencing and assembly methods. The completion of the project was announced in 2003 and improvements are still seen continuously. Despite this progress, both the cost and performance of DNA technologies are far from competing with existing storage solutions. On the other hand, important biotechnology applications are expected to continue driving the growth in the field and as a result DNA storage systems are expected to become a realistic platform for storing archival data.

There are two main approaches to store data in DNA, namely DNA storage in vivo (inside an organism) and in vitro (outside). For the in vivo case, the data is stored in the DNA molecule inside a living cell. This approach was implemented in [15], in which 1280 characters were encoded in a bacterial genome and were recovered successfully. In the in vitro setup, the data is stored in a DNA pool external to a living organism. To the best of our knowledge, in vivo storage is inapplicable to large scale storage, as opposed to in vitro methods, which get more attention recently; see [8, 9, 12, 14, 18, 19, 39] for more details.

The first large scale DNA storage system was implemented by Church et al.
in 2012 [12]. Church’s approach was to divide the data into blocks of size 96 bits. A unique id of length 19 bits was attached to each block for the purpose of identifying the location of the block in the file. Then, the 115 bits were converted symbol by symbol to DNA bases: 0 was converted to A or C and 1 to T or G, taking chemical limitations into account (in order to avoid problematic patterns such as repetitions of the same symbol and to ensure balanced sequences). The resulting sequences were synthesized (i.e. written) with density of $0.7 \cdot 10^{15}$ Bytes per gram (B/g). Each sequence was stored multiple times in order to avoid errors (that is, a repetition error-correcting code was applied). After a few post processing steps the data was decoded perfectly using Illumina’s sequencing technology. Using this method, a file of size 5.27MB was stored.

Later, in [18], Goldman et al. designed a similar storage system to the previous one with improved density of $2.2 \cdot 10^{15}$ B/g. This improvement was achieved by using several coding techniques such as Huffman, differential, and repetition coding.

### 2.3 Random Access in DNA Storage

The systems described in Section 2.2 established the first step towards large scale DNA based storage systems. However, they both lack an important feature for any storage system (even archival systems) of allowing random access. In order to read even a single bit, one has to retrieve the entire DNA pool, decode all the sequences to the binary data, assemble the file and then choose the relevant information.

A solution to this problem was proposed in 2015 by Yazdi et al. [39]. In order to enable random access memory, the authors of [39] suggest the following method. As part of the writing process, prior to synthesis, each DNA block is equipped with a unique address, also known as *primer*. Then, the reading channel receives the primer of the requested DNA block as an input. Recall that sequencing is done with the help of an enzyme. The enzyme binds the single strands with free nucleotides, resulting in double strands which are read optically. In the random access system, the sequencing process starts with an additional phase aimed to identify the requested DNA block. During the identification process the complementary sequence of the unique primer is sent to the DNA pool and hybridize with the primer part of the corresponding DNA block. By that, the selection of the DNA block is done and the enzyme acts solely on the selected DNA block. Consequently, only the selected DNA block binds with the complementary nucleotides
to form a double strand, and thus only the selected DNA block is read.

Selecting a good set of primers is a key aspect in achieving the desired random access feature. A good set ensures correct identification of the DNA blocks by guaranteeing the success of the involved chemical processes. We now present the constraints that a good primers set should satisfy, according to Yazdi et al. [39].

- (C1) GC content of approximately 50%, also known as balanced sequences which are more stable chemically.
- (C2) Large mutual Hamming distance, to avoid identification errors due to substitutions in the sequencing and synthesis processes.
- (C3) Uncorrelatedness of sequences is imposed to prevent overlaps of two addresses.
- (C4) Absence of folding structure. We do not look into this constraint in this work.

In addition, the authors of [37] mentioned that deletion errors are introduced during synthesis. We therefore also extend our study to MU codes with edit distance.

2.4 Definitions and Previous work

**Definition 2.1.** Two not necessarily distinct words \(a, b \in \Sigma_q^n\) are mutually uncorrelated if any non-trivial prefix of \(a\) does not match any non-trivial suffix of \(b\). A code \(C \subseteq \Sigma_q^n\) is a mutually uncorrelated (MU) code if any two not necessarily distinct codewords of \(C\) are mutually uncorrelated.

Let us denote by \(A_{MU}(n, q)\) the largest cardinality of an MU code of length \(n\) over \(\Sigma_q\). Levenshtein showed in [28] that for all \(n\) and \(q\)

\[
A_{MU}(n, q) \leq \left( \frac{n-1}{n} \right)^{n-1} \frac{q^n}{n} < \frac{q^n}{e(n-1)},
\]

or in other words, the redundancy is lower bounded by \(\log e + \log(n-1)\). Later, in [7] Blackburn suggested a slightly weaker bound, \(A_{MU}(n, q) < \frac{q^n}{2(n-1)}\), with the advantage of its proof simplicity.
We now recall a well studied family of MU codes. It was suggested independently first by Gilbert in [16], later by Levenshtein in [25] and recently by Chee et al. in [11].

**Construction I.** Let $n, k$ be two integers such that $1 \leq k < n$ and let $C_1(n, q, k) \subseteq \Sigma_q^n$ be the following code:

$$C_1(n, q, k) = \{ a \in \Sigma_q^n \mid \forall i \in [k], a_i = 0, a_{k+1} \neq 0, a_{n-1} \neq 0,$$

$$a_{n-k+1}^{n-2} \text{ has no zeros run of length } k \}.$$

We denote

$$C_1(n, q) = \max_{1 \leq k < n} \{|C_1(n, q, k)|\}.$$

It was proved in [11, 16, 25] that there exists an appropriate choice of $k$ for which the following lower bounds hold

$$C_1(n, q) \gtrsim q^{-\frac{q-1}{q-1}} \ln q \frac{q^n}{n}, \tag{2.2}$$

as $n \to \infty$, and more specifically

$$C_1(n, q) \gtrsim q^{-1} q^e \cdot \frac{q^n}{n}, \tag{2.3}$$

where $n \to \infty$ over the subsequence $\frac{q^i-1}{q-1}, i \in \mathbb{N}$. However, it was not established in these works what the asymptotic cardinality of $C_1(n, q)$ is and for which $k$ it is achieved. We answer these questions in Section 3.2 and show that the asymptotic inequalities in (2.2), (2.3) are asymptotically tight.

Additional interesting approach for constructing binary MU codes was proposed in [6]. This work is based on a combinatorial term named *Dyck words*. A dyck word is a word over the binary alphabet, of even length $n$, that has equal numbers of ones an zeros. Moreover, any prefix of a Dyck word contains more ones than zeros. We can think of a Dyck word as a legal ordering of opening and closing parenthesis, where the opening parenthesis represents 1 and the closing, 0. The construction from [6] is the set of binary words of the form $1d$ such that $d$ is a Dyck word. This is an MU-code because any prefix of a codeword contains more ones than zeros, as apposed to a suffix of a codeword. The idea of this construction was later generalized in [2] to alphabets of size $q > 2$. The idea behind
this work was based on Motzkin path which is a generalization of the term Dyck word, for \( q > 2 \). In this case, it was shown that for some small values of \( n \) the approach in [2] achieves larger cardinality than Construction I. Both works [2, 6] do not offer asymptotic analysis of the constructions’ redundancy, however, for the binary case it is commonly believed that Construction I provides the best known asymptotic code size. Furthermore, the author of [7] extended Construction I for \( q > 2 \) and showed that if \( q \) is a multiple of \( n \) then this extension results with strictly optimal codes. Lastly, we note that in [4, 5] Gray codes were presented for listing the vectors of the code \( C_1(n, q, k) \).

Due to the structure of Construction I, the tools that we use for analyzing its cardinality are taken from studies in the field of constraint coding. We recall the well known Run Length Limited (RLL) constraint in which the lengths of the zero runs are limited to a fixed range of values.

**Definition 2.2.** Let \( k \) and \( n \) be two integers. We say that a vector \( \mathbf{a} \in \Sigma_q^n \) satisfies the \( k \)-run length limited (RLL) constraint, and is called a \( k \)-RLL vector, if \( n < k \) or \( \forall i \in [n - k + 1] : w_H(\mathbf{a}_{i+k-1}^{i+k-1}) \geq 1 \).

In other words, a vector is called a \( k \)-RLL vector if it does not contain a zeros run of length \( k \). We denote by \( A_q(n, k) \) the set of all \( k \)-RLL length-\( n \) vectors over the alphabet \( \Sigma_q \), and \( a_q(n, k) = |A_q(n, k)| \).

The definition of the \( k \)-RLL constraint is similar to the \((d, k)\)-RLL constraint from [31] that states that any zeros run is of length at least \( d \) and at most \( k \). In other words, the \( k \)-RLL constraint is equivalent to the well studied \((0, k - 1)\)-RLL constraint. We chose the notation of \( k \)-RLL for simplicity of the following analysis to come.

The capacity of the \( k \)-RLL constraint, is defined for any fixed values of \( k \) and \( q \) as

\[
E_{k,q} = \lim_{n \to \infty} \frac{\log(a_q(n, k))}{n}.
\]

Kato and Zeger [21] provided the following result for the binary case and later Jain et. al generalized it for \( q \geq 2 \) in [20]

\[
E_{k,q} = \log q - \frac{(q - 1) \log e}{q^{k+2}} (1 + o(1)).
\]
Chapter 3

Mutually Uncorrelated Codes

3.1 Run Length Limited Constraint

The $k$-RLL constraint was studied extensively in the literature previously by many works; see [31] and references therein. However, this study was focused solely for the case in which $k$ is fixed, meaning $k$ is independent of $n$. As it will be explained later, the MU codes problem requires analysis of values of $k$ which are dependent on $n$. In Subsection 3.1.1, we resort to previous results on the $k$-RLL constraint, when $k$ is fixed to provide a better understanding on the asymptotic behavior of $a_q(n, k)$ when $k$ depends on $n$. Later in Subsection 3.1.2 we present efficient encoding and decoding algorithms to avoid zeros runs of length $\lceil \log n \rceil + 1$, with linear time and space complexity, and with only a single bit of redundancy.

3.1.1 Cardinality Analysis

We start by giving general bounds on $a_q(n, k)$ for most values of $n, k$ in Lemma 3.1 and Lemma 3.3. The intuition of the proofs is as follows. In Lemma 3.1, we consider the set of all length-$n$ vectors which are a concatenation of multiple length-$2k$ $k$-RLL vectors, i.e. each length-$2k$ vector satisfies the $k$-RLL constraint. The set $A_q(n, k)$ is a subset of this set and hence $a_q(n, k)$ is upper bounded by its size. For the lower bound we consider the set of all vectors of length $n$ which are again, a concatenation of multiple shorter $k$-RLL vectors, however, now we extract vectors that start or end with $\lceil k/2 \rceil$ zeros. The new set is a subset of $A_q(n, k)$ and a lower bound is derived accordingly.

In Lemma 3.3 we study the value $a_q(mn, k)$ and bound it similarly to the...
bounds in Lemma 3.1 by considering a concatenation of \( m \) vectors of length \( n \). Both the upper and lower bounds on \( a_q(mn,k) \) will include the size \( a_q(n,k) \) and thus equivalent upper and lower bounds on \( a_q(n,k) \) will be established. We then derive the final bounds using the expression \( E_{k,q} = \lim_{m \to \infty} \frac{\log(a_q(mn,k))}{mn} \) for the capacity.

**Lemma 3.1.** Let \( n,k \) be positive integers such that \( 5 \leq k \leq n \), then

\[
q^n \left( \frac{q - 1}{q} \right)^e \frac{q}{q+1} e^{-\frac{2}{e+1}} \leq a_q(n,k) \leq q^n \frac{e^{-\frac{1}{2}}}{\sqrt{q-1}} ,
\]

where \( c_1 = \frac{(q-1)^{\frac{1}{2}}}{2q}, c_2 = \frac{q^k}{q+1}, c_3 = \frac{\log_q(e(q-1)^2)}{2q^2} \).

**Proof** Part 1: Upper Bound

We consider the set \( A_q(2k,k) \) of \( \frac{n}{2k} \) vectors of length \( n \). The resulting set of length-\( n \) vectors is denoted by

\[
B_q(n,k) = A_q(2k,k) \left\lfloor \frac{n}{2k} \right\rfloor \Sigma_q^{(n)2k}.
\]

Note that \( A_q(n,k) \subseteq B_q(n,k) \) and \( |B_q(n,k)| = a_q(2k,k) \left\lfloor \frac{n}{2k} \right\rfloor q^{(n)2k} \), hence

\[
a_q(n,k) \leq a_q(2k,k) \left\lfloor \frac{n}{2k} \right\rfloor q^{(n)2k}.
\]

Let \( b(k) \) be the number of vectors of length \( 2k \) with a zeros run of length exactly \( k \) and with no zeros run of length greater than \( k \). There are \( q^{2k-(k+1)}(q-1) \) vectors of length \( 2k \) that start with a zeros run of length exactly \( k \) and \( q^{2k-(k+1)}(q-1) \) different vectors that end with such a run. There are \( 2k - (k+1) - 2 = k-1 \) other positions in which a zeros run of length \( k \) can start within the \( 2k \) vector. For each such a position there are \( q^{2k-(k+2)}(q-1)^2 \) different vectors with a zeros run of length exactly \( k \). In total we have

\[
b(k) = 2q^{2k-(k+1)}(q-1) + (k-1)q^{2k-(k+2)}(q-1)^2 \\
= 2q^{k-1}(q-1) + (k-1)q^{k-2}(q-1)^2 \\
\geq (k+1)q^{k-2}(q-1)^2.
\]

All length-\( 2k \) vectors with a zeros run of length exactly \( k \) are not included in

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14
\[ A_q(2k, k). \] Therefore,
\[ a_q(2k, k) \leq q^{2k} - b(k) \leq q^{2k} - (k + 1)q^{k-2}(q - 1)^2. \]

(3.2)

By combining inequalities (3.1) and (3.2), we get
\[
\begin{align*}
a_q(n, k) &\leq (q^{2k} - (k + 1)q^{k-2}(q - 1)^2) \left\lfloor \frac{n}{2k} \right\rfloor q^{(n)2k} \\
&= (q^{2k}(1 - \frac{(k + 1)(q - 1)^2}{q^{k+2}})) \left\lfloor \frac{n}{2k} \right\rfloor q^{(n)2k} \\
&= q^n(1 - \frac{(k + 1)(q - 1)^2}{q^{k+2}}) \left\lfloor \frac{n}{2k} \right\rfloor \\
\leq q^n(\left( e^{-\frac{(k+1)(q-1)^2}{q^{k+2}q}} \right) \left\lfloor \frac{n}{2k} \right\rfloor \\
\leq q^n(\frac{\log_q e^{\frac{(k+1)(q-1)^2}{q^{k+2}}}}{2k}) \\
\end{align*}
\]

where (a) results from the inequality \( 1 - x \leq e^{-x} \) for all \( x \).

Part 2: Lower Bound

We start by presenting an upper bound on the number of length-\( n \) vectors with a zeros run of length at least \( k \). There are \( n - k + 1 \) positions in which a zeros run can start and for each position we have at most \( q^{n-k} \) different vectors. From the union bound we have that the number of length-\( n \) vectors with a zeros run of length at least \( k \) is upper bounded by \( q^{n-k}(n - k + 1) \), and by excluding those vectors from the set of all length-\( n \) vectors we get
\[ a_q(n, k) \geq q^n - q^{n-k}(n - k + 1). \]
This bound is irrelevant for values of $k$ smaller than $\log_q n$ as it is less than zero. However, if we choose $k \geq \log_q n + 1$ it becomes useful and we get

$$a_q(n,k) \geq q^n - q^{n-\log_q n}(n - \log_q n)$$

$$= q^n(1 - \frac{n - \log_q n}{qn})$$

$$\geq q^n(1 - \frac{n}{qn})$$

$$= q^{n-1}(q - 1).$$

Since we are also interested in the case of $k < \log_q n + 1$, or equivalently $n > q^{k-1}$, we continue our analysis by breaking down the length-$n$ vector to blocks of length $q^{k-1}$ and applying the bound on them in the following manner.

For $\ell \geq k$ we denote the set

$$C_q(\ell, k) = \{ a \mid a \in A_q(\ell, k), \ a \text{ starts or ends with } \left\lceil \frac{k}{2} \right\rceil \text{ zeros} \}$$

and $c_q(\ell, k) = |C_q(\ell, k)|$. Note that

$$c_q(\ell, k) \leq 2 \cdot q^{\ell - \left\lfloor k/2 \right\rfloor}.$$  

We also denote $D_q(\ell, k) = A_q(\ell, k) \setminus C_q(\ell, k)$ and

$$d_q(\ell, k) = |D_q(\ell, k)| = a_q(\ell, k) - c_q(\ell, k)$$

$$\geq a_q(\ell, k) - 2q^{\ell - \left\lceil k/2 \right\rceil}.$$  

For $k \geq \log_q \ell + 1$, or equivalently $\ell \leq q^{k-1}$ we have that

$$d_q(\ell, k) \geq q^{\ell - 1}(q - 1) - 2q^{\ell - \left\lceil k/2 \right\rceil}$$

$$= q^{\ell - 1}(q - 1)(1 - \frac{2q}{(q - 1)q^{\left\lceil k/2 \right\rceil}}).$$  

Consider the set of vectors which are a concatenation of $\left\lfloor \frac{n}{q^{k-1}} \right\rfloor$ vectors from
$D_q(q^{k-1}, k)$ appended by a vector from $D_q(\langle n \rangle_{q^{k-1}}, k)$. We denote this set by

$$E_q(n, k) = D_q(q^{k-1}, k)^{\lfloor \frac{n}{q^{k-1}} \rfloor} D_q(\langle n \rangle_{q^{k-1}}, k).$$

Note that $E_q(n, k) \subseteq A_q(n, k)$ and

$$|E_q(n, k)| = d_q(q^{k-1}, k)^{\lfloor \frac{n}{q^{k-1}} \rfloor} d_q(\langle n \rangle_{2^{k-1}}, k).$$

Hence,

$$a_q(n, k) \geq d_q(q^{k-1}, k)^{\lfloor \frac{n}{q^{k-1}} \rfloor} d_q(\langle n \rangle_{q^{k-1}}, k)$$

$$\geq (q^{q^{k-1}-1} (q - 1)(1 - \frac{2q}{(q - 1)q^{[k/2]}}))^{\lfloor \frac{n}{q^{k-1}} \rfloor}$$

$$= q^n (\frac{q - 1}{q})^{\lfloor \frac{n}{q^{k-1}} \rfloor + 1} (1 - \frac{2q}{(q - 1)q^{[k/2]}})^{\lfloor \frac{n}{q^{k-1}} \rfloor + 1}.$$

For all $x < -1$, it is known that $(1 + \frac{1}{x})^{x+1} < e$. We denote $x = -\frac{(q-1)q^{[k/2]}}{2q} = -c_1 q^{\frac{k}{2}}$, for some constant $c_1 > 0$. For $k \geq 5, q \geq 2$ we have that $x < -1$ and we deduce that

$$(1 - \frac{2q}{(q - 1)q^{[k/2]}})^{\lfloor \frac{n}{q^{k-1}} \rfloor + 1} = (1 + \frac{1}{x})^{(x+1)((\lfloor \frac{n}{q^{k-1}} \rfloor + 1)/(x+1))}$$

$$> e^{((\lfloor \frac{n}{q^{k-1}} \rfloor + 1)/(x+1))}$$

$$= e^{(c_2 \frac{n}{q^{k-1}})/(-c_1 q^{\frac{k}{2}})}$$

$$= e^{-\frac{c_2}{c_1} \frac{n}{q^{\frac{k}{2}}}},$$

where (a) follows from a choice of $c_2 > 0$ as the constant which satisfies $c_2 \frac{n}{q^{k-1}} = \lfloor \frac{n}{q^{k-1}} \rfloor + 1$. Finally, we conclude that

$$a_q(n, k) \geq q^n (\frac{q - 1}{q})^{c_2 \frac{n}{q^{k-1}}} e^{-\frac{c_2}{c_1} \frac{n}{q^{\frac{k}{2}}}}.$$
From Lemma 3.1 we have that
\[
\frac{c_3 n - 2k}{q^k} \leq \text{red}(A_q(n, k)) \leq \log\left(\frac{q}{q-1}\right) c_2 \frac{n}{q^{k-1}} + \frac{\log(\epsilon) c_2}{c_1} \frac{n}{q^{1.5k-1}}.
\]

If \( n - 2k = \Theta(n) \) the lower and upper bounds are of the same order, meaning there exist constants \( C_1, C_2 > 0 \) such that for large enough \( n \)
\[
C_1 \frac{n}{q^k} \leq \text{red}(A_q(n, k)) \leq C_2 \frac{n}{q^k},
\]
and the next corollary follows.

**Corollary 3.2.** Let \( f(n) \) be a function such that \( n - 2(\log_q n - f(n)) = \Theta(n) \) and \( \log_q n - f(n) \) is a positive integer. Then, the redundancy of \( A_q(n, \log_q n - f(n)) \) is \( \Theta(q^{f(n)}) \).

Note that \( f(n) \) can be negative. The result from Corollary 3.2 provides us with a general understanding on how the redundancy of the set \( A_q(n, k) \) behaves for different values of \( k \). We can conclude for example that for \( k = 0.5 \log_q n \) the redundancy is \( \Theta(\sqrt{n}) \). Motivated by the MU problem, we are interested in further exploring the case in which the redundancy is constant. According to the result from Lemma 3.1 and Corollary 3.2, this holds when \( f(n) \) is a constant function. That is, we are interested in the asymptotic behavior of \( a_q(n, \lceil \log_q n \rceil + z), z \in \mathbb{Z} \) up to a better precision than the one suggested in Lemma 3.1. For this purpose, we provide in the next Lemma additional results on the asymptotic behavior of \( a_q(n, k) \).

**Lemma 3.3.** Let \( n, k \) be integers such that \( 0 < k \leq n \). Then,
\[
2^{nE_{k,q}} \leq a_q(n, k) \leq 2^{nE_{k,q}} + 2q^{n - \lceil k/2 \rceil}.
\]

**Proof** Part 1: Upper Bound
We use the notations \( D_q(n, k), d_q(n, k) \) from the proof of Lemma 3.1 and consider the set \( G(m) = D_q(n, k)^m \). We also denote \( g(m) = |G(m)| = d_q(n, k)^m \). Note
that the set $G(m)$ satisfies the $k$-RLL constraint, and hence
\[
\lim_{m \to \infty} \frac{\log(g(m))}{nm} \leq E_{k,q}.
\]
We therefore get
\[
\lim_{m \to \infty} \frac{\log(d_q(n,k)^m)}{nm} = \frac{\log(d_q(n,k))}{n} \leq E_{k,q}.
\]
By using Equation (3.3) we have
\[
\frac{\log(a_q(n,k) - 2q^{n-[k/2]})}{n} \leq E_{k,q},
\]
and
\[
a_q(n,k) \leq 2^{nE_{k,q}} + 2q^{n-[k/2]}.\]

**Part 2: Lower Bound**

For all positive integer $m$ we have $A_q(mn,k) \subseteq A_q(n,k)^m$ and therefore we deduce that
\[
E_{k,q} = \lim_{m \to \infty} \frac{\log(a_q(mn,k))}{mn} \leq \lim_{m \to \infty} \frac{\log(a_q(n,k)^m)}{mn} = \frac{\log(a_q(n,k))}{n},
\]
and the lower bound follows directly.

For some values of $k$ the result from Lemma 3.3 gives weaker bounds than the one presented in Lemma 3.1. However, for the case of $k = \lceil \log_q n \rceil + z, z \in \mathbb{Z}$, we next show that the lower and upper bound of Lemma 3.3 are asymptotically tight.

Recall that Jain et. al showed in [20] that
\[
E_{k,q} = \log q - \frac{(q - 1) \log e}{q^{k+2}}(1 + o(1)) \quad (3.4)
\]
or in other words,
\[
\lim_{k \to \infty} \frac{(q - 1) \log eq^{-k-2}}{\log q - E_{k,q}} = 1. \tag{3.5}
\]

According to this result and the properties proved in Proposition 3.4 and Lemma 3.5 we conclude in Lemma 3.6 what the asymptotic behavior of \(2^{nE_{k,q}}\) is. Then, in Theorem 3.7 we show that \(2^{n-[k/2]+1}\) is negligible relatively to \(2^{nE_{k,q}}\), therefore the bounds in Lemma 3.3 meet and the asymptotic behavior of \(a_q(n,k)\) is established.

The following proposition will be in use in the proof of Lemma 3.6 and its proof is given in Appendix A.

**Proposition 3.4.** Let \(f(n), g(n)\) be functions such that \(\lim_{n \to \infty} g(n) = 1\) and \(1 \leq f(n) \leq C\) for a constants \(C\). Then,

\[
f(n)^{g(n)} \approx f(n).
\]

Note that the requirement that \(f(n)\) is bounded is essential, otherwise the proposition does not hold. For example if \(f(n) = 2^n\) and \(g(n) = 1 + \frac{1}{n}\) we have that \(\lim_{n \to \infty} \frac{f(n)^{g(n)}}{f(n)} = 2\).

**Lemma 3.5.** There exists an integer \(N\) such that for all \(n \geq N\) and \(k = \lceil \log_q n \rceil + z, z \in \mathbb{Z}\)

\[
\log q - \frac{C}{n} \leq E_{k,q} \leq \log q,
\]

for some constant \(C > 0\) which is independent of \(n\).

**Proof** First, \(E_{k,q} \leq \log q\) from the definition of \(E_{k,q}\). From Corollary 3.2 there exists a constant \(C'\) such that for large enough \(n\)

\[
\text{red}(A_q(n, \lceil \log_q n \rceil + z)) \leq C',
\]

and equivalently \(a_q(n, \lceil \log_q n \rceil + z) \geq q^{n-C'}\). From Lemma 3.3

\[
E_{k,q} \geq \frac{\log(a_q(n, k) - 2q^{n-[k/2]})}{n} \\
\geq \frac{\log(q^{n-C'} - 2q^{n-[k/2]})}{n} \\
= \frac{n \log q + \log(q^{-C'} - 2q^{-[k/2]})}{n}.
\]
There exists an integer $N$ such that for all $n \geq N$, $2q^{-\lceil k/2 \rceil} \leq 0.5q^{-C'}$. Thus,

$$E_{k,q} \geq \frac{n \log q + \log(q^{-C'} - 0.5q^{-C'})}{n}$$
$$= \frac{n \log q + \log(0.5q^{-C'})}{n}$$
$$= \log q - \frac{1 + C' \log q}{n}$$

and by choosing $C = 1 + C' \log q$ the result follows.

We are now ready to use the result of Jain et al. mentioned in Equation (3.5), in order to establish the following result.

**Lemma 3.6.** For $k = \lceil \log_q n \rceil + z, z \in \mathbb{Z}$,

$$2^n E_{k,q} \approx \frac{q^n}{e^{(q-1)q^\Delta n-z-1}},$$

where $\Delta_n = \log_q n - \lceil \log_q n \rceil$.

**Proof.** We use Proposition 3.4 with $f(n) = 2^n(\log q - E_{k,q})$ and $g(n) = \frac{(q-1) \log eq^{-k-2}}{\log q - E_{k,q}}$.

From Lemma 3.5, when $k = \lceil \log_q n \rceil + z$,

$$1 \leq 2^n(\log q - E_{k,q}) = f(n) \leq 2^C.$$

From Eq. (3.5) $\lim_{n \to \infty} g(n) = 1$ thus the requirements of the proposition are satisfied and we get

$$2^n(\log q - E_{k,q}) \approx 2^n(q-1) \log eq^{-(k-1)^2}.$$

We conclude that

$$2^n E_{k,q} = 2^n \log q + n(\log q - E_{k,q})$$
$$\approx 2^n \log q - n(q-1) \log eq^{-k-1}$$
$$\approx 2^n \log q - n(q-1) \log eq^{\lceil \log_q n \rceil-z-1}$$
$$\approx \frac{q^n}{e^{(q-1)q^\Delta n-z-1}},$$
where $\Delta_n = \log_q n - \lceil \log_q n \rceil$.

Finally, we can now apply the result of Lemma 3.6 in Lemma 3.3 and obtain the asymptotic behavior of $a_q(n, \lceil \log_q n \rceil + z)$ in the following theorem.

**Theorem 3.7.** For $k = \lceil \log_q n \rceil + z$, $z \in \mathbb{Z}$,

$$a_q(n, k) \approx \frac{q^n}{e^{(q - 1)q^{\Delta_n - z - 1}}},$$

where $\Delta_n = \log n - \lceil \log n \rceil$.

**Proof.** Lemma 3.3 gives us

$$1 \leq \frac{a_q(n, k)}{2^{nE_{k,q}}} \leq 1 + \frac{2q^{n - \lceil k/2 \rceil}}{2^{nE_{k,q}}}.$$

By using Lemma 3.6 we get that for $k = \lceil \log_q n \rceil + z$

$$\lim_{n \to \infty} \frac{2q^{n - \lceil k/2 \rceil}}{2^{nE_{k,q}}} = \lim_{n \to \infty} \frac{2q^{n - \lceil k/2 \rceil}}{q^n e^{-(q-1)q^{\Delta_n - z - 1}}} = 0,$$

and conclude that

$$1 \leq \lim_{n \to \infty} \frac{a_q(n, k)}{2^{nE_{k,q}}} \leq \lim_{n \to \infty} 1 + \frac{2q^{n - \lceil k/2 \rceil}}{2^{nE_{k,q}}} = 1.$$

Therefore,

$$a_q(n, k) \approx 2^{nE_{k,q}},$$

and together with Lemma 3.6 the result follows directly.

**Remark 1.** We would like to note that the results in this section match the results by Schilling [33] on the distribution of the longest runs in arbitrary vectors. In particular, it is stated in [33] that the typical length of the longest run in $n$ flips of a fair coin converges to $\log n - 1$. However, we find the proof in this section to be more accurate for the purpose of exactly calculating the number of redundancy bits of $k$-RLL codes.
3.1.2 Efficient Encoding and Decoding Algorithms

In this subsection we provide an algorithm to efficiently encode and decode vectors which avoid zeros run of length $\lceil \log_q n \rceil + 1$. We show the algorithm for the binary case, however, it is straightforward to extend it for $q > 2$. According to Theorem 3.7 the redundancy of $A_2(n, \log n + 1)$ is approximately $\frac{\log e}{4} \approx 0.36$ when $n = 2^i, i \in \mathbb{N}$. The algorithm described next uses one redundancy bit, however, it has linear encoding and decoding complexities.

**Algorithm 1** Zero Run-Length Encoding

**Input:** Sequence $x \in \Sigma^{n'}$, $n' \leq n$

**Output:** $y \in \Sigma^{n'+1}$ with zeros run length $\leq \lceil \log n \rceil$

1: Define $y = x1 \in \Sigma^{n'+1}$

2: Set $i = 1$ and $i_{\text{end}} = n'$

3: while $i \leq i_{\text{end}} - \lceil \log n \rceil$ do

4: if $w_H(y_i^{i + \lceil \log n \rceil}) = 0$ then

5: remove the zeros run $y_i^{i + \lceil \log n \rceil}$ from $y$

6: $p(i)$: binary representation of $i$ with $\lceil \log n \rceil$ bits

7: append $p(i)0$ to the right of $y$

8: set $i_{\text{end}} = i_{\text{end}} - \lceil \log n \rceil - 1$

9: else

10: set $i = i + 1$

11: end if

12: end while

The following lemma proves the correctness of Algorithm 1.

**Lemma 3.8.** For all $n' \leq n$, given any vector $x \in \Sigma^{n'}$, Algorithm 1 outputs a sequence $y \in \Sigma^{n'+1}$, where any zeros run has length at most $\lceil \log n \rceil$ and such that $x$ can be uniquely reconstructed given $y$. Furthermore, the time and space complexity of the algorithm and its inverse is $\Theta(n)$.

**Proof** The algorithm starts by initializing $y = x1$. We then iterate over the indices of $y$ that correspond to the indices of the input word $x$. If we encounter an index in which a $\lceil \log n \rceil + 1$ zeros run starts, we remove the run and append $p(i)0$, which we call a pointer, to the right of $y$, where $p(i)$ is the binary representation of the index $i$.

First, notice that each appended pointer $p(i)0$ has the same length as the corresponding removed run. Therefore throughout the algorithm the length of $y$ does
not change. There exists an index \( 1 \leq t \leq n' \) such that the output \( y \) is of the form 
\[ y_1^t y_{t+2}^{n'+1} \]
where \( y_1^t \) is the remainder of \( x \) after removing the zeros runs and \( y_{t+2}^{n'+1} \) is the list of the pointers \((p(i), 0)\) representing the indices of the removed zeros runs.

To reconstruct \( x \) given \( y \) we start by locating the separating bit 1 on position \( t \). We start from the right and check whether the rightmost bit is 1 or 0. In case it is 0, the \( \lceil \log n \rceil \) bits to the left correspond to a pointer, we skip them and repeat the process until we encounter the separating 1. We then construct the original \( x \) by inserting zeros runs of length \( \lceil \log n \rceil + 1 \) to the remainder part \( y_1^t \) according to the pointers part \( y_{t+2}^{n'+1} \).

Next, we show that \( y \) does not contain a zeros run of length greater than \( \lceil \log n \rceil \). It is clear that \( y_1^t \) does not contain such a run. The separating 1 ensures that there is no zeros run which starts in \( y_1^t \) and ends in \( y_{t+2}^{n'+1} \). The structure of \( y_{t+2}^{n'+1} \) is a sequence of concatenated pointers of the form \( p(i)0 \). It suffices to show that any sub-vector of \( y_{t+2}^{n'+1} \) of the form \( p(j)0p(k) \) does not consist a zeros run of length greater than \( \lceil \log n \rceil \). Here, \( j \) and \( k \) represent the indices of two consecutive locations where a \( \lceil \log n \rceil + 1 \) zeros run was found in the while loop in the algorithm and note that \( 0 < j \leq k \).

We consider the leftmost one in the binary representations \( p(j) \) and \( p(k) \). Since \( j \leq k \), the position of the leftmost one within \( p(j) \) is smaller or equal to the position of the leftmost one within \( p(k) \), thus any window of length \( \lceil \log n \rceil + 1 \) must contain at least one of the leftmost ones from \( p(j) \) and \( p(k) \).

Note that since the indices of the input vector \( x \) are indexed starting from 1, we do not write 0 in any of the pointers. We do not write \( n' \) either since the largest index we consider for zeros runs is \( n' - \lceil \log n \rceil \). Lastly, the complexity of the algorithm is \( \mathcal{O}(n) \) since the complexity of each pointer update when encountering a zeros run is \( \Theta(\log n) \) and the number of these operations is at most \( n'/\log n \).

The following example demonstrates how the encoding in Algorithm 1 works.

**Example 1.** Let \( n' = n = 13 \) and therefore \( \lceil \log n \rceil = 4 \) and \( \lceil \log n \rceil + 1 = 5 \). Consider the following sequence:

\[ x = 1000000000001, \]

Let us go through the steps of Algorithm 1.

1. \( y = x1 = 10000000000011 \)
2. \( i = 1 \) and \( i_{\text{end}} = 13 \)

3. \( i = 1: \ w_H(y_{1}^5) \neq 0, \ i = i + 1 \)

4. \( i = 2: \ w_H(y_{2}^5) = 0, \)
   
   (a) Remove \( y_{2}^5 \) from \( y : \ y = 100000011 \)
   
   (b) Define \( p(2) = 0010 \)
   
   (c) Append \( p(2)0 = 00100 : \ y = 10000001100100 \)
   
   (d) Set \( i_{\text{end}} = 13 - 5 = 8 \)

5. \( i = 2: \ w_H(y_{2}^5) = 0, \)
   
   (a) Remove \( y_{2}^5 \) from \( y : \ y = 101100100 \)
   
   (b) Define \( p(2) = 0010 \)
   
   (c) Append \( p(2)0 = 00100 : \ y = 10110010000100 \)
   
   (d) Set \( i_{\text{end}} = 8 - 5 = 3 \)

6. \( i = 2: \ w_H(y_{2}^5) \neq 0, \ i = i + 1. \)

The decoding works as described in the proof of Lemma 3.8.

A similar algorithm to the problem solved by Algorithm 1 was recently proposed in [34] to efficiently encode sequences that do not contain runs of zeros and ones of length \( k \) with a single redundancy bit. Specifically, in [34] the authors showed how to accomplish this task with \( k = \lfloor \log n \rfloor + 4 \). Algorithm 1 can be slightly adjusted in order to solve the problem in [34] with \( k = \lfloor \log n \rfloor + 2 \). Lastly, we note that Kauts presented in [22] an algorithm which encodes all words avoiding zeros runs of any specific length with optimal redundancy. However, the space complexity of this algorithm is \( \Theta(n^2) \).

Algorithm 1 solves the problem of avoiding zeros runs of length \( \lceil \log n \rceil + 1 \). However, it can be extended to avoid zeros runs of any length in the following manner. Assume for example that we want to avoid zeros runs of length \( 0.5 \log n \).

We start with a length-\( n \) vector, divide it into blocks of length \( \sqrt{n} \) and apply Algorithm 1 on each block since \( 0.5 \log n = \log(\sqrt{n}) + 1 \). We append 1 to each of the output vectors and the final resulting vector is the concatenation of all of them.

This approach can be applied for any value of \( k \) and it achieves optimal order of redundancy, with linear encoding and decoding complexities.
3.2 Cardinality of MU Code Construction

In this section we expand the study of MU codes. Specifically, we show that the lower bounds in (2.2), (2.3) are asymptotically tight.

We are interested in maximizing the value of $|C_1(n, q, k)|$ over all values of $k$. Notice that

$$|C_1(n, q, k)| = (q - 1)^2 a_q(n - k - 2, k),$$

and therefore the proof of the main theorem in this section highly relies on the analysis in Section 3.1. However, in Section 3.1 we analyzed the asymptotic size $a_q(n, k)$ while we are actually interested in $a_q(n - k - 2, k)$. We chose to present the analysis of $a_q(n, k)$ in Section 3.1 since it is similar to the analysis of $a_q(n - k - 2, k)$ and we believe it will be of use in other problems as well. In this section we complete some missing parts to obtain the asymptotic behavior of the maximal value of $|C_1(n, q, k)| = (q - 1)^2 a_q(n - k - 2, k)$. We mainly focus on the trade-off formed by the choice of $k$: on one hand, reducing the value of $k$ results with a larger value for $n - k - 2$ and thus smaller redundancy due to the fixed $k$-length prefix in the construction. On the other hand, smaller $k$ implies a stronger $k$-RLL constraint which requires larger redundancy in the remaining part.

We start by showing that it is sufficient to look only into values of $k$ which are of the form $\lceil \log_q n \rceil + z, z \in \mathbb{Z}$.

**Lemma 3.9.** The value of $k$ that minimizes the redundancy of $C_1(n, q, k)$ is $\lceil \log_q n \rceil + z, z \in \mathbb{Z}$.

**Proof** From Corollary 3.2, in the case of $k = \lceil \log_q n \rceil + z, z \in \mathbb{Z}$ the redundancy of $C_1(n, q, k)$ is $k + \Theta(1) = \lceil \log_q n \rceil + z + \Theta(1)$. Since the redundancy of $C_1(n, q, k)$ is at least $k$, greater values of $k$ i.e., $k = \lceil \log_q n \rceil + \omega(1)$, lead to higher redundancy, and so we disregard them. For values $k$ of the form $\log_q n - f(n)$, where $f(n) = \omega(1)$ we again turn to Corollary 3.2 to claim that the redundancy is $\Theta(k + 2f(n)) = \Theta(\log_q n - f(n) + 2f(n))$ which is also asymptotically greater than $\lceil \log_q n \rceil + z + \Theta(1)$. We therefore summarize that the minimal redundancy, or the maximal cardinality of $C_1(n, q, k)$ is achieved when $k$ is of the form $\lceil \log_q n \rceil + z, z \in \mathbb{Z}$.

We next look further into the specific choice of $k$ that maximizes the cardinality and obtain the asymptotic behavior of the maximal cardinality.
For the rest of this section we denote \( n' = n - k - 2 \). From Lemma 3.3 we have that
\[
2^{n' E_{k,q}} \leq a_q(n', k) \leq 2^{n' E_{k,q}} + 2q^{n' - \lfloor k/2 \rfloor}.
\] (3.6)

In the following lemma we establish how \( 2^{n' E_{k,q}} \) behaves asymptotically for the values of \( k \) of interest to us. The proofs of Lemma 3.10 and Lemma 3.11 are attached in Appendix B as they share similar ideas to the proofs in Section 3.1.

**Lemma 3.10.** For \( k = \lceil \log_q n \rceil + z, z \in \mathbb{Z} \),
\[
2^{n' E_{k,q}} \approx \frac{q^n}{n} \cdot \frac{q^{\Delta_n - z - 2}}{e^{(q-1)q^{\Delta_n - z - 1}}},
\]
where \( \Delta_n = \log_q n - \lceil \log_q n \rceil \).

We apply the result of Lemma 3.10 in the inequality (3.6) to obtain the asymptotic behavior of \( a_q(n', \lceil \log n \rceil + z) \) and establish the following

**Lemma 3.11.** For \( z \in \mathbb{Z} \),
\[
|C_1(n, q, \lceil \log n \rceil + z)| \approx \frac{q^n}{n} \cdot \left(\frac{q-1}{q}\right)^2 q^{\Delta_n - z - \log_q e(q-1)q^{\Delta_n - z - 1}}
\]
where \( \Delta_n = \log_q n - \lceil \log_q n \rceil \).

Next we optimize this term over all values of \( z \in \mathbb{Z} \) in order to establish the maximal cardinality \( C_1(n, q) = \max_{1 \leq k < n} \{|C_1(n, q, k)|\} \).

**Theorem 3.12.**
\[
C_1(n, q) \approx \frac{q^n}{n} \cdot \left(\frac{q-1}{q}\right)^2 q^{F(\Delta_n)} \leq \frac{q^n}{n} \cdot \frac{q-1}{eq},
\]
where \( \Delta_n = \log_q n - \lceil \log_q n \rceil \) and
\[
F(\Delta_n) = \max_{z \in \{-2, -1, 0\}} \{\Delta_n - z - \log_q e(q-1)q^{\Delta_n - z - 1}\}.
\]
The inequality is tight when \( n \to \infty \) over any subsequence of \( n \) that satisfies \( \Delta_n = - \log_q (q-1) \).
Proof From Lemma 3.11, when \( k = \lceil \log_q n \rceil + z \) we get

\[
|C_1(n, q, k)| \approx \frac{q^n}{n} \left( \frac{q-1}{q} \right)^2 q^{\Delta_n - z - \log_q e (q-1) q^{\Delta_n - z - 1}}.
\]

Let us denote

\[
f(\Delta_n, z) = \Delta_n - z - \log_q e (q-1) q^{\Delta_n - z - 1}.
\]

We are interested in the value of \( z \in \mathbb{Z} \) that maximizes the size of \( C_1(n, q, k) \), that is, the value of \( z \in \mathbb{Z} \) that maximizes \( f(\Delta_n, z) \) for each \( \Delta_n \).

\[
\frac{\partial f}{\partial z} = -1 + (q-1) q^{\Delta_n - z - 1},
\]

so the only maximum of the function \( f(\Delta_n, z) \) is achieved for \( z_0 = \log_q (q-1) - 1 + \Delta_n \). However, \( z_0 \) is not necessarily an integer while we are interested in integers only. Since \( -1 < \Delta_n \leq 0 \) we have \( -2 < z_0 < 0 \) thus the maximum over \( z \in \mathbb{Z} \) is achieved by one of the options \( z \in \{-2, -1, 0\} \).

We therefore obtain the following:

\[
C_1(n, q) \approx \frac{q^n}{n} \left( \frac{q-1}{q} \right)^2 q^{F(\Delta_n)},
\]

where \( \Delta_n = \log_q n - \lceil \log_q n \rceil \) and

\[
F(\Delta_n) = \max_{z \in \{-2, -1, 0\}} \{ \Delta_n - z - \log_q e (q-1) q^{\Delta_n - z - 1} \}.
\]

In Appendix B we analyze how \( F(\Delta_n) \) behaves for each value of \( \Delta_n \) and obtain that when \( q = 2 \)

\[
F(\Delta_n) = \begin{cases} 
    f(\Delta_n, -2), & \text{for } -1 < \Delta_n \leq \log(\ln 2) \\
    f(\Delta_n, -1), & \text{otherwise} 
\end{cases}
\]

and when \( q > 2 \)

\[
F(\Delta_n) = \begin{cases} 
    f(\Delta_n, -1), & \text{for } -1 < \Delta_n \leq \delta_0 \\
    f(\Delta_n, 0), & \text{otherwise} 
\end{cases}
\]

28
for \( \delta_0 = -\log_q \frac{(q-1)^2}{q \ln q} \). We also discuss in Appendix B the maximal value of \( F(\Delta_n) \) which leads to the maximal cardinality \( \frac{q^n}{n} \cdot \frac{q-1}{q^z} \).

The result of Theorem 3.12 aligns with the results from [11, 16, 25] which we recalled in Equations (2.2) and (2.3). The lower bound from (2.2) states that

\[
C_1(n, q) \geq q^{-\frac{q}{q-1}} \ln q \frac{q^n}{n} \quad (2.2)
\]

and it holds when \( n \to \infty \) over any series of \( n \). The lower bound (2.3) claims that

\[
C_1(n, q) \geq \frac{q-1}{qe} \cdot \frac{q^n}{n} \quad (2.3)
\]

and it holds when \( n \to \infty \) over the subseries \( \frac{q^i}{q-1}, i \in \mathbb{N} \). Observe that when \( n = \frac{q^i}{q-1} \), \( \lim_{n \to \infty} \Delta_n = -\log_q (q - 1) \) thus the result of Theorem 3.12 and the bound (2.3) agree.

The bounds in (2.2) and (2.3) provide lower asymptotic bounds on the cardinality \( C_1(n, q) \), while in Theorem 3.12 we obtain the explicit expression of the asymptotic behavior for any value of \( \Delta_n \) and show that the lower bounds are asymptotically tight. According to Theorem 3.12 the term

\[
|C_1(n, q, \lceil \log_q n \rceil + z)| \cdot \frac{n}{q^n}
\]

depends only on \( z, q \) and \( \Delta_n = \log_q n - \lceil \log_q n \rceil \). Specifically it does not grow with \( n \). In Fig. 3.1 we plot this term for the three interesting values of \( z \in \{-2, -1, 0\} \) as a function of \( \Delta_n \), alongside the lower bounds from (2.2) and (2.3). The first, second graph in Fig. 3.1 corresponds to \( q = 2, q = 5 \), respectively. When \( q = 5 \) for example, the maximal cardinality is achieved when \( z = -1 \) or when \( z = 0 \), depending on the value of \( \Delta_n \). Among the range of values of \( \Delta_n \), \( \delta_0 = -\log_5(\frac{16}{5\ln 5}) \approx -0.43 \) yields the minimal cardinality which aligns with the bound (2.2). The maximal cardinality is achieved for \( \delta_1 = -\log_q (q - 1) = -\log_5 4 \approx -0.86 \) where it meets the bound (2.3), and the maximal cardinality is \( \frac{q-1}{qe} \cdot \frac{q^n}{n} \). So far we referred to \( q \) as a constant, however, an interesting fact to notice when \( q \to \infty \) and \( n \to \infty \), the maximal cardinality \( \frac{q-1}{qe} \cdot \frac{q^n}{n} \) approaches the upper bound from [28], \( \frac{q^n}{e(n-1)} \). This is illustrated in Fig. 3.2. Having said that, note that in the binary case there is a gap of factor 2

29
between the maximal cardinality and the upper bound and closing this gap remains an open problem.
Figure 3.1: Comparison between the construction’s cardinality according to Lemma 3.11, multiplied by $\frac{n}{q}$ for different values of $z$, and the bounds (2.2) and (2.3) from [11, 16, 25].
Figure 3.2: A comparison between the construction’s maximal cardinality according to Theorem 3.12 and the upper bound of the cardinality of MU codes from [28], mentioned in (2.1).
Since substitution errors may also happen both in the writing and reading processes of DNA molecules, we impose in this chapter a stronger and more general version of the mutual uncorrelatedness constraint, by requiring the prefixes and suffixes to not only differ but also have a large Hamming distance.

**Definition 4.1.** A code $C \subseteq \Sigma_q^n$ is called a $(d_h, d_m)$-MU code if

1. the minimum Hamming distance of the code is $d_h,$

2. for every two not necessarily distinct words $a, b \in C,$ and $i \in [n - 1]:$
$$d_H(a_i^{1}, b_{n-i+1}^{n}) \geq \min\{i, d_m\}.$$ 

### 4.1 Window Weight Limited Constraint

In this section we introduce a natural extension to the RLL constraint which we call the window weight limited constraint. We study this constraint for the purpose of constructing a new family of $(d_h, d_m)$-Mutually Uncorrelated Codes, which will be presented later in Section 4.2. Our main results in this section are an upper bound on the size of the set of vectors satisfying the window weight limited constraint and a construction with efficient encoding and decoding, and almost optimal cardinality. We start with the definition of this constraint.
Definition 4.2. Let \( d \) and \( k \) be positive integers. We say that a vector \( \mathbf{a} \in \Sigma_q^n \) satisfies the \((d, k)\)-window weight limited (WWL) constraint, and is called a \((d, k)\)-WWL vector, if \( n < k \) or \( \forall i \in [n - k + 1] : w_H(\mathbf{a}_i^{i+k-1}) \geq d \).

We call a set of \((d, k)\)-WWL vectors a \((d, k)\)-WWL code and we denote by \( A_q(n, k, d) \) the set of all \((d, k)\)-WWL vectors in \( \Sigma_q^n \). Lastly, \( a_q(n, k, d) = |A_q(n, k, d)| \).

This constraint states that a vector \( \mathbf{a} \in \Sigma_q^n \) is a \((d, k)\)-WWL vector if the Hamming weight of every consecutive subsequence of length \( k \) in \( \mathbf{a} \) is at least \( d \). A similar constraint was studied in [32] when studying time space codes for phase change memories. However, the authors of [32] studied the opposite constraint in which the weight of every window of length \( k \) is at most \( d \). Furthermore, as will be explained later, we are interested in the case where \( k \) depends on the word length \( n \) as opposed to the analysis in [32] where \( k \) is fixed. Notice that the \((1, k)\)-WWL constraint is equivalent to the \( k \)-RLL constraint. In the next lemma we provide an upper bound on the size of \( A_q(n, k, d) \). The proof is deferred to Appendix C since it shares similar ideas with the proof of the upper bound in Lemma 3.1.

Lemma 4.1. Let \( n, k, d \) be positive integers such that \( d \leq k \leq n \). Then, there exists a constant \( C > 0 \) such that for \( n \) large enough

\[
a_q(n, k, d) \leq q^{n-C(n-2k)^{d-1}/q^k}.
\]

For the rest of the paper we let \( \mathcal{F}(n,d) \) be

\[
\mathcal{F}(n,d) = \lceil \log_q n \rceil + (d - 1)(\lceil \log_q \lceil \log_q n \rceil \rceil + C) + 2,
\]

where \( C \) is a constant equal to the minimum integer such that \( \lceil \log_q \lceil \log_q n \rceil \rceil + C \geq \lceil \log_q (\mathcal{F}(n,d) + 2) \rceil \). From Lemma 4.1 we also have that

\[
C \frac{(n-2k)k^{d-1}}{q^k} \leq \text{red}(A_q(n, k, d)),
\]

where \( C > 0 \) is a constant. By setting \( k \) to be \( \mathcal{F}(n,d) - f(n), f(n) > 0 \) we get that the redundancy of \( A_q(n, \mathcal{F}(n,d) - f(n), d) \) is \( \Omega(q^{f(n)}) \).

Next we present an explicit algorithm for encoding and decoding \((d, \mathcal{F}(n,d))\)-WWL vectors with \( n' \leq n \) information bits and \( d \) redundancy symbols.

The next lemma proves the correctness of Algorithm 2. Its proof is deferred to Appendix C.

34
Algorithm 2 Window Weight Limited Encoding

**Input:** $x \in \Sigma_q^{n'}$ and an integer $d > 1$

**Output:** $(d, F(n, d))$-WWL vector $y \in \Sigma_{q^d}^{n'+d}$

1: Define $y = x1^d \in \Sigma_q^{n'+d}$
2: Set $i = 1$ and $i_{\text{end}} = n'$
3: while $i \leq i_{\text{end}} - F(n, d) + 1$ do
4:   if $w_H(y_{i+i-F(n,d)-1}) < d$ then
5:     remove $y_{i+F(n,d)-1}$ from $y$
6:     $p(i)$: $q$-ary representation of $i$ with $\lceil \log_q n \rceil$ symbols
7:     for $j = 1, \ldots, d - 1$ do
8:       if there are at least $j$ ones in $y_{i+F(n,d)-1}$ then
9:         $t(j)$: $q$-ary index of the $j$-th 1 in $y_{i=F(n,d)-1}$
10:            with $\lceil \log_q \lceil \log_q n \rceil \rceil + C$ symbols
11:        else
12:          $t(j) = 1^{\lceil \log_q \lceil \log_q n \rceil \rceil + C}$
13:        end if
14:     end for
15:   else
16:       set $i = i - F(n, d) + 1$
17:   end if
18: set $i = i + 1$
19: end while
Table 4.1: Redundancy summary of \(k\)-RLL and \((d, k)\)-WWL constraints

<table>
<thead>
<tr>
<th>Constraint</th>
<th>(k)-RLL</th>
<th>(\log_q e^{(q-1)^q^{\cdots 1}})</th>
<th>((d, k))-WWL</th>
</tr>
</thead>
<tbody>
<tr>
<td>Redundancy</td>
<td>(\Theta(\frac{q}{n^q}))</td>
<td>(\log_q e^{(q-1)^q^{\cdots 1}})</td>
<td>(\Theta\left(\frac{n^q}{q^n}\right))</td>
</tr>
<tr>
<td>Comments</td>
<td>(n-2k = \Theta(n)) (k = \lceil \log_q n \rceil + z), (z \in \mathbb{Z}, \Delta_n = \log_q n - \lceil \log_q n \rceil) (k = \log n + (d-1)\log \log n - \Omega(1))</td>
<td>(\Theta(\frac{n^q}{q^n}))</td>
<td>(\Theta(\frac{n^q}{q^n}))</td>
</tr>
<tr>
<td>Stated in</td>
<td>Corollary 3.3</td>
<td>Theorem 3.7</td>
<td>Corollary 4.3</td>
</tr>
</tbody>
</table>

Lemma 4.2. For all \(n' \leq n\), given any vector \(x \in \Sigma_q^{n'}\), Algorithm 2 outputs a \((d, F(n, d))\)-WWL vector \(y \in \Sigma_q^{n'+d}\) such that \(x\) can be uniquely reconstructed given \(y\). The time and space complexity of the algorithm and its inverse is \(\Theta(n)\).

Algorithm 2 solves the problem of WWL encoding by replacing each subvector of small weight with the index of the subvector (denoted by \(p(i)\)), appended by the indices of the ones within it (denoted by \(t(j)\)s). The algorithm works when the window’s length \(k\) is \(F(n, d) = \log_q n + (d-1)(\log_q (\log_q n) + \Theta(1))\), with a constant number of redundancy symbols. From Lemma 4.1, the redundancy in this case is at least \(\Theta(1)\), hence the redundancy of the algorithm is optimal up to a constant number of bits. Lastly, we can also extend this algorithm to encoding WWL vectors with smaller values of \(k\), that is \(k = F(n, d) - f(n)\), \(f(n) > 0\), by breaking down the length-\(n\) vector to \(q^{f(n)}\) blocks of length \(n/q^{f(n)}\), applying Algorithm 2 on each block, and stitching the output vectors with \(d\) ones between each two subvectors. This approach yields redundancy of approximately \(2dq^{f(n)}\), which is also optimal up to a constant factor, according to Lemma 4.1. The following corollary is established

Corollary 4.3. Let \(f(n)\) be a positive function such that \(F(n, d) - f(n)\) is a positive integer. The redundancy of \(A_q(n, F(n, d) - f(n), d)\) is \(\Theta(q^{f(n)})\).

We hereby summarize our study on the \(k\)-RLL and the \((d, k)\)-WWL constraints. Our main results are presented in Table 4.1. Note that the first column of the \(k\)-RLL constraint lists our result for all values of \(k\) such that \(n - 2k = \Theta(n)\) and provides the redundancy order in this case. On the other hand, the second column is targeted towards specific values of \(k\) and gives an exact value of the redundancy.

36
4.2  \((d_h, d_m)\)-MU Codes Cardinality Analysis

4.2.1 Upper Bound  

We set \(A_{MU}(n, q, d_h, d_m)\) to be the largest cardinality of a \((d_h, d_m)\)-MU code over \(\Sigma_q^n\), and \(M(n, q, d)\) as the largest cardinality of a length-\(n\) code over \(\Sigma_q^n\) with minimum Hamming distance \(d\). Motivated by Levenshtein’s upper bound [27], we first provide an upper bound on the value \(A_{MU}(n, q, d_h, d_m)\).

**Theorem 4.4.** For all positive integers \(n, q, d_h, d_m\),

\[
A_{MU}(n, q, d_h, d_m) \leq \frac{M(n, q, d)}{\lceil n/d_m \rceil},
\]

where \(d = \min\{d_h, 2d_m\}\).

**Proof** Let \(C \subseteq \Sigma_q^n\) be a \((d_h, d_m)\)-MU code. We let

\[
\hat{C} = \{(aa)^{i+1}_{i+1} | a \in C, i = \alpha \cdot d_m, \alpha \in [0, \lfloor n/d_m \rfloor - 1]\}.
\]

That is, the code \(\hat{C}\) consists of all cyclic shifts of words from \(C\) by \(\alpha d_m\) bits. For \(a, b \in C\) let \(\hat{a} = (aa)^{i+1}_{i+1}, \hat{b} = (bb)^{j+1}_{j+1}\) with \(i = \alpha_i d_m\) and \(j = \alpha_j d_m, \alpha_i, \alpha_j \in [0, \lfloor n/d_m \rfloor - 1]\). Note that \(\hat{a}, \hat{b} \in \hat{C}\). We prove that if \(i \neq j\) or \(a \neq b\), then \(d_H(\hat{a}, \hat{b}) \geq \min\{d_h, 2d_m\}\). First, if \(i = j\) and \(a \neq b\) we have that \(d_H(\hat{a}, \hat{b}) = d_H(a, b) \geq d_h\). Otherwise, \(i \neq j\) and we assume without loss of generality that \(i > j\). Notice that \(\tilde{a}_{n-i+1}^{n-j}\) is a prefix of \(a\) and \(\tilde{b}_{n-i+1}^{n-j}\) is a suffix of \(b\). Moreover, the length of \(\tilde{a}_{n-i+1}^{n-j}\) is at least \(d_m\), therefore, from the definition of \((d_h, d_m)\)-MU code, \(d_H(\tilde{a}_{n-i+1}^{n-j}, \tilde{b}_{n-i+1}^{n-j}) \geq d_m\). Similarly, \(\tilde{b}_{n-j+1}^{n-i}\) is a prefix of \(b\) and \(\tilde{a}_{n-j+1}^{n-i}\) is a suffix of \(a\), thus \(d_H(\tilde{a}_{n-j+1}^{n-i}, \tilde{b}_{n-j+1}^{n-i}) \geq d_m\). Therefore,

\[
d_H(\hat{a}, \hat{b}) = d(\tilde{a}_{n-i+1}^{n-j}, \tilde{a}_{n-j+1}^{n-i}, \tilde{b}_{n-i+1}^{n-j}, \tilde{b}_{n-j+1}^{n-i}) \geq 2d_m.
\]

We showed that \(d_H(\hat{a}, \hat{b}) \geq \min\{d_h, 2d_m\}\) and thus

\[
|\hat{C}| = \lfloor n/d_m \rfloor \cdot |C| \leq M(n, q, d),
\]

where \(d = \min\{d_h, 2d_m\}\), and the theorem follows directly.
According to the sphere packing bound we have that 
\[ M(n, q, d) \leq q^n / (Cn^{\frac{d}{2} - 1}) \]
for some constant \( C \), hence the minimum redundancy of any \((d_h, d_m)\)-MU code is
\[ \left\lceil \frac{d + 1}{2} \right\rceil \log_q n - \log_q d_m + \mathcal{O}(1). \]

### 4.2.2 Construction

We are now ready to show a construction of \((d_h, d_m)\)-MU codes. For the sake of simplicity, all the constructions presented in the rest of the paper are for the binary case.

We say that a vector \( u \in \Sigma^\ell \) is a \textit{\( d \)-auto-cyclic} vector if for every \( 1 \leq i \leq d \), \( d_H(u, 0^i u_1^{\ell-i}) \geq d \). In the next construction we use the following \( d \)-auto-cyclic vector \( u \) of length \( \ell(d) = d \lceil \log d \rceil + d \), which is given by \( u = 1^d u_0 \cdots u_{\lceil \log d \rceil - 1} \in \Sigma^{\ell(d)} \) such that \( u_i = ((1^d 0^2)^d)^i \). For example, if \( d = 5 \) we have \( \lceil \log d \rceil = 3 \) and \( u = 11111 \ 10101 \ 11001 \ 11110 \). \( u \) is \( d \)-auto-cyclic vector because for \( 1 \leq i \leq d \) there exist \( j \) odd and \( k \geq 0 \) such that \( i \equiv j \cdot 2^k \). \( d_H(u^1, 0^i) = i \) and \( d_H(u_{(u^1)}^{d-1}, u_{(u^2)}^{d+1}) = d - i \) so \( d_H(u, 0^i u^{\ell-i}) \geq d \).

**Construction II.** Let \( n, k \) be two integers such that \( k \geq \ell(d_m) \) and \( n \geq k + \ell(d_m) + 2d_m \). Denote \( n' = n - k - \ell(d_m) - 2d_m \) and let \( C_{H} \) be a length-\( n' \) \((d_h, k)\)-WWL code with minimum Hamming distance \( d_h \). We define the following code.

\[ C_2(n, k, d_h, d_m) = \{ 0^k u 1^d_n c 1^d_m \mid c \in C_{H} \}. \]

The correctness of Construction II is proved in the next theorem.

**Theorem 4.5.** The code \( C_2(n, k, d_h, d_m) \) is a \((d_h, d_m)\)-MU code.

**Proof** For simplicity of notation let \( C = C_2(n, k, d_h, d_m) \) and \( a, b \in C \). The code \( C \) has minimum distance \( d_h \) since \( a_{k+\ell(d_m)+d_m+1}^{n-d_m} b_{k+\ell(d_m)+d_m+1}^{n-d_m} \in C_{H} \) and \( C_{H} \) has minimum distance \( d_h \). For the second part of the proof we use the following claim. The proof is left to the reader.

**Claim 4.6.** Let \( x, y \) be two \((d, k)\)-WWL vectors, then the vector \( x 1^d y \) is also a \((d, k)\)-WWL vector.

We show that for any \( a, b \) which are not necessarily distinct, for all \( i \in [n-1] \):
\[ d_H(a_i^1, b_i^1 = i, d_m) \geq \min\{i, d_m\} \]. We consider the following cases:

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Technion - Computer Science Department - M.Sc. Thesis  MSC-2018-08 - 2018
1. For \( i \in [1, d_m] \), \( a_i = 0^i, b_{n-i+1}^n = 1^i \), and thus \( d_H(a_i^i, b_{n-i+1}^n) = i = \min\{i, d_m\} \).

2. For \( i \in [d_m + 1, k] \), \( a_i^i = 0^i, b_{n-i+1}^n = b_{n-i+1}^n 1^i \), and hence \( d_H(a_i^i, b_{n-i+1}^n) \geq d_m \).

3. For \( i \in [k + 1, n - d_m] \), notice that \( b_{dn-i+1}^n = 0^{k-d_m} 1^d_n u_{d_n-i+1}^{(d_n)} 1^d_m \), where \( c \in C_H, 0^{k-d_m} \), and \( u_{d_n-i+1}^{(d_n)} \) are all \((d_m, k)\)-WWL vectors. From Claim 4.6 \( b_{dn-i+1}^n \) is also a \((d_m, k)\)-WWL vector. Since \( n - i + 1 > d_m \), \( b_{n-i+1}^n \) is a subvector \( b_{dn-i+1}^n \) and as such, its first \( k \) positions consist at least \( d_m \) ones. Hence, \( d_H(a_i^i, b_{n-i+1}^n) \geq d_m \).

4. For \( i \in [n - d_m + 1, n - 1] \), let \( j = n - i \), and \( \hat{a} = a_i^i, \hat{b} = b_{n-i+1}^n = b_{n-i+1}^n \). Notice that \( \hat{a}_{k-j}^{k-j+\ell(d_m)} = 0^j 1^{d_n-j} \) and \( \hat{b}_{k-j}^{k-j+\ell(d_m)} = u \). \( u \) is a \( d_m \)-auto-cyclic vector, hence \( d(n, (d_n-1), u) \geq d_m \) and we get \( d_H(a_i^i, b_{n-i+1}^n) = d_{H}(\hat{a}, \hat{b}) = d_H(\hat{a}_{k-j}^{k-j+\ell(d_m)}, \hat{b}_{k-j}^{k-j+\ell(d_m)}) \geq d_m \).

Next, we are interested in determining the maximum value of the codes’ cardinality from Construction II, when optimizing over all possible values of \( k \). We start with the case \( d_h = 1 \), that is, the case in which we don’t require the codewords to differ by more than one symbol, but we require the MU property with distance \( d_m \).

**Lemma 4.7.** The redundancy of the code \( C_2(n, k, 1, d_m) \) is minimized when \( k \) is of the form \( k = \log n + (d_m - 1) \log \log n + \Theta(1) \). The minimal redundancy is \( k + \Theta(1) = \log n + (d_m - 1) \log \log n + \Theta(1) \).

**Proof** The construction’s redundancy includes the \( k + 2 \) fixed bits and the redundancy bits of \( C_H \). Therefore, values of \( k \) of the form \( k = \mathcal{F}(n, d_m) + \omega(1) \) result in redundancy greater than \( \log n + (d_m - 1) \log \log n + \Theta(1) \). When \( k = \mathcal{F}(n, d_m) + \Theta(1) \) we can use Algorithm 2 to construct \( C_H \) with \( \Theta(1) \) redundancy bits and get total redundancy of \( \log n + (d - 1) \log \log n + \Theta(1) \) bits. Lastly, from Corollary 4.3, when \( k = \mathcal{F}(n, d_m) - f(n), f(n) \geq 0 \) the redundancy of \( C_2(n, k, 1, d_m) \) is \( k + \Theta(2^f(n)) = \log n + (d - 1) \log \log n - f(n) + \Theta(2^f(n)) \). This term is minimized when \( f(n) = \Theta(1) \) and again, it yields total redundancy of \( \log n + (d - 1) \log \log n + \Theta(1) \).

According to Lemma 4.7, Construction II provides existence of \( (d_h = 1, d_m) \)-MU codes which are \( \Theta(\log \log n) \) away from the lower bound on redundancy in
Theorem 4.4. Note that the results so far did not include an explicit description of an encoder and decoder for general values of $d_h, d_m$. In order to provide such an efficient construction we present in our next result a $(d_h, d_m)$-MU code with linear encoding and decoding complexities, and \( \left\lceil \frac{d_h+1}{2} \right\rceil \log n + (d_m - 1) \cdot \log \log n + \mathcal{O}(d_m \log d_m) \) redundancy bits. In this case, it is also \( \Theta(\log \log n) \) away from the bound on redundancy in Theorem 4.4.

For the suggested construction, we choose $k = F(n, d_m) + 1$ in Construction II, and construct the code $C_{ij}$ with Algorithm 2 to provide the WWL property. We also allow greater values of $d_h$ by incorporating a systematic code of minimum Hamming distance $d_h$. This result is stated in the following corollary.

**Corollary 4.8.** There exists a binary $(d_h, d_m)$-MU code with redundancy

\[
\left\lceil \frac{d_h + 1}{2} \right\rceil \log n + (d_m - 1) \log \log n + \mathcal{O}(d_m \log d_m)
\]

and linear time and space encoding and decoding complexities.

**Proof sketch** We use Algorithm 2 to generate a $(d_m, F(n, d_m))$-WWL code of length $\tilde{n} < n$, where the choice of $\tilde{n}$ will be explained later. We then guarantee minimum distance $d_h$ by applying a systematic BCH encoder on the output from Algorithm 2. The approximately $\left\lceil \frac{d_h-1}{2} \right\rceil \log(\tilde{n})$ redundancy bits are spread within the $\tilde{n}$ bits such that the difference in the positions of each two redundancy bits is greater than $F(n, d_m) + 1$ (we assume that $\tilde{n}$ is large enough to allow this). That way the resulting vector is a $(d_m, F(n, d_m) + 1)$-WWL vector. We denote this resulting set of vectors by $C'$, so the code $C'$ is a $(d_m, F(n, d_m) + 1)$-WWL code with minimum distance $d_h$ and its length is a function of $\tilde{n}$ denoted by $f(\tilde{n})$. The redundancy of $C'$ is $\left\lceil \frac{d_h-1}{2} \right\rceil \log(\tilde{n}) + \mathcal{O}(1)$. We construct the code $C_{ij}(n, k)$ by choosing $k = F(n, d_m) + 1$ and using $C'$ as $C_{ij}$. The choice of $\tilde{n}$ is determined in a way that $f(\tilde{n}) = n' = n - k - \ell(d_m) - 2d_m$ is satisfied. Thus, the total redundancy of this construction is upper bounded by

\[
F(n, d_m) + \ell(d_m) + 2d_m + 1 + \left\lceil \frac{d_h - 1}{2} \right\rceil \log n + \mathcal{O}(1)
\]

Lastly, we note that Construction II improves upon Construction 3 from [38], which
solves the problem of \((d_h, d_m = 1)\)-MU codes but requires \(O(\sqrt{n})\) redundancy bits.

### 4.3 Related Families of Codes

In this section we present two families of codes which are closely related to \((d_h, d_m)\)-MU Codes. Namely, \((d, \rho)\) comma-free codes and prefixed synchronized codes with index \(\rho\) [27]. We discuss these codes and their connection with \((d_h, d_m)\)-MU codes.

#### 4.3.1 \((d, \rho)\) Comma-free Codes

Comma-free codes were studied in [17], motivated by the problem of word synchronization for block codes. A code \(C \subseteq \Sigma_q^n\) is a comma-free code if for any two not necessarily distinct vectors \(a, b \in C\) and \(i \in [2, n]\), \((ab)_i^{n+i-1} \not\in C\). We denote by \(A_{CF}(n, q)\) the maximal cardinality of a comma-free code of length \(n\) over \(\Sigma_q\). It was shown in [17] that

\[
A_{CF}(n, q) \leq \frac{1}{n} \sum_{d|n} \mu(d) q^{n/d},
\]

where \(\mu(d)\) is the M"{o}bius function and the sum is taken over all divisors of \(n\). Later in [13], an optimal construction of odd length comma-free codes was introduced leading to

\[
A_{CF}(n, q) = \frac{1}{n} \sum_{d|n} \mu(d) q^{n/d} \approx \frac{q^n}{n}
\]

for odd \(n\).

Note that the comma-free property is weaker than the MU property, therefore an MU-code is a comma-free code but not vice versa; in particular, \(A_{MU}(n, q) \leq A_{CF}(n, q)\).

An extension to comma-free codes by Levenshtein [27] states that \(C \subseteq \Sigma_q^n\) is a \((d, \rho)\) comma-free code if for any \(a, b, c \in C\), \(i \in [2, n]\), \(d_H((ab)_i^{n+i-1}, c) \geq \rho\), and \(C\) has a minimum Hamming distance \(d\).

Levenshtein proved that there exists a construction of \((d, \rho)\)-comma-free codes with redundancy of approximately \([\frac{d+1}{2}] \log n + c(\rho)\) bits, where \(c(\rho)\) is a constant that depends on \(\rho\). However, the encoding and decoding of such codes are
complex. To allow efficient encoding and decoding, Levenshtein suggested in [29] to use constructions based on linear codes. But, in this case, it was shown in [3] that the redundancy is at least $\sqrt{\rho n}$. Note that any $(d, \rho)$-MU code is also a $(d, \rho)$-comma-free code, so Construction II is a $(d, \rho)$-comma-free code and we can use the result of Corollary 4.8 to construct efficient $(d, \rho)$-comma-free code with significantly less redundancy, as described next.

**Corollary 4.9.** There exists a construction of a $(d, \rho)$-comma-free code with efficient encoding and decoding and $\left\lfloor \frac{d+1}{2} \right\rfloor \log n + (\rho - 1) \log \log n + o(\log \log n)$ redundancy bits.

For the case of $\rho \geq d$ this construction is $O(\log \log n)$ away from the optimal possible redundancy according to [27]. The details of the construction are described in the proof of Corollary 4.8.

### 4.3.2 Prefix Synchronized Codes with Index $\rho$

For a set $H \subseteq \Sigma^m$, a prefix synchronized code $C_H \subseteq \Sigma^n_q$ is defined to be the set of words $a \in \Sigma^n_q$ such that for any $h \in H$, the word $ah$ contains a word from $H$ only in the first and last $m$ positions [16, 29]. Construction I is in fact a prefix synchronized code with the set $H = \{0^k\}$. Prefix synchronized codes can be defined for any set of prefixes $H$. A related problem is discussed in [27], namely, prefix synchronized codes with index $\rho$. A code $C \subseteq \Sigma^n$ is said to be prefix synchronized with a set $H \subseteq \Sigma^m$, $m \leq n$ and index $\rho$ if for any $a \in C$, $h \in H$, $i \in [2, n], d_H((ah)^{i+m-1}, h) \geq \rho$. Levenshtein stated in [27] that when $n$ goes to infinity, a lower bound on the redundancy of a prefix synchronized code with index $\rho$ is $\log n + (\rho - 1) \log \log n + \log \log \log n$. The next theorem provides a prefix synchronized code which is close to optimal.

**Theorem 4.10.** The code $C_2(n, k, 1, d_m)$ is prefix synchronized with $H = \{0^k u\}$ and $\rho = d_m$.

Hence, by Corollary 4.8 we provide an efficient construction to this problem with only $o(\log \log n)$ additional bits of redundancy.
Chapter 5

Other MU Codes Extensions

5.1 MU Codes with Edit Distance

5.1.1 Definitions and Preliminaries

In this section we turn to another extension of MU codes which imposes a minimum edit distance between prefixes and suffixes as well as on the code. This extension is motivated by several works such as [37] which report on deletion errors during the synthesis process of DNA strands.

The edit distance, denoted by \( d_E(a, b) \), of two words \( a, b \) is the minimum number of insertions and deletions that transform \( a \) to \( b \). The minimum edit distance of a code \( C \) is the maximal \( d \) such that for any two distinct words \( a, b \in C \), \( d_E(a, b) \geq d \). For a word \( a \) we say that \( (a_{i_1}a_{i_2}\ldots a_{i_k}) \) is a subsequence of \( a \) if \( 1 \leq i_1 < i_2 < \cdots < i_k \leq n \). A common subsequence of two words \( a, b \) is a sequence which is a subsequence of \( a \) and \( b \). We say that a sequence \( c \) is a largest common subsequence (lcs) of \( a, b \) if there is no common subsequence of \( a \) and \( b \) with length greater than the length of \( c \). Note that every lcs of \( a, b \) has the same length, which will be denoted by \( \ell(a, b) \).

Next, we list several commonly known claims that will be helpful in the following proofs of this section. The claims’ proofs appear in Appendix D for completeness.

Claim 5.1. For \( a \in \Sigma_q^n, b \in \Sigma_q^m \): \( d_E(a, b) = n + m - 2\ell(a, b) \).

Claim 5.2. For \( a \in \Sigma_q^n, b \in \Sigma_q^n, c \in \Sigma_q^m, d \in \Sigma_q^m \): \( d_E(ac, bd) \geq \max\{d_E(a, b), d_E(c, d)\} \).

43
Claim 5.3. For \( a \in \Sigma_q^n, b \in \Sigma_q^n, c \in \Sigma_q^m, d \in D(a, b) \geq d_E(a, b)/2 \).

Claim 5.4. For \( a = 0^n, b \in \Sigma_q^n, d_E(a, b) = 2 \omega_H(b) \).

We are now ready to present the definition of MU codes with edit distance.

Definition 5.1. A code \( C \) is called a \((d_e, d_m)\)-EMU code if

1. the minimum edit distance of the code is \( d_e \),
2. for every two not necessarily distinct words \( a, b \in C \), and \( i, j \in [n - 1] \), if \( i, j \in [d_m, n - d_m] \): \( d_E(a_i^a, b_{n-j+1}^b) \geq d_m \), otherwise \( d_E(a_i^a, b_{n-j+1}^b) \geq \min\{i, j, n - i, n - j\} \).

The second condition in Definition 5.1 is different from the MU constraints we introduced so far since now we require large distance of suffixes and prefixes of different lengths, while in Section 4.2 we required large Hamming distance between prefixes and suffixes of the same length. This choice of the constraint will assure that suffixes and prefixes of the addresses will not get confused even if they experienced deletions and insertions.

5.1.2 Upper Bound

We set \( A_{EMU}(n, q, d_e, d_m) \) to be the largest cardinality of any \((d_e, d_m)\)-EMU code over \( \Sigma_q^n \), and \( E(n, q, d) \) is the largest cardinality of a code over \( \Sigma_q^n \) with minimum edit distance \( d \). The following is an upper bound on the value \( A_{EMU}(n, q, d_e, d_m) \).

Theorem 5.5. For all \( n, q, d_e, d_m \),

\[
A_{EMU}(n, q, d_e, d_m) \leq \frac{E(n, q, d)}{\lfloor n/d_m \rfloor},
\]

where \( d = \min\{d_e, d_m\} \).

Proof Given a \((d_e, d_m)\)-EMU code, \( C \subseteq \Sigma_q^n \), we define \( \hat{C} \) and \( \hat{a}, \hat{b} \in \hat{C} \) similarly to the proof of Theorem 4.4. The code \( \hat{C} \) is defined as the code of all cyclic shifts of \( \alpha \cdot d_m \) bits of codewords of \( C \). If \( \hat{a}, \hat{b} \in \hat{C} \) are of the same shift, their edit distance is at least \( d_e \). Otherwise, we denote \( \hat{a} = (aa)_{i+1}^n, \hat{b} = (bb)_{j+1}^n \), where \( a, b \in C \), \( i, j \) are multiples of \( d_m \), and we assume without loss of generality that \( i > j \). The subsequence \( \hat{a}_{n-i+1}^{n-j} \) is a prefix of \( a \) and \( \hat{b}_{n-i+1}^{n-j} \) is a suffix of \( b \). Moreover, the
length of \( \hat{a}_{n-i+1}^{n-j} \) is at least \( d_m \), therefore, from the definition of \((d_e, d_m)\)-EMU codes, \( d_E(\hat{a}_{n-i+1}^{n-j}, \hat{b}_{n-i+1}^{n-j}) \geq d_m \). We now apply Claim 5.2 twice and get that

\[
d_E(\hat{a}, \hat{b}) \geq d_E(\hat{a}_{n-i+1}^{n-j}, \hat{b}_{n-i+1}^{n-j}) \geq d_m.
\]

We showed that \( d_E(\hat{a}, \hat{b}) \geq \min\{d_e, d_m\} \), thus

\[
|\hat{C}| = \lfloor n/d_m \rfloor \cdot |C| \leq E(n, q, d),
\]

where \( d = \min\{d_h, d_e\} \) and the theorem follows directly.

In [24], it was shown that

\[
E(n, q, 4) \leq \frac{q^n - q}{(q - 1)n}
\]

which aligns with the asymptotic upper bounds by [26, 35] \( E(n, q, 4) \lesssim \frac{q^n}{(q-1)n} \).

Theorem 5.5 therefore implies that the minimum redundancy when \( \min\{d_e, d_m\} = 4 \) is at least \( 2 \log_q n + \Theta(1) \).

5.1.3 Constructions

The following lemma is given without a proof as it shares similar ideas with the proof of Theorem 5.7 that follows.

**Lemma 5.6.** The code \( C_2(n, k, 1, d_m) \) is a \((2, d_m)\)-EMU code.

Next, we slightly modify Construction II to allow values of \( d_e \) greater than 2.

**Construction III.** Let \( n, k \) be two integers such that \( k \geq d_m \) and \( n \geq k + 2d_m \).

Let \( n' = n - k - 2d_m \) and \( C_E \) be a \((d_m, k)\)-WWL code of length \( n' \) with minimum edit distance \( d_e \). The code \( C_3(n, k, d_e, d_m) \) is defined as follows,

\[
C_3(n, k, d_e, d_m) = \{0^k1^{d_m}\epsilon 1^{d_m} | \epsilon \in C_E\}.
\]

The correctness of Construction III is proved in the next theorem.

**Theorem 5.7.** The code \( C_3(n, k, d_e, d_m) \) is a \((d_e, d_m)\)-EMU code.
Proof We use the notation $C = C_3(n,k,d_e,d_m)$ for simplicity. Any two words $a,b \in C$ satisfy $d_E(a^n_{k+d_m+1}, b^n_{k+d_m+1}) \geq d_E$. Applying Claim 5.2 twice gives us $d_E(a, b) \geq d_E$.

We denote $a^i_1 = x, b^n_{n-j+1} = y$. Notice that a word $b \in C$ has the following structure $b^j_{d_m+1} = 0^{k-d_m}1^{d_m} b^j_{k+d_m+1} 1^{d_m}$ such that $b^n_{k+d_m+1}$ is a $(d_m,k)$-WWL vector. Therefore, according to Claim 4.6, $b^n_{d_m+1}$ is also a $(d_m,k)$-WWL vector, and since $y$ is a subsequence of it, $y$ is a $(d_m,k)$-WWL vector as well. The following cases are considered,

1. $j \in [n-d_m], i = j$: in that case we show a stronger property of $x,y$ which claims that $d_E(x,y) \geq 2 \min\{d_m,i,j\}$. If $i = j \leq k$ then $x = 0^i$ and $w_H(y) \geq \min\{d_m,i,j\}$ since $y$ ends with $\min\{d_m,i,j\}$ 1s. From Claim 5.4 we get $d_E(x,y) \geq 2 \min\{d_m,i,j\}$. If $i = j > k$, $x^j_1 = 0^k$ and $w_H(y^j_1) = d_m$ as $y$ is a $(d_m,k)$-WWL vector. Claim 5.4 yields $d_E(x^j_1,y^j_1) \geq 2d_m$ and $d_E(x^j_1 x^j_{k+1}, y^j_1 y^j_{k+1}) = d_E(x,y) \geq \max\{2d_m,w_H(x^j_1,y^j_{k+1})\} \geq 2d_m$ according to Claim 5.2.

2. $j \in [n-d_m+1,n-1], i \neq j$: we assume that $i > j$. $x^j_1 = a^j_1$ and $y = b^n_{n-j+1}$, hence $d_E(x,y) \geq 2 \min\{d_m,i,j\}$ following the previous case. Claim 5.3 implies $d_E(x^j_1 x^j_{k+1}, y) = d_E(x,y) \geq 2 \min\{d_m,i,j\}/2 = \min\{d_m,i,j\}$. The proof for $j > i$ is similar.

3. $j \in [n-d_m+1,n-1], i < j$: $y = b^n_{n-j+1} - 0^{k-(n-j)}1^{d_m} b^n_{k+d_m+1} 1^{d_m}$. Since $n-j < d_m$ and $b^n_{k+d_m+1}$ is a $(d_m,k)$-WWL vector, we deduce that $y$ is a $(n-j,k)$-WWL vector. In the case of $i \geq k$, following a similar path to the proof when $j \in [n-d_m]$ leads us to $d_E(x,y) \geq n-j \geq \min\{i,j,n-i,n-j\}$. If $i < k$ then $j-i > d_m$ and hence $d_E(x,y) \geq d_m > n-j$.

Even though Construction III provides a construction of $(d_e,d_m)$-EMU codes for all $d_e$ and $d_m$ it heavily depends on the existence of codes with edit distance. The knowledge on codes with large minimum edit distance is quiet limited, in the sense that there exist codes with rate 1, however their structure is complex and there is no explicit expression for their redundancy [10]. Hence, for the rest of this section we focus on the case of edit distance four, i.e. codes correcting a single deletion or insertion.

There exists an explicit efficient method to construct a $(d_m,F(n,d_m)+1)$-WWL codes with minimum edit distance 4, which will be used as the code $C_E$.
in Construction III. For this, we use Algorithm 2 and the well known Varshamov
Tenengolts (VT) codes with edit distance four in their systematic version [1, 36].
Namely, the VT code is defined for all \( n \) and \( b \in \mathbb{Z}^n \) by

\[
VT(b) = \{ a = (a_1, \ldots, a_n) \in \Sigma^n | \sum_{i=1}^n i \cdot a_i \equiv b (\text{mod } n + 1) \}.
\]

The systematic version of a VT code converts any vector of length \( n' = n - \lceil \log(n + 1) \rceil \) to a VT vector of length \( n \) by adding \( \lceil \log(n + 1) \rceil \) redundancy
bits in locations corresponding to powers of 2 [1]. Any integer \( i \in [0, n] \) can
be represented by a sum of a subset of those indices (the subset corresponding
to its binary representation). We choose to represent in those indices the integer
that guarantees the fulfillment of the VT constraint by the resulting vector. The
complexity of the encoding and decoding of this method is linear, we use it to
achieve the following result.

**Theorem 5.8.** There exists a construction of \((d_m, F(n, d_m) + 1)\)-WWL code of
length \( n \), with minimum edit distance 4, redundancy \( \log n + O(d_m) \) and linear
time and space encoding and decoding complexities.

**Proof sketch** To construct the code \( C \) which is a \((d_m, F(n, d_m) + 1)\)-WWL code
with minimum edit distance 4, we start with a \((d_m, F(n, d_m))\)-WWL code \( C_{wwl} \)
of length \( n' \), where \( n' + 2d_m + \lceil \log(n' + 2d_m) \rceil = n \). An efficient algorithm
for encoding and decoding of such a code was presented in Lemma 4.2. We then
define

\[
C_{ewwl} = \{ a_{i_1}^{i_1} 1^{d_m} a_{i_2}^{i_2+1} 1^{d_m} a_{i_3}^{i_3} \mid a \in C_{wwl} \}
\]

where \( i_1 \leq i_2 \) and the choice of their values will be explained later. The extension
to \( C_{ewwl} \) is aimed to ensure that the final resulting code satisfies the \((d_m, F(n, d_m) + 1)\)-WWL constraint. It is readily verified that the code \( C_{ewwl} \) is also a \((d_m, F(n, d_m))\)-WWL code. We now apply the systematic VT code on \( C_{ewwl} \) to get the code \( C \) with
minimum edit distance 4. The length of \( C_{ewwl} \) is \( n' + 2d_m \), hence the length of \( C \)
is \( n' + 2d_m + \lceil \log(n' + 2d_m) \rceil = n \). We are only left with showing that \( C \) is a
\((d_m, F(n, d_m) + 1)\)-WWL code.

Recall that the redundancy bits are located in indices which are powers of
2. As a result, for any \( a \in C \), \( a_2^{\lceil \log n \rceil + 1} \) does not include a window of length
\( F(n, d_m) = \log n + o(\log n) \) that consists of more than one redundancy bit.
Combining it with the fact that $C_{ewwl}$ is a $(d_m, F(n, d_m))$-WWL code we get that $a_{2^{|\log n|}+1}^n$ is a $(d_m, F(n, d_m) + 1)$-WWL vector.

Lastly, we can choose $i_1$ and $i_2$ when constructing the code $C_{ewwl}$ such that a vector $a \in C$ satisfies $a_{|\log n|}^n = 1^{d_m}$ and $a_{2|\log n|}^{2^n} = 1^{d_m}$, or in other words $a = x^1 y^1 z^1$ where $x, y$ are of length $|\log n| - d_m$ and $z$ of length $n - 2|\log n|$. In that case, we showed that $z$ is a $(d_m, F(n, d_m) + 1)$-WWL vector and since the length of $x, y$ is smaller than $F(n, d_m)$ they are also $(d_m, F(n, d_m) + 1)$-WWL vectors. Using Claim 4.6 we conclude that $a \in C$ is a $(d_m, F(n, d_m) + 1)$-WWL vector.

We integrate the code $C$ from the proof above as $C_E$ in Construction III and conclude the result in the following corollary.

**Theorem 5.9.** There exists a construction of $(4, d_m)$-EMU codes with redundancy $2 \log n + (d_m - 1) \log \log n + O(d_m)$ and linear time and space complexity.

To summarize, in this section we first introduced in Theorem 5.5 a lower bound on the redundancy of $(d_e, d_m)$-EMU codes. We then presented a general structure of a $(d_e, d_m)$-EMU code in Construction III, and used this structure to obtain an explicit construction, with efficient encoder and decoder, when $d_e = 4$. For this case, the construction is $(d_m - 1) \log \log n + \Theta(1)$ redundancy bits away from the lower bound in Theorem 5.5.

### 5.2 Balanced Mutually Uncorrelated Codes

#### 5.2.1 Definitions and Upper Bound

In this section we study yet another extension of MU codes. Under this setup we seek the codes to be balanced. A binary word of length $n$, when $n$ is even, is said to be **balanced** if its Hamming weight is $n/2$. It is well known that the number of balanced words is $\binom{n}{n/2} \approx \frac{2^{n+1}}{\sqrt{2\pi n}}$. For the extension of $q > 2$, when $q$ is even, we follow the balanced definition from [38] and say that a code $C \subseteq \Sigma_q^n$ is balanced if for any $a \in C$ the number of positions $i$ such that $a_i \in [0, \frac{q}{2} - 1]$ is $n/2$. Hence, the number of $q$-ary balanced words is $\binom{n}{n/2}(q/2)^n \approx \frac{2^{q^n}}{\sqrt{2\pi n}}$. For the rest of this section we assume that $n$ and $q$ are even. A code $C \subseteq \Sigma_q^n$ is said to be a **balanced MU code** if $C$ is balanced and is also an MU code. Let $A_{BMU}(n, q)$ denote the maximum cardinality of balanced MU codes.
Theorem 5.10. For all $n, q$, $A_{BMU}(n, q) \leq \frac{(\frac{n}{2})^n}{n} \approx \frac{2^n}{n^{\sqrt{2n}}}$. In particular, the redundancy of balanced MU codes is at least $1.5 \log n + O(1)$.

Proof Assume that $C$ is a balanced MU code and let $\hat{C}$ be the following code.

\[ \hat{C} = \{(a, a)_{i+n}^i \mid a \in C, i \in [1, n]\}. \]

That is, $\hat{C}$ is the code of all cyclic shifts of words from $C$. The code $\hat{C}$ is balanced and hence $\hat{C}$ is balanced as well. Furthermore, since the code $C$ is an MU code, the cardinality of $\hat{C}$ is $n \cdot |C|$, and we obtain $|\hat{C}| = n \cdot |C| \leq (\frac{n}{2})^n$.

5.2.2 Construction

Next we present a construction of binary balanced MU codes.

Construction IV. Let $n, k$ be two integers such that $1 \leq k < n$. The code $C(n, k) \subseteq \Sigma^n$ is defined as follows,

\[ C_4(n, k) = \{0^k1c1 \mid w_H(c) = \frac{n}{2} - 2, \ c \text{ has no zeros run of length } k\}. \]

The correctness and redundancy result of Construction IV are stated in the next theorem. It follows from the proof that the redundancy of the maximal $C_4(n, k)$ is $1.5 \log n + O(1)$ and from Theorem 5.10, this result is optimal up to a constant number of redundancy bits.

Theorem 5.11. The code $C_4(n, k)$ is a balanced MU code, and for an integer $k = \log n + a$, $|C_4(n, \log n + a)| \geq C \frac{2^n}{n^{\sqrt{2n}}}$, where $C = \frac{2^a-1}{2^{a+1} \sqrt{2\pi}}$.

Proof The cardinality of $C_4(n, k)$ is the number of binary words of length $n' = n - k - 2$ which consist of $n/2 - 2$ ones and do not contain a zeros run of length $k$. To lower bound the number of these words we count all words of length $n'$ and weight $n/2 - 2$, $(\frac{n'}{2})$, and reduce an upper bound on the number of such words that contain a zeros run of length $k$. For any word $a \in \Sigma^{n'}$ of weight $n'/2 - 2$ with zeros run of length $k$ there exists an index $i \in [1, n' - k + 1]$ such that $a_i^{i+k-1} = 0^k$ and the remaining $n' - k$ bits consist of exactly $n/2 - 2$ 1s. There are $n' - k + 1$ possibilities for $i$, and for any fixed $i$, there are $(\frac{n'}{2})$ possibilities for
the remaining \( n' - k \) symbols. Therefore, the number of vectors of weight \( n/2 - 2 \) that have a zeros run of length \( k \) is upper bounded by \((n' - k + 1)\binom{n' - k - 2}{n/2 - 2}\). Based on this observation we show that \(|C_4(n, k)| \gtrsim \frac{2^n}{n^{\sqrt{n}}} \frac{2^{n+1} - 1}{\sqrt{2\pi}} n^{n/2 - 2}\). The technical details of this step appear in Appendix E.

The extension of Construction IV for non-binary is direct, since every symbol can store \( q/2 \) values after the assignment of binary values. Hence the number of redundancy symbols remains the same, i.e., \( 1.5 \log_q n + O(1) \). This meets the result from [38] where a balanced MU code over the alphabet \( \{A, C, G, T\} \) was suggested, with redundancy of \( 1.5 \log_4(n) + O(1) \) symbols. However, our construction is also applicable for the binary case as opposed to the one in [38].

Lastly, we discuss efficient implementation of Construction IV. In [23], Knuth presented an efficient (linear complexity) algorithm to construct balanced words. His algorithm is based on the observation that for every binary vector \( a \) there exists an index \( i \in [1, n] \) such that the vector \( a_i \overline{a}_{i+1} \) is balanced, where \( a_i \) is the complement vector of \( a \). To convert an arbitrary vector \( a \) to a balanced vector, one can store the balanced vector \( a_i \overline{a}_{i+1} \), and append a binary balanced representation of the index \( i \). The total redundancy of this method is \( \log n + \log \log n + o(\log \log n) \). We extend Knuth’s method to provide an efficient construction of balanced MU codes with linear encoding and decoding complexity and redundancy \( 2 \log n + \log \log n + o(\log \log n) \).

**Theorem 5.12.** There exists a construction of balanced MU codes with \( 2 \log n + \log \log n + o(\log \log n) \) redundancy bits and linear time and space complexity.

**Proof** We describe an algorithm for efficiently encoding words of length \( n \), that agree with the structure presented in Construction IV. We assume for simplicity that \( \log n, \log \log n, \log \log \log n \) are integers and we denote by \#(a), \#(a) the number of ones and zeros in \( a \), respectively. We start with a vector \( x \) of length

\[
n' = n - 2 \log n - \log \log n - 2 \log \log \log n - 14
\]

and apply Algorithm 3.

After Step 1, \( x \) has no zeros runs of length \( \log n + 1 \). According to Step 2, \( v \) is the integral vector of \( x \), therefore, it does not contain a zeros or ones run of length \( \log n + 2 \). Then, we add in Step 3 the required prefix \( 0^{\log n + 31} \). We are now left with balancing the vector \( v \). For that purpose, we use Knuth’s Algorithm with
Algorithm 3 Extended Knuth’s Algorithm for balanced MU codes

Input: $x \in \Sigma^m$

Output: balanced $y \in \Sigma^n$, $y = 0^\log n + 3 y^*1$ where $y^*$ does not contain a run of zeros of length $\log n + 3$.

1: Execute Algorithm 1 to remove zeros runs of length $\log n + 1$ from $x$

2: Let $v \in \Sigma^m$, $v_i = \sum_{j=1}^{i} x_j$

3: $v = 0^{\log n + 3} 1 v$

4: If $v$ is balanced set $i = 0$, otherwise find an index $i$ such that $v_i v_{i+1}^*$ is balanced.

5: if $0.5 \log n + 2 < i \leq \log n + 4$ then

6: $w = v_1^{\log n + 4} v_{i+1}^{\log n + 5} 01$

7: else

8: $w = v 10$

9: end if

10: If $w$ is balanced set $i = 0$, otherwise find an index $i$ such that $w_i w_{i+1}^*$ is balanced

11: if $i > \log n + 4$ then

12: $p(i)$: balanced binary representation of $i$ with $\log n \log \log n + 2 \log \log \log n + 2$ bits, such that it does not contain a zero run of length $\log n$

13: $y = w_1^{\log n + 4} w_i^{\log n + 5} 1 p(i) 0001$

14: else ($i \leq 0.5 \log n + 2$)

15: $\ell = \#_0(w) - \#_1(w)$

16: $y = w_1^{\ell} z 11$ such that $z 11$ is balanced and $y \in \Sigma^n$

17: end if
some adaptations that ensure the overall structure of the output vector remains as
required by Construction IV, i.e., with a prefix $0^{\log n + 3}1$, followed by a sequence
with no zeros run of length $\log n + 3$, and ends with 1. As in Knuth’s algorithm,
we first find an index $i$ in $v$ such that $v^i_{1^i v^i_{n + 1}}$ is balanced. We consider the following
cases in Step 4:

1. $i > \log n + 4$ : in Step 8 we set $w = v10$ and in Step 10, $i > \log n + 4$
is still satisfied. We then set $y$ to $w_i^i 1 w_{i+1}^i$. Since $v$ originally did not have
a zeros or ones run of length $\log n + 2$ other than its prefix, $w$ does not
contain a run of length $\log n + 3$ and $y = w_i^i 1 w_{i+1}^i$ does not contain a zeros
run of length $\log n + 3$. Similarly to Knuth’s algorithm, we also append a
balanced binary representation of $i$. Such a representation is available with
$\log n + \log \log n + 2 \log \log \log n$ bits. Since we additionally require it
to not include a zeros run of length $\log n + 3$ we insert 10 in its $\log n$'th
position to get $p(i)$. The returned vector is appended with 0001 for balancing
purposes.

2. $i < \log n + 4, i \neq 0$ : here, we do not simply apply the same approach as
in the previous case because we want to guarantee $0^{\log n + 3}1$ is a prefix of $y$.
$v^i_{1^i v^i_{n + 1}}$ is balanced, hence

$$\#_0(v^i_{1^i v^i_{n + 1}}) = \#_1(v^i_{1^i v^i_{n + 1}})$$

and since

$$\#_0(v^i_{1^i v^i_{n + 1}}) = i + \#_1(v), \#_1(v^i_{1^i v^i_{n + 1}}) = \#_0(v) - i$$

we have that

$$2i = \#_0(v) - \#_1(v). \tag{5.1}$$

(a) If $i \leq 0.5 \log n + 2$, then $w = v10$ and in Step 10 $i$ remains the same.
In Step 16,

$$\ell = \#_0(w) - \#_1(w) \leq \log n + 4$$

and therefore we balance $y$ by simply appending the sequence $1^\ell z11$.

(b) If $0.5 \log n + 1 < i < \log n + 4$ then $\ell$ might exceed our desirable
redundancy number of bits. Our alternative solution is to set $w$ in Step 6

52
to $v_1^{\log n + 4}v_{\log n + 5}^{n'}01$. After Step 6

$$
\#_0(w) = \#_0(v_1^{\log n + 4}v_{\log n + 5}^{n'}01) = \log n + 3 + \#_1(v),
$$

$$
\#_1(w) = \#_1(v_1^{\log n + 4}v_{\log n + 5}^{n'}01) = \#_0(v) - \log n - 1
$$

and we have that

$$
\#_0(w) - \#_1(w) = \log n + 3 + \#_1(v)
$$

$$
- (\#_0(v) - \log n - 1)
$$

$$
= 2\log n + 4 - (\#_0(v) - \#_1(v))
$$

$$
\overset{\text{Eq. (5.1)}}{=} 2\log n + 4 - 2i
$$

$$
< \log n + 2.
$$

Therefore, after Step 10, similarly to Equation (5.1),

$$
2i = \#_0(w) - \#_1(w) < \log n + 3
$$

and the algorithm follows the path as in case 2a.

3. $i = \log n + 4$ or $i = 0$: in both cases $w$ is balanced after the if clause in Step 5. In Step 10 $i = 0$, the if statement in Step 11 is not satisfied and in Step 16 $\ell = 0$. We append some balanced suffix of the form $z11$ and of proper length such that $y = w1^\ell z11$ is a balanced length-$n$ vector as expected.

The vector $y$ is uniquely decodable in the following manner. By looking at the two rightmost bits we can detect whether the if statement in Step 11 was satisfied and reconstruct $w$. Then, again, we look at the two rightmost bits to detect whether the if statement in Step 5 was satisfied and reconstruct $v$. Finally, $x$ is derived from $v$ by removing the prefix $0^k1$, and computing the differences vector of $v$.

To conclude, we showed in Theorem 5.10 that the problem of balanced MU codes can be solved with at least $1.5 \log n + \Theta(1)$ redundancy bits. Construction IV and Theorem 5.11 provide existence and structure of a balanced MU code with such a redundancy. Lastly, in Theorem 5.12 we present an explicit construction with efficient encoding and decoding algorithms, with $2 \log n + \Theta(1)$ redundancy bits.
Chapter 6

Conclusions and Future Work

6.1 Summary

In this work we studied MU codes and their extension to MU codes with Hamming distance. For that purpose we looked into two interesting constraints, the $k$-RLL and the $(d, k)$-WWL constraints when $k$ is a function of the word’s length, $n$. The results of this study are presented in Table 4.1. We then continued to additional variations of MU codes, that is MU codes with minimum Edit distance and balanced MU codes. Similar techniques can be applied to construct balanced MU codes together with minimum Hamming distance and thereby satisfy three of the constraints listed in [38]. The results on the variations of MU codes are summarized in Table 6.1. For each case we first give the lower bound on the redundancy, then the construction that solves this case, and finally the best redundancy we could get with linear encoding and decoding complexity.

Table 6.1: Redundancy summary for binary MU codes

<table>
<thead>
<tr>
<th>Property</th>
<th>MU $(d_h, d_m)$: MU</th>
<th>$(4, d_m)$: EMU</th>
<th>Balanced MU</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lower bound</td>
<td>$\log n + \log(e)$</td>
<td>$2 \log n - \log d_m + O(1)$</td>
<td>$1.5 \log n + \log \sqrt{2\pi} - 1$</td>
</tr>
<tr>
<td>Construction</td>
<td>Construction I</td>
<td>Construction II</td>
<td>Construction III</td>
</tr>
<tr>
<td>Efficient $\leq$</td>
<td>$\left\lceil \log n + 4 \right\rceil$</td>
<td>$\left\lceil \log n + (d_m - 1) \log \log n \right\rceil$</td>
<td>$d_m \neq 4$</td>
</tr>
<tr>
<td>upper bound</td>
<td>$\frac{d_m}{2}$ $\log n$</td>
<td>$2 \log n + (d_m - 1) \log \log n$</td>
<td>$2 \log n + \log \log n$</td>
</tr>
<tr>
<td>Comments</td>
<td>$d = \min{2d_m, d_h}$</td>
<td>$+ O(d_m \log d_m)$</td>
<td>$+ o(\log \log n)$</td>
</tr>
</tbody>
</table>

Technion - Computer Science Department - M.Sc. Thesis  MSC-2018-08 - 2018
6.2 Future Work

Interesting question which remains open is the gap between the upper and lower bounds on the MU codes cardinality in the binary case. The author of [11] conducted a computer program that found that when $n \leq 16$, $n \neq 9$ the code $C_1(n, k, 2)$ with a proper choice of $k$, is the maximal MU code that exists. However, for general values of $n$ it is still not known whether the upper bound (2.1) is tight or Construction I is optimal.

Furthermore, in this work we obtain the order of redundancy of the $(d, k)$-WWL constraint, as mentioned in Table 4.1. An interesting problem is to further analyze the constraint and to obtain tighter bounds on the redundancy.

Lastly, DNA storage systems gain a lot of attention recently. Studies and analysis of errors in the writing and reading channels are starting to gather and many interesting coding problems are formed. While this work is a first step in studying this new research topic, we encourage the readers to further investigate coding problems motivated by DNA storage.
Appendix A

Missing Proofs: Section 3.1

**Proposition 3.4** Let \( f(n), g(n) \) be functions such that \( \lim_{n \to \infty} g(n) = 1 \) and \( 1 \leq f(n) \leq C \) for a constant \( C \) then

\[
|f(n)g(n)| \approx f(n)
\]

**Proof** For any \( 0 < \delta < 1 \) we choose

\[
\delta' = \min\{-\frac{\log(1-\delta)}{\log C}, \frac{\log(1+\delta)}{\log C}\} > 0
\]

such that \( 1 - \delta \leq C^{-\delta'} \) and \( C^{\delta'} \leq 1 + \delta \) are satisfied. There exists \( N' \) such that for every \( n \geq N' \)

\[
1 - \delta' \leq g(n) \leq 1 + \delta'.
\]

Therefore, for every \( n \geq N' \)

\[
C^{-\delta'} \leq f(n)^{-\delta'} \leq \frac{f(n)g(n)}{f(n)} \leq f(n)^{\delta'} \leq C^{\delta'}
\]

and from the choice of \( \delta' \) we get that for every \( n \geq N' \)

\[
1 - \delta \leq \frac{f(n)g(n)}{f(n)} \leq 1 + \delta
\]

hence

\[
\lim_{n \to \infty} \frac{f(n)g(n)}{f(n)} = 1
\]
Appendix B

Missing Proofs: Section 3.2

**Lemma 3.10** For \( k = \lceil \log_q n \rceil + z, z \in \mathbb{Z} \),

\[
2^{n'E_{k,q}} \approx \frac{q^n}{n} \cdot \frac{q^{\Delta_n - z - 2}}{e^{(q-1)q^{\Delta_n - z - 1}}},
\]

where \( \Delta_n = \log_q n - \lceil \log_q n \rceil \).

**Proof** From Lemma 3.5, when \( k = \lceil \log_q n \rceil + z \), and for \( n \) large enough there exists a constant \( C \geq 0 \) such that

\[
1 \leq 2^{n'(\log q - E_{k,q})} \leq 2^{n'\frac{C}{n}} \leq 2^C.
\]

We use Proposition 3.4 with \( f(n) = 2^{n'(\log q - E_{k,q})} \) and \( g(n) = \frac{(q-1)log_{eq}^{-k-2}}{\log q - E_{k,q}} \) to get

\[
2^{n'(\log q - E_{k,q})} \approx 2^{n'(q-1)log_{eq} - (k-1)}
\]

and conclude that

\[
2^{n'E_{k,q}} = 2^{n'(\log q - E_{k,q})} \approx 2^{n'(\log q - n'(q-1)log_{eq} - (k-1))}
\]

\[
= q^{n'}e^{-n'(q-1)q^{-k-1}}.
\]
The term \( n'(q - 1)q^{-k-1} \) satisfies
\[
\begin{align*}
n'(q - 1)q^{-k-1} &= (n - ([\log_q n] + z) - 2)(q - 1)q^{-(\lceil \log_q n \rceil + z) - 1} \\
&= (n - ([\log_q n] + z) - 2)(q - 1)q^{\Delta_n - z - 1} \\
&= (q - 1)q^{\Delta_n - z - 1} + o(1).
\end{align*}
\]

Finally, we conclude that
\[
2^{n'E_{k,q}} \approx q^{n'}e^{-(q-1)q^{\Delta_n - z - 1} + o(1)}
\]
\[
\approx \frac{q^n}{n^{e(q-1)q^{\Delta_n - z - 1}}}
\]

where \( \Delta_n = \log_q n - [\log_q n] \).

**Lemma 3.11** For \( z \in \mathbb{Z} \),
\[
|C_1(n, q, [\log n] + z)| \approx \frac{q^n}{n^{\Delta_n - z - 1}} q^{\Delta_n - z - 1} e^{(q-1)q^{\Delta_n - z - 1}}
\]

where \( \Delta_n = \log_q n - [\log_q n] \).

**Proof** From (3.6) we have
\[
1 \leq \frac{a_q(n', k)}{2^{n'E_{k,q}}} \leq 1 + \frac{2q^{n' - [k/2]}}{2^{n'E_{k,q}}},
\]

By using Lemma 3.10 we get that for \( k = [\log_q n] + z \)
\[
\lim_{n \to \infty} \frac{2q^{n' - [k/2]}}{2^{n'E_{k,q}}} = \lim_{n \to \infty} \frac{2q^{n' - [k/2]}}{2^{n'E_{k,q}}} = 0,
\]

and conclude that
\[
1 \leq \lim_{n \to \infty} \frac{a_q(n', k)}{2^{n'E_{k,q}}} \leq 1 + 2q^{n' - [k/2]}/2^{n'E_{k,q}} = 1.
\]

Hence,
\[
a_q(n', k) \approx 2^{n'E_{k,q}}.
\]
Recall that \(|C_1(n, q, k)| = (q - 1)^2 a_q(n', k)\). Together with Lemma 3.10 the result follows directly.

**Theorem 3.12**

\[
C_1(n, q) \approx \frac{q^n}{n} \cdot \frac{q - 1}{q} q^{F(\Delta_n)} \leq \frac{q^n}{n} \cdot \frac{q - 1}{eq},
\]

where \(\Delta_n = \log_q n - \lceil \log_q n \rceil\) and

\[
F(\Delta_n) = \max_{z \in \{-2, -1, 0\}} \{\Delta_n - z - \log_q (e)(q - 1)q^{\Delta_n - z - 1}\}.
\]

The inequality is tight when \(n \to \infty\) over any subsequence of \(n\) that satisfies \(\Delta_n = -\log_q (q - 1)\).

**Completion of the Proof of Theorem 3.12** In this part, we complete the proof of Theorem 3.12 by analyzing the function \(F(\Delta_n)\). Recall that

\[
f(\Delta_n, z) = \Delta_n - z - \log_q e(q - 1)q^{\Delta_n - z - 1}
\]

and the value of \(z\) that achieves the only maximum of \(f(\Delta_n, z)\) over the real numbers is \(z_0 = \log_q (q - 1) - 1 + \Delta_n\). Since we are interested in integers, we next investigate which \(z \in \mathbb{Z}\) maximizes \(f(\Delta_n, z)\). We consider the following two cases.

1. \(q = 2\): In this case, \(z_0 = \Delta_n - 1\), and therefore \(-2 < z_0 \leq -1\), so the maximal value is achieved for \(z \in \{-2, -1\}\). Note that

\[
\frac{\partial f}{\partial \Delta_n} \bigg|_{z=-1} = 1 - (q - 1)q^{\Delta_n - z - 1} = 1 - 2^{\Delta_n - z - 1},
\]

and

\[
\frac{\partial f}{\partial \Delta_n} \bigg|_{z=-2} = 1 - 2^{\Delta_n + 1} \leq 0.
\]

Hence \(f(\Delta_n, -1)\) is increasing and \(f(\Delta_n, -2)\) is decreasing when \(-1 < \Delta_n \leq 0\). Moreover, they meet in \(\Delta_n = \log(\ln 2) \approx -0.53\). From the
analysis above we summarize that when $q = 2$,

$$F(\Delta_n) = \begin{cases} 
    f(\Delta_n, -2), & \text{for } -1 < \Delta_n \leq \log(\ln 2) \\
    f(\Delta_n, -1), & \text{otherwise}
\end{cases}$$

To understand which $\Delta_n$ achieves the maximal $F(\Delta_n)$ we can consider the only two options $\Delta_n \in \{-1, 0\}$ and since we get $f(-1, -2) = f(0, -2) = 1 - \log e$ we can conclude that

$$C_1(n, q) \lesssim \frac{2^n}{4n} \cdot 2^{1 - \log e} = \frac{2^n}{2\epsilon n}.$$  

Lastly, when $n \to \infty$ over any subsequence of $n$ that satisfies $\lim_{n \to \infty} \Delta_n \in \{-1, 0\}$, we have that

$$C_1(n, q) \approx \frac{2^n}{2\epsilon n}.$$

2. $q > 2$:

For any $q > 2$ and $-1 < \Delta_n \leq 0$ we have that $-2 < z_0 < 0$, and so the maximum over $z \in \mathbb{Z}$ is achieved by one of the options $z \in \{-2, -1, 0\}$. Note that

$$f(\Delta_n, -1) - f(\Delta_n, -2) = \Delta_n + 1 - \log_q e(q - 1)q^{\Delta_n} - (\Delta_n + 2 - \log_q e(q - 1)q^{\Delta_n + 1}) = -1 + \log_q e(q - 1)^2 q^{\Delta_n} \geq -1 + \frac{(q - 1)^2}{q \ln q} > 0,$$

where the last step holds for any $q > 2$. Therefore, we are only left with maximizing over $z \in \{-1, 0\}$. The functions $f(\Delta_n, -1)$ and $f(\Delta_n, 0)$ meet when $\Delta_n = \delta_0 = -\log_q \frac{(q-1)^2}{q \ln q}$. Since for $q > 2$, $1 \leq \frac{(q-1)^2}{q \ln q} < q$, we have $-1 < \delta_0 \leq 0$. In addition,

$$\frac{\partial f}{\partial \Delta_n} \bigg|_{z=0} = 1 - \frac{q-1}{q} \cdot q^{\Delta_n} > 0,$$
hence $f(\Delta_n, 0)$ is increasing for $-1 < \Delta_n \leq 0$. Also, 

$$\frac{\partial f}{\partial \Delta_n} \bigg|_{z=-1,\Delta_n=\delta_0} = 1 - (q - 1) \cdot q^{\delta_0} = 1 - \frac{q \ln q}{q - 1} < 0,$$

so $f(\Delta_n, -1)$ is decreasing at the point $\Delta_n = \delta_0$. We conclude that for $\Delta_n \leq \delta_0$, $f(\Delta_n, 0) \leq f(\Delta_n, -1)$ and for $\Delta_n \geq \delta_0$, $f(\Delta_n, -1) \leq f(\Delta_n, 0)$. That is,

$$F(\Delta_n) = \begin{cases} f(\Delta_n, -1), & \text{for } -1 < \Delta_n \leq \delta_0 \\ f(\Delta_n, 0), & \text{otherwise} \end{cases}$$

We next study that values of $\Delta_n$ that maximize the function $F(\Delta_n)$. We showed that $f(\Delta_n, 0)$ is increasing so its maximal value is achieved when $\Delta_n = 0$ and it is $F(0) = f(0, 0) = \frac{q-1}{q} - 1$. The maximal value of $f(\Delta_n, -1)$ is achieved when

$$\frac{\partial f}{\partial \Delta_n} \bigg|_{z=-1} = 0,$$

that is, when $\Delta_n = \delta_1 = -\log_q (q - 1)$. Since $-1 \leq \delta_1 \leq \delta_0$ we get $F(\delta_1) = f(-1, \delta_1) = 1 - \log_q (e(q - 1))$ and $F(\delta_1) \geq F(0)$. We are now ready to summarize and conclude that

$$C_1(n, q) \lesssim \frac{q^n}{n} \left(\frac{q-1}{q}\right)^2 \cdot q^{F(\delta_1)} = \frac{q^n}{n} \frac{q - 1}{eq}$$

and when $n \to \infty$ over any subsequence of $n$ that satisfies $\lim_{n \to \infty} \Delta_n \in \{-1, 0\}$ we have

$$C_1(n, q) \approx \frac{q^n}{n} \frac{q - 1}{eq}.$$
Appendix C

Missing Proofs: Section 4.1

**Lemma 4.1** Let $n, k, d$ be positive integers such that $d \leq k \leq n$. Then, there exists a constant $C > 0$ such that for $n$ large enough

$$a_q(n, k, d) \leq q^{n - C \frac{(n-2k+2)}{2} - 1}.$$

**Proof** We consider the set $A_q(2k, k, d) \lfloor \frac{n}{2k} \rfloor$, that is, the set of vectors which are a concatenation of $\lfloor \frac{n}{2k} \rfloor$ vectors from $A_q(2k, k, d)$. We then append it with the set of all length-$\langle n \rangle$ vectors. The resulting set of length-$n$ vectors is denoted by $B_q(n, k, d) = A_q(2k, k, d) \lfloor \frac{n}{2k} \rfloor \Sigma_q^{\langle n \rangle}$. Note that $A_q(n, k, d) \subseteq B_q(n, k, d)$ and

$$|B_q(n, k, d)| = a_q(2k, k, d) \lfloor \frac{n}{2k} \rfloor q^{\langle n \rangle}. \quad (C.1)$$

Hence,

$$a_q(n, k, d) \leq a_q(2k, k, d) \lfloor \frac{n}{2k} \rfloor q^{\langle n \rangle}. \quad (C.1)$$

Let $b(k)$ be the number of vectors of length $2k$ with a subsequence of the form $[1, q - 1]^d u [1, q - 1]^d$ where $u$ is a vector of length $k$ and weight smaller than $d$ and $[1, q - 1]^d$ corresponds to a sequence of $d$ non zero symbols. The value of $b(k)$ is given by

$$b(k) = q^{2k - (k+2d)}(q - 1)^{2d}(2k - (k+2d) + 1) \sum_{i=0}^{d-1} \binom{k}{i}$$

$$= q^{k - 2d}(q - 1)^{2d}(k - 2d + 1) \sum_{i=0}^{d-1} \binom{k}{i}.$$
All those length $2k$ vectors are not included in the set $A_q(2k, k, d)$. Therefore,

$$a_q(2k, k, d) \leq q^{2k} - b(k) \leq q^{2k} - q^{k-2d}(q-1)^{2d}(k-2d+1) \sum_{i=0}^{d-1} \binom{k}{i}.$$  \hspace{1cm} (C.2)

We denote $B = \sum_{i=0}^{d-1} \binom{k}{i}$. Note that $B = \Theta(k^{d-1})$, when $d$ is fixed and $k$ is arbitrary large. Combining inequalities (C.1) and (C.2) we get

$$a_q(n, k, d) \leq (q^{2k} - q^{k-2d}(q-1)^{2d}(k-2d+1)B)\frac{1}{e^n} q^{(n)2k}$$

$$= (q^{2k}(1 - \frac{(k-2d+1)(q-1)^{2d}B}{q^{k+2d}}))\frac{1}{e^n} q^{(n)2k}$$

$$= q^n(1 - \frac{(k-2d+1)(q-1)^{2d}B}{q^{k+2}})\frac{1}{e^n} q^{(n)2k}$$

$$\leq q^n(1 - \frac{(k-2d+1)(q-1)^{2d}B}{q^{k+2}})\frac{1}{e^n} q^{(n)2k}$$

$$(a) \leq q^n(\exp(-\frac{(k-2d+1)(q-1)^{2d}B}{q^{k+2}}))\frac{1}{e^n} q^{(n)2k}$$

$$\leq q^n(\exp(-\frac{(k-2d+1)(q-1)^{2d}B}{q^{k+2}}))\frac{1}{e^n} q^{(n)2k}$$

$$(b) \leq q^n(\exp(-\frac{(k-2d+1)(q-1)^{2d}B}{q^{k+2}}))\frac{1}{e^n} q^{(n)2k}$$

$$(c) \leq q^n\left(1 - \frac{(k-2d+1)(q-1)^{2d}B}{q^{k+2}}\right)\frac{1}{e^n} q^{(n)2k}$$

$$(d) \leq q^n\left(1 - \frac{(k-2d+1)(q-1)^{2d}B}{q^{k+2}}\right)\frac{1}{e^n} q^{(n)2k}$$

$$(e) \leq q^n\left(1 - \frac{(k-2d+1)(q-1)^{2d}B}{q^{k+2}}\right)\frac{1}{e^n} q^{(n)2k}$$

where (a) results from the inequality $1 - x \leq e^{-x}$ for all $x$ and (b) since there exists a constant $C$ such that for $n$ large enough the inequality holds.

**Lemma 4.2** For all $n' \leq n$, given any vector $x \in \Sigma_n'$ Algorithm 2 outputs a $(d, F(n, d))$-WWL vector $y \in \Sigma_n^{n'}$ such that $x$ can be uniquely reconstructed given $y$. The time and space complexity of the algorithm and its inverse is $\Theta(n)$.

**Proof** First, we notice that according to the choice of $F(n, d)$ and $C$ the length of $y$ does not change throughout the execution of the algorithm, therefore $y \in \Sigma_{n'}$. There exists an index $1 \leq t \leq n'$ such that the output vector $y$ satisfies $y = (y_1, 1^d, y_{t+d+1}^{n'}$, where $y_1$ is the remainder of $x$ after removing the low weight vectors and $y_{t+d+1}^{n'}$ is the list of pointers of the form $p(i)\cdot t(1) \cdots t(d-1)$ representing the indices of the low weight subvectors and the positions of the ones inside each subvector.

To reconstruct $x$ we first locate the index $t$ by scanning $y$ from the right. We read the two rightmost symbols of $y$, if they are $01$ we conclude that the following
\( \mathcal{F}(n, d) - 2 \) symbols are a pointer of the form \( p(i)t(1) \cdots t(d - 1)01 \), we skip them and repeat that process until we encounter with two symbols 11. We then construct the original \( \mathbf{x} \) by inserting proper low weight vectors of length \( \mathcal{F}(n, d) \) to the remainder part \( y'_1 \). The positions of the vectors we insert and the positions of the ones within them are determined according to the indices listed in the pointers part in \( y^{n'+d}_{t+d+d+1} \).

We next show that \( y \) does not contain a vector of length \( \mathcal{F}(n, d) \) of weight less than \( d \). The remainder part \( y'_1 \) does not contain such a vector since we removed all low weight vectors within the while loop. Also, the separating part \( 1^d \) ensures that there is no vector of length \( \mathcal{F}(n, d) \) that originates in \( y'_1 \) and ends in \( y^{n'+d}_{t+d+d+1} \). Next we contradict the case of low weight vector in the addressing part. Recall that the structure of \( y^{n'+d}_{t+d+d+1} \) is a concatenation of pointers of the form \( p(i)t(1) \cdots t(d - 1)01 \). Note that the weight of every index \( p(i) \) or \( t(j) \) is at least 1. Every vector of length \( \mathcal{F}(n, d) \) in \( y^{n'+d}_{t+d+d+1} \) consists at least \( d - 1 \) full indices (counting both \( p_i \) and \( t_j \)) and an additional one from the appended 01 pairs. Therefore the total weight of such vectors is at least \( d \) as required.

Lastly, the algorithm’s complexity is \( \Theta(n) \) since the complexity of every pointer update \( \Theta(\log n) \) and there are at most \( n/\log n \) updating operations.
Appendix D

Missing Proofs: Section 5.1

Claim 5.1 For $a \in \Sigma^n_q, b \in \Sigma^m_q$: $d_E(a, b) = n + m - 2\ell(a, b)$.

Proof We say a series of insertions and deletions that transforms $a$ to $b$ is canonic if all of its deletions are of symbols from the original $a$. In other words, there were no new symbols that were inserted and then deleted. To determine the edit distance of $a$ and $b$ we are interested in the minimal length of a series that transforms $a$ to $b$. We therefore imitate our discussion, without loss of generality, to canonic series, because if a series is not canonic there exists a shorter equivalent canonic series to transform $a$ to $b$.

Any canonic series is associated with a common subsequence of $a$ and $b$ which is received by applying all the deletions in the series on $a$. We denote this common subsequence by $x \in \Sigma^\ell_q$. The number of deletions in the initial canonic series is $n - \ell$ and number of insertions is $m - \ell$, so the length of the series is $n + m - 2\ell$. Equivalently, any common subsequence of $a$ and $b$, denoted by $x \in \Sigma^\ell_q$ is associated with a canonic series of length $n + m - 2\ell$ which consists deletions of symbols of $a$ that do not belong to $x$ followed by insertions of symbols of $b$ that do not belong to $x$. Thus, there exists a common subsequence of $a$ and $b$ of length $\ell$ if and only if there exists a canonic series from $a$ to $b$ of length $n + m - 2\ell$. From the definition of edit distance, the claim follows.

Claim 5.2 For $a \in \Sigma^n_q, b \in \Sigma^n_q, c \in \Sigma^m_q, d \in \Sigma^m_q$: $d_E(ac, bd) \geq \max\{d_E(a, b), d_E(c, d)\}$.

Proof Without loss of generality we assume that $d_E(a, b) = \max\{d_E(a, b), d_E(c, d)\}$. We set $\ell(a, b) = \ell_1$, $\ell(ac, bd) = \ell_2$. For any common subsequence $x$ of $ac, bd$ of
length $\ell \geq m$, the prefix of length $\ell - m$, $x_1^{\ell - m}$ is a common subsequence of $a, b$ hence $\ell_1 \geq \ell - m$. If we take $x$ to be an lcs of $ac, bd$ we get that $\ell_1 \geq \ell_2 - m$. Combining this with Claim 5.1 we have

$$d_E(ac, bd) = 2n + 2m - 2\ell_2 \geq 2n + 2m - 2\ell_1 - 2m = d_E(a, b).$$

**Claim 5.3** For $a \in \Sigma_q^n, b \in \Sigma_q^n, c \in \Sigma_q^m, d_E(ac, b) \geq d_E(a, b) / 2$.

**Proof** We will show that

$$\max\{d_E(a, b) - m, m\} \leq d_E(ac, b),$$

since every $m$ satisfies $d_E(a, b) / 2 \leq \max\{d_E(a, b) - m, m\}$ the claim follows directly. An lcs of $ac, b$ is a subsequence of $b$, therefore $\ell(ac, b) \leq n$ and from Claim 5.1, $d_E(ac, b) = n + n + m - 2\ell(ac, b) \geq 2n + m - 2n = m$.

We set $\ell(a, b) = \ell_1, \ell(ac, b) = \ell_2$. Note that since $a$ and $b$ are of the same length, $d_E(a, b)$ is even. As in the proof of claim 5.2, $\ell_1 \geq \ell_2 - m$. Combining it with Claim 5.1 we have $d_E(ac, b) = 2n + m - 2\ell_2 \geq 2n + m - 2\ell_1 - 2m = d_E(a, b) - m$.

**Claim 5.4** For $a = 0^n, b \in \Sigma_q^n, d_E(a, b) = 2w_H(b)$.

**Proof** Since $a = 0^n$ the lcs of $a, b$ is $0^{n - w_H(b)}$ and $\ell(a, b) = n - w_H(b)$. From Claim 5.1 we get $d_E(a, b) = 2w_H(b)$. 

66
Appendix E

Missing Proofs: Section 5.2

**Theorem 5.11** The code $C_4(n,k)$ is a balanced MU code, and for an integer $k = \log n + a$, $|C_4(n, \log n + a)| \gtrsim C \frac{2^n}{n \sqrt{n}}$, where $C = \frac{2^a - 1}{2^{2a+1} \sqrt{2\pi}}$.

**Completion of the Proof of Theorem 5.11**

$$|C_4(n,k)| \geq \binom{n'}{n/2 - 2} - (n' - k + 1) \binom{n' - \frac{k}{2} - 2}{n/2 - 2}$$

$$= \binom{n' - k}{n/2 - 2} \left[ \prod_{i=0}^{k-1} \frac{n' - i}{n' - (n/2 - 2) - i} - n' + k - 1 \right]$$

$$\geq \binom{n - 2k - 2}{n/2 - 2} \left[ \prod_{i=0}^{k-1} \frac{n - k - 2 - i}{n/2 - k - i} - n + 2k + 1 \right]$$

$$= \binom{n - 2k - 2}{n/2 - 2} \cdot \frac{2^k - n + 2k + 1}{n} \cdot \frac{2^k + 2k + 1}{n} - 1.$$
Note that
\[
\binom{n - 2k - 2}{n/2 - 2} = \binom{n - 2k - 2}{\frac{n}{2} - 2} \frac{(\frac{n}{2} - 2)!}{(\frac{n}{2} - 2)!} \frac{(\frac{n}{2} - k - 1)!}{(\frac{n}{2} - k)!} \\
= \binom{n - 2k - 2}{\frac{n}{2} - 2} \frac{(\frac{n}{2} - k - 1)!}{(\frac{n}{2} - 2)!} \frac{(\frac{n}{2} - 2)!}{(\frac{n}{2} - k)!} \\
= \binom{n - 2k - 2}{\frac{n}{2} - 2} \prod_{i=1}^{k-1} \frac{n}{2} - 1 - i \\
\geq \binom{n - 2k - 2}{\frac{n}{2} - 2} \left( \frac{n}{2} - 2k + 1 \right)^{k-1} \\
= \binom{n - 2k - 2}{\frac{n}{2} - 2} \left( 1 - \frac{4k - 2}{n} \right)^{k-1}. \\
\]

For \( k = \log n + a \) we get the following
\[
= \binom{n - 2k - 2}{\frac{n}{2} - 2} \left( 1 - \frac{4k - 2}{n} \right)^{n \approx \frac{(4k - 2)(k - 1)}{n}} \\
\approx \binom{n - 2k - 2}{\frac{n}{2} - 2} e^{-\frac{(4k - 2)(k - 1)}{n}} \\
\approx \frac{2^{n - 2k - 2} + 1}{\sqrt{2\pi n} - 2k - 2} \\
\geq \frac{2^n}{2^{2a+1} n^2 \sqrt{2\pi n}}.
\]

Finally, we conclude that
\[
|C_4(n,k)| \approx \frac{2^n}{2^{2a+1} n^2 \sqrt{2\pi n}} \left[ \frac{2^k + 2k + 1}{n} - 1 \right] \\
= \frac{2^n}{2^{2a+1} n \sqrt{2\pi n}} \left[ \frac{n2^a + 2 \log n + 2a + 1}{n} - 1 \right] \\
\approx \frac{2^n}{n \sqrt{n} 2^{2a+1} \sqrt{2\pi}}. \\
\]
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Bibliography


צפנים ללא תאים הדדי

עבור אחסון מבוסס DNA

מאיה לוי
צפונים ללא האימוג הדרדרית
עבור אחסון מבוסס DNA
היתוך על מחקר
לשם مليוי חלקי של הדרישות לקבלת תואר
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מאתיה לוי

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כרת תודה
ברצוני להודות למנחה שלי, פרופמר איתן יעקובי, על המецוותיות, המוסריות, התמיכה והדרכה לאלוף המדריך.

אני מודה לテכניון על התמיכה כספית בהשתלמותי

העבודה הוצגה בכנס "International Symposium on Information Theory"
באאכן, גרמניה, יולי 2017
The Mutually uncorrelated codes (Mutually uncorrelated codes), or simply MU codes, are a class of codes in which the correlation between the members of a group of codes is zero. In other words, the ciphertexts are mutually uncorrelated.

Such codes were first studied in the 1960s for synchronization purposes, and they have since attracted renewed attention with applications in random access DNA storage.

The potential of DNA as a storage medium was first recognized in the 1960s due to the unique properties of DNA molecules. At that time, however, the technology for creating and reading DNA molecules was not yet developed enough for such systems. Over the years, as the understanding of DNA and its manipulation has advanced, significant progress has been made in developing practical applications of DNA storage.

DNA molecules are composed of four types of bases (nucleotides): adenine (A), cytosine (C), guanine (G), and thymine (T). They can be considered as a linear sequence of characters from the alphabet \{A, C, G, T\}.

A breakthrough came in 2012 when a method for DNA-based storage was proposed. The writing in the described system was done chemically, and the reading was done using DNA-sequencing technologies. These technologies are still not competitive with existing storage technologies in terms of cost and access time, but their applications are expected to continue to push and develop their capabilities, as we have seen in recent years.

The length of DNA molecules created in the synthesis process is limited to a few hundred characters, so to store a large amount of data, a large number of DNA molecules were stored unordered or randomly. In order to read a piece of information, a specific sequence of DNA must be reconstructed, which leads to a non-random access.

Since 2012, several similar systems have been implemented, and in 2016, a system for DNA storage was proposed that solves the access problem. The aim of the proposed scheme is to enable random access.

The constraints that a group of writing sequences must meet are described in our work from 2016. In our research, we focus on three of these constraints:

1. No mutual correlation is required to prevent overlap between two sequences, thereby reducing the probability of misidentifying the required sequence.

In this work, we used some of the above constraints in the construction of a new DNA-based storage system.


We expanded the research also to MU codes in a similar way. The mechanism. Let c be an unstructured bi-unique codeword and assume that it is a codeword of length n.

We used the known algorithm of Varshamov Tenengolts for the correction of a deletion and an insertion. In this way, we extended it so that it also satisfies the mutual exclusion constraint.

Finally, for balanced MU codes, we showed that the minimal disparity is 1.5 log n, and for balanced MU codes, the disparity is 2 log n.

To enable efficient encoding and decoding, we presented a method with disparity 2 log n by using Knuth's algorithm for balancing binary words.