Exact Reconstruction from Insertions in Synchronization Codes

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Abstract—This work studies problems in data reconstruction, an important area with numerous applications. In particular, we examine the reconstruction of binary and nonbinary sequences from synchronization (insertion/deletion-correcting) codes. These sequences have been corrupted by a fixed number of symbol insertions (larger than the minimum edit distance of the code), yielding a number of distinct traces to be used for reconstruction. We wish to know the minimum number of traces needed for exact reconstruction. This is a general version of a problem tackled by Levenshtein for uncoded sequences.

We introduce an exact formula for the maximum number of common supersequences shared by sequences at a certain edit distance, yielding an upper bound on the number of distinct traces necessary to guarantee exact reconstruction. Without specific knowledge of the codewords, this upper bound is tight. We apply our results to the famous single deletion/insertion-correcting Varshamov-Tenengolts (VT) codes and show that a significant number of VT codeword pairs achieve the worst-case number of outputs needed for exact reconstruction. We also consider extensions to other channels, such as adversarial deletion and insertion/deletion channels and probabilistic channels.

Index Terms—Insertions and deletions; sequence reconstruction; edit distance; synchronization codes.

I. INTRODUCTION

This paper is concerned with the problem of reconstructing sequences (selected from error-correcting codes) from many corrupted copies, referred to as traces. In particular, we are concerned with sequences from insertion/deletion-correcting codes that have been corrupted by insertion errors. We are interested in answering the following question. If we are given a codeword from an \((\ell - 1)\)-insertion/deletion-correcting \(q\)-ary code of length \(n\), what is the minimum number \(M\) of distinct traces (produced by a \(t\)-insertion channel) that always allows for exact reconstruction of the codeword?

Our main result is that exact reconstruction is always possible when \(M\) is at least

\[
\sum_{j=0}^{t} \sum_{i=0}^{t-j} \binom{2j}{j} \binom{n+t}{i+2j} (q-1)^{t+j-i} + 1.
\]

Without further knowledge of the code, the bound in (1) is tight. In other words, if \(M\) is smaller than the right hand side of (1), there exists a pair of sequences of length \(n\) with \(M\) common traces, so that reconstruction cannot be guaranteed, and, moreover, these sequences are at edit distance at least \(2\ell + 1\) (we make this terminology precise later on), and could thus be part of an \((\ell - 1)\)-insertion/deletion-correcting code. The result generalizes the uncoded version from [1] and [2], which can be recovered by taking \(\ell = 1\) in (1) and simplifying the result through a series of combinatorial identities. It is surprising that exact formulas like (1) exist, given the paucity of exact expressions in insertion and deletion problems. Before we further discuss results such as (1), we briefly introduce the context for this work.

Data reconstruction from traces is an important problem with numerous applications, including data recovery, genomics and other areas of biology, chemistry, sensor networks, and many others. The general problem of reconstruction is broadly divided into probabilistic and adversarial variants. In the probabilistic version, the traces are formed by passing the data through a noisy channel (typically an edit channel with certain deletion and insertion probabilities) and the goal is to reconstruct the data to within a certain error probability. In [3], an algorithm is introduced based on bitwise majority alignment that reconstructs an original sequence of length \(n\) with high probability from \(O(\log n)\) traces when the deletion probability is \(O(\log^2 \frac{1}{n})\). These results were extended to the deletion/insertion channel in [4] and improved upon in [5]. Sequence reconstruction with constant deletion probability was also studied in [6], where the authors showed that when the sequence length is \(n\), reconstruction is possible with high probability from a number of traces polynomial in \(n\) in time polynomial in \(n\).

The adversarial variant, which we are concerned with in this paper, allows for traces formed from a worst-case number of errors and seeks to determine what is the smallest number of traces needed for zero-error reconstruction [1]. This setup for sequence reconstruction has also been applied to associative memories [7]. In these memories, each entry is associated with neighboring entries; when searching for a particular entry, “clues” are given in the form of such neighboring entries.
This notion leads to a generalization of sequence reconstruction; here, the question becomes how many sequences are associated with (i.e., of maximum Hamming distance from) three or more sequences. The resulting intersection is called an output set, and the size of the maximum output set is the uncertainty of the memory. This line of research was extended in [10], which studies efficient codes for information retrieval in memories with small uncertainty, and in [11], where the number of input clues is varied. We note that all of these works target the Hamming metric.

By contrast, we are specifically interested in the following problem: if a codeword from a synchronization (specifically, insertion/deletion-correcting) code, i.e., a code with a certain minimum edit distance, is repeatedly transmitted through a noisy channel, how many distinct channel outputs (traces) are necessary for zero-error reconstruction? This question is indeed meaningful; consider, for example, phylogenomics, where we wish to reconstruct the genetic sequence of an ancestor organism from a large number of sequences of evolutionary descendants. Each of the descendant sequences is formed from a number of base pair insertions. The related question of how to efficiently perform the reconstruction is tackled from a coding-theoretic point of view in the recent work [12].

Moreover, reconstruction of encoded data has a natural application to data storage. We can partition the operational lifetime of a disk drive, memory, or other data storage device into two periods. The first period is regular, short-term use; popular devices and common error-correcting codes all target this scenario. Here, a small number (often one) of channel outputs are used to read the data. The second period refers to extremely long-term use of the device, well beyond the guaranteed operating lifetime. In this case, many reads can be performed, resulting in a large number of traces that can be used to recover the original data. This type of very long term use is increasingly relevant. For example, DNA storage is targeted as a medium to store data for 10^4 or more years, and, indeed, over the long term, DNA sequences are affected by insertions and other errors that can be modeled by insertions (duplications, tandem/block duplications, block insertions) [13]. The present paper studying reconstruction from insertions can therefore be viewed as complementary to the many recent efforts studying coding for data storage in DNA [14]–[22].

Our work also joins recent coding-theoretic studies on insertions and deletions, such as [23]–[25]. Our main result [1], as we shall see, computes a maximal common intersection of error balls centered at codewords. A similar result, but in the easier case of codes correcting substitution errors, has been used to provide improved Gilbert-Varshamov bounds in [26] and [27], in the cases of binary and non-binary codes, respectively. Our work can thus be seen as a building block towards providing improve cardinality bounds for insertion/deletion-correcting codes.

In a classic paper [1], Levenshtein explored several variations of the reconstruction problem, studying both adversarial and probabilistic channels and exact and approximate reconstruction. However, the problem of reconstructing sequences affected by insertions and deletions in the case where the sequences are themselves part of a code (e.g., have a certain minimum edit distance) was left open. We tackle this problem for the insertion case in the current work. We target insertions for two reasons. First, insertions (and deletions) are edit errors, which are of particular interest as we often wish to reconstruct strings. In keeping with our biology theme, as described above, we note that insertions are a common type of mutation affecting genetic sequences. Second, unlike deletions, insertions offer symmetries that allow a tractable search for exact formulas.

The remainder of this paper is organized in the following way. In the next section, we introduce our problem setup, discuss prior work, and describe notation. In Section III, we prove and interpret the general, non-binary common supersequences problem. We also discuss important special cases and corollaries. In Section IV, we apply our result to the single deletion/insertion-correcting Varshamov-Tenengolts (VT) codes. Finally, in Section V, we consider extensions to the adversarial deletion and insertion/deletion channels and briefly consider probabilistic channels. We conclude the paper in Section VI.

II. PRELIMINARIES

A. Problem setup

Levenshtein provided a general framework for reconstruction problems in [1]. He considered the problem of reconstructing a sequence \( X \in V \subseteq \mathbb{F}_q^n \) for a set \( V \) and an alphabet \( \mathbb{F}_q \). In this case, the traces were selected from \( B_t(X,H) \), the ball produced by applying up to \( t \) single errors from a set of error functions \( H \) to \( X \). Levenshtein noted that it is always possible to exactly reconstruct \( X \) given \( N^H_q(V,t) + 1 \) distinct elements of \( B_t(X,H) \), where \( N^H_q(V,t) \) is defined by

\[
N^H_q(V,t) := \max_{x,z \in V, x \neq z} |B_t(X,H) \cap B_t(Z,H)|. \tag{2}
\]

The idea here is that \( X \) cannot share any more than \( N^H_q(V,t) \) traces with any other element of \( V \), so that \( N^H_q(V,t) + 1 \) distinct traces uniquely determine \( X \). Thus, the problem of exact reconstruction of sequences can be identified with the combinatorial problem of counting the (maximum number of) common distorted sequences. In [1], expressions were given for \( N^H_q(V,t) \) for many choices of error sets \( H \). In [2], the focus was specifically on deletion and insertion channels. In this paper, we focus on the case of insertion channels, so that \( H \) is made up of single symbol insertions, and we wish to reconstruct \( X \) from its supersequences (sequences formed from \( X \) by insertions). In [1], an expression was provided for \( N^H_q(V,t) \) in this scenario, but only for the specific case of \( V = \mathbb{F}_q^n \), the uncoded case. For this problem setup, we may write the balls \( B_t(X,H) \) as insertion balls \( I_t(X) \) and denote the expression \( N^H_q(\mathbb{F}_q^n,t) \) as \( N^+_q(n,t) \). In [1], \( N^+_q(n,t) \) was shown to be

\[
N^+_q(n,t) = \sum_{i=0}^{t-1} \binom{n+t}{i} (q-1)^i (1-(-1)^{t-i}). \tag{3}
\]
An example is shown in Figure 1 for codewords in insertion-correcting codes. Without specifying a structure of the code, we can always perform exact reconstruction if we know \( \ell \) sequences are known outside of the single insertion case. Instead, we can always perform exact reconstruction if we know \( N_q^{+}(C',t) + 1 \) distinct supersequences of \( X \), where

\[
N_q^{+}(C',t) = \max_{X,Z \in C',X \neq Z} |I_t(X) \cap I_t(Z)|.
\]

Computing this maximal intersection requires knowing the structure of the code \( C' \). This is challenging, since few such codes are known outside of the single insertion case. Instead, we focus on deriving an expression for the maximum number of common supersequences for sequences at a minimum particular edit distance \( 2\ell \).

\[
N_q^{+}(n,t,\ell) = \max_{X,Z \in \mathbb{F}_q^n, d_s(X,Z) \geq 2\ell} |I_t(X) \cap I_t(Z)|.
\]

An example is shown in Figure 1 for \( n = 4 \) and \( t = 2 \). Specifically, we prove that \( N_q^{+}(n,t,\ell) \) is given by

\[
\sum_{j=0}^{t} \sum_{i=0}^{t-j} \binom{2j}{j} \binom{n+t}{i} (q-1)^{i(-1)^{t+j-i}}.
\]

Evaluating the above expression provides an upper bound on the number of channel outputs needed for reconstruction of codewords in insertion-correcting codes. Without specifying a particular code construction, this bound is the best possible. Note that Levenshtein’s formula \( N_q^{+}(n,t) \) in [27] can be written as \( N_q^{+}(n,t) = \max_{t \geq 1} N_q^{+}(n,t,\ell) \).

We will see, in fact, that this maximum is always attained at \( \ell = 1 \). In other words, the maximum number of common supersequences occurs for sequences that are as “close” as possible.

As part of our study, we provide an even more general version of this result where we allow the sequences \( X \) and \( Z \) to have different lengths. This result can be interpreted as a generalization of Levenshtein’s formula \( N_q^{+}(n,t) \).

### B. Notation

We introduce some useful notation. Let \( \mathbb{F}_q \) denote the set \( \{0,1,\ldots,q-1\} \) for \( q \geq 2 \) and \( [a,b] \) denote the set \( \{a,a+1,a+2,\ldots,b-1,b\} \) if \( a \leq b \). We denote sequences with upper-case letters and individual symbols with lower-case letters, so that, for example, \( X = x_1x_2\ldots x_n \in \mathbb{F}_q^n \) while \( x_i \in \mathbb{F}_q \) for \( 1 \leq i \leq n \). We write \( XY \) for the concatenation of sequences \( X \) and \( Y \); similarly, \( aX \) is the concatenation of a symbol \( a \) with sequence \( X \). We sometimes use the notation \( XS \) where \( S \) is a set. In this case, \( XS \) refers to the set that contains the concatenation of \( X \) with all sequences in \( S \), \( \{XY : Y \in S\} \).

We use the generalized binomial coefficient: for \( a,b \in \mathbb{Z} \),

\[
\binom{q}{a} = \frac{(a-1)\ldots(a-b+1)}{b!}.
\]

We assume \( 0! = 1 \). We set \( \binom{q}{b} = 0 \) for \( b < 0 \). We also use the convention \( \binom{a}{0} = 1 \) for all \( a \in \mathbb{Z} \) and \( \binom{a}{b} = 0 \) if \( a = 0 \) and \( b > 0 \). We sometimes rely on the useful fact that \( \binom{a}{b} = 0 \) if \( a > 0 \) and \( a < b \).

If \( n \geq v \geq 0 \), \( Z \in \mathbb{F}_q^{n-v} \) is a subsequence of \( X \in \mathbb{F}_q^n \) if \( Z \) can be formed by \( X \) by deleting \( v \) symbols. If \( n = v \), \( Z \) is the empty sequence, with length 0. Similarly, for \( t \geq 0 \), \( W \in \mathbb{F}_q^{n-t} \) is a supersequence of \( X \) if it can be formed by \( t \) symbol insertions into \( X \). The set of all length \( n-v \) subsequences of \( X \) (also called the \( v \)-deletion ball centered at \( X \)) is denoted \( D_v(X) \); similarly, the set of all length \( n+t \) supersequences of \( X \) (the \( t \)-insertion ball centered at \( X \)) is written \( I_t(X) \).

In general, the size of \( D_v(X) \) depends on the sequence \( X \). For example, \( |D_1(X)| = \tau(X) \), where \( \tau(X) \) is the number of maximal-length runs of identical symbols in \( X \). On the other hand, \( |I_t(X)| \) does not depend on \( X \) for any \( t \geq 0 \). There is a formula for the size of the supersequence set that only depends on \( n,t \), and the alphabet size \( q \) [28].

\[
|I_t(X)| := I_q(n,t) = \sum_{i=0}^{t} \binom{n+t}{i} (q-1)^i.
\]

The distance between sequences \( X, Y \) can be measured with the edit distance \( d_e(X,Y) \). This distance is defined in the following way: \( d_e(X,Y) = k \) if \( k \) is the smallest number of insertions and deletions that can be used to transform \( X \) to \( Y \). Note that it is not necessary that \( X \) and \( Y \) have the same length for \( d_e(X,Y) \) to be defined. For example, take \( X = 00 \) and \( Y = 010 \). Then, \( d_e(X,Y) = 1 \), since we require just one insertion of a 1 into \( X \) to form \( Y \). If \( X = 110 \) and \( Y = 000 \), then \( d_e(X,Y) = 4 \). Note that our definition of edit distance does not include substitutions. Studying codes correcting insertions, deletions, and substitutions is a different, even more challenging problem.

A \( t \)-insertion-correcting code \( \mathcal{C} \) is a subset of \( \mathbb{F}_q^n \) such that if \( X, Y \in \mathcal{C} \) and \( X \neq Y \), then \( I_t(X) \cap I_t(Y) = \emptyset \). This is equivalent to requiring that \( \mathcal{C} \) has minimum edit distance \( 2t+2 \). We also note that a \( t \)-insertion-correcting code is also a \( t \)-deletion-correcting code (and also an \( a \)-deletion/b-insertion-correcting code for any pair \( (a,b) \) with \( a+b \leq t \)). This equivalence was proved in [28].

\footnote{The required minimum distance is \( 2t+1 \); however, since all the codewords in \( \mathcal{C} \) are of the same length, the distance between any codeword pair must be even, since to go from one codeword to another, there must be a deletion for every insertion. Thus the minimum distance is in fact \( 2t+2 \). For example, the minimum edit distance of the single deletion/insertion-correcting Varshamov-Tenengolts codes is 4.}
As described, we are concerned with computing the maximum number of common supersequences between sequences with edit distance of at least \(2\ell\) for \(\ell \geq 1\). That is, we are interested in the quantity \(N^+_q(n, t, \ell)\) defined as

\[
N^+_q(n, t, \ell) = \max_{X, Y \in \mathbb{F}_q^n} \{|I_t(X) \cap I_t(Y)|\}.
\]

We refer to \(n, t, \ell\) as the length, insertion, and distance arguments, respectively.

Additionally, in our results, we consider a more general version of the problem where the sequences need not be of the same length. One of the two sequences (\(Y\)) continues to be of length \(n\) while the common supersequences for distance \(\ell + 1\) is at least that for distance argument \(\ell\).

We also have another easy fact regarding distances. Let \(Z = z_1z_2\ldots z_n\). Then, \(I_t(Z)\) is the union of the set of sequences starting with \(z_1\) which can be formed by placing all \(t\) insertions into \(z_2\ldots z_n\) and the set of sequences starting with the element \(x \in \mathbb{F}_q \setminus z_1\) (which require \(x\) to be inserted at the head of \(Z\), leaving \(t - 1\) remaining insertions into \(Z\)). Formally, we have that

\[
I_t(Z) = z_1I_t(z_2z_3\ldots z_n) \cup \bigcup_{x \in \mathbb{F}_q \setminus z_1} xI_{t-1}(Z).
\]

Here, recall that \(xI_{t-1}(Z)\) refers to appending all the sequences in \(I_{t-1}(Z)\) to the element \(x\). The idea in Claim 3 can be exploited to derive recursive expressions for the number of common supersequences. A variant of the following observation was used by Levenshtein in [2]; we provide a proof for completeness.

Claim 3. If \(Z = z_1z_2\ldots z_n \in \mathbb{F}_q^n\) is a sequence and \(t \geq 1\), then

\[
I_t(Z) = z_1I_t(z_2z_3\ldots z_n) \cup \bigcup_{x \in \mathbb{F}_q \setminus z_1} xI_{t-1}(Z).
\]

We refer to \(n, t, \ell\) as the length, insertion, and distance arguments, respectively.

Additionally, in our results, we consider a more general version of the problem where the sequences need not be of the same length. One of the two sequences (\(Y\)) continues to be of length \(n\) while the common supersequences for distance argument \(\ell + 1\) is at least that for distance argument \(\ell\).

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Here, recall that \(xI_{t-1}(Z)\) refers to appending all the sequences in \(I_{t-1}(Z)\) to the element \(x\). The idea in Claim 3 can be exploited to derive recursive expressions for the number of common supersequences. A variant of the following observation was used by Levenshtein in [2]; we provide a proof for completeness.

Claim 4. Let \(n\) be a positive integer, \(t, k\) be non-negative integers with \(t \geq k\). Let \(X' \in \mathbb{F}_q^{n+t-k}\). Write \(X' = aX\) and \(Y' = bY\) with \(a, b \in \mathbb{F}_q\). Then, if \(a = b\),

\[
|I_k(X') \cap I_t(Y')| = |I_k(X) \cap I_t(Y)| + (q-1)|I_{k-1}(aX) \cap I_{t-1}(aY)|.
\]

If \(a \neq b\),

\[
|I_k(X') \cap I_t(Y')| = |I_k(X) \cap I_{t-1}(bY)| + |I_{k-1}(aX) \cap I_t(Y)| + (q-2)|I_{k-1}(aX) \cap I_{t-1}(bY)|.
\]

Proof: First we consider the case of \(a = b\). A common supersequence \(W'\) of \(X' = aX\) and \(Y' = bY\) either starts with \(a\) or an element in \(\mathbb{F}_q \setminus \{a\}\). If \(W'\) starts with \(a\), we write \(W' = aW\). Using Claim 3 \(W\) can be formed by \(k\) insertions into \(X\) and \(t\) insertions into \(Y\), so \(W\) is a common supersequence of \(X\) and \(Y\). That is, it is in the set \(I_k(X) \cap I_t(Y)\). Similarly, if \(W' = w_0W\) where \(w_0\) is one of the \(q\) elements in \(\mathbb{F}_q \setminus \{a\}\), it must be formed by inserting \(w_0\) at the head of \(X' = aX\) and at the head of \(Y' = bY\). Therefore, \(W\) is in \(I_k(X) \cap I_{t-1}(bY)\). There are thus \((q-1) \times |I_k(X) \cap I_{t-1}(bY)|\) choices for such supersequences \(W'\). This establishes (4).

For the case of \(a \neq b\), if \(W' = aW\), \(W\) can be formed from \(X'\) by inserting \(k\) elements into \(X\), while \(W'\) can be formed from \(Y'\) by inserting \(t\) elements into \(Y\). If \(W' = w_0W\), it is formed from \(X'\) by inserting \(b\) at the head and \(k-1\) elements into \(X' = aX\) while \(W'\) is formed from \(Y'\) by inserting \(t-1\) elements into \(Y\). Lastly, if \(W'\) starts with \(w_1\), one of the \((q-2)\) elements in \(\mathbb{F}_q \setminus \{a, b\}\), it is formed from \(X'\) by inserting \(w_1\) at the head and \(k-1\) elements into \(X' = aX\) and from \(Y'\) by inserting \(w_1\) at the head and \(t-1\) elements into \(Y' = bY\). The sets given by the three possibilities are disjoint, giving (7).

III. MAXIMUM NUMBER OF COMMON SUPERSEQUENCES

We are ready to prove the main result of the present work.
exact reconstruction given traces formed by $t = 15$ random symbol insertions.

The rest of this section is organized as follows. First, we provide the proof. Afterwards, we discuss important special cases of Theorem 5. In particular, we recover Levenshtein’s result for $N_q^+(n,t)$ from [1], along with a simplified binary result as special cases.

### A. Proof

We start with a roadmap for the proof of Theorem 5. First, we denote by $N_q^+(n,t,k,\ell)$ the formula

$$N_q^+(n,t,k,\ell) = \sum_{j=\ell}^{k} \sum_{i=0}^{k-j} \binom{t+1}{j} \binom{n+t}{i} (q-1)^j (-1)^{k-j}.$$  

If $t = k$,

$$N_q^+(n,t,\ell) = \sum_{j=\ell}^{k} \binom{t+1}{j} \binom{n+t}{i} (q-1)^j (-1)^{t-j}.$$  

We begin with some observations on Theorem 5. We are more interested in the formula in (9) compared to that in (8) because most insertion/deletion-correcting codes have equal-length codewords. We will later see (from the proof of Theorem 5) that the sequences that yield the maximum $N_q^+(n,t,\ell)$ common supersequences are those at distance precisely $2\ell$. This confirms the intuitive idea that the maximum number of common supersequences is attained by sequences as “close” as possible.

Results in the spirit of (9) encourage us to examine the tradeoff between code rate and reconstruction requirements. For example, the expression in (9) is decreasing in $\ell$, the code’s insertion/deletion-correcting ability. Increasing $\ell$ allows us to reconstruct with fewer and fewer traces, but this comes at the cost of decreased code rate. We show an example of this tradeoff for insertion/deletion-correcting codes of lengths $n = 30, 50, 70, 90$ in Figure 2. Here, we plot two curves for each code based on the error-correcting ability; the dashed curves show code rates (based on non-asymptotic upper bounds from [24]), while the thick curves show the percentage reduction (normalized to 1) in the number of traces needed to guarantee exact reconstruction given traces formed by $t = 15$ random symbol insertions.

We briefly discuss two important improvements of our proof technique compared to that of Levenshtein for the $N_q^+(n,t)$ result. First, we generalize the problem to the different-lengths case where $\ell$ need not be equal to $k$. This enables us to involve the second recursion (for sequences starting with different symbols) directly in the induction. This was not possible in Levenshtein’s proof, as the second recursion immediately breaks down into different-length cases (and formulas for such cases were not known); however, for $\ell = 1$, this issue can be overcome. In the cases $\ell > 1$, this is not possible. Interestingly, our approach mirrors some proofs in combinatorics, where it is easier to prove a general formula compared to a special case. In addition, we note that unlike in Levenshtein’s proof, we require a careful accounting of the recursions’ effects on the distance.

#### Lemma 6.

For $n \geq 1$, $q \geq 2$ and $t,k,\ell \geq 1$ with $t \geq k \geq \ell$, $N_q^+(n,t,k,\ell)$ satisfies the recursions

$$N_q^+(n,t,k,\ell) = N_q^+(n-1,t,k,\ell) + (q-1)N_q^+(n,t-1,k-1,\ell).$$
\[ N_q^+ (n, t, k, \ell) = N_q^+ (n, t-1, k, \ell) + N_q^+ (n, t-1, k, \ell - 1) + (q-2)N_q^+ (n, t-1, k-1, \ell). \]

We defer the proof of Lemma 7 to the appendix. We prove that the quantity \( N_q^+ (n, t, k, \ell) \) satisfies similar recursions:

**Lemma 7.** Let \( n \) and \( q \geq 2 \) be positive integers and \( t, k, \ell \) be non-negative integers such that \( t \geq k \geq \ell \). Let \( X' = aX, Y' = bY \) be two sequences satisfying \( X' \neq Y' \), \( X' \in \mathbb{F}_q^{n-t-k}, Y' \in \mathbb{F}_q^n \) and \( d_e(X', Y') = t - k + 2\ell \). Then, if \( a = b \),

\[ |I_k(X') \cap I_l(Y')| \leq N_q^+ (n-1, t, k, \ell) + (q-1)N_q^+ (n, t-1, k-1, \ell), \]

and if \( a \neq b \),

\[ |I_k(X') \cap I_l(Y')| \leq N_q^+ (n-1, t, k, \ell) + N_q^+ (n-1, t, k, \ell - 1) + (q-2)N_q^+ (n, t-1, k-1, \ell). \]

**Proof:** We are given \( X', Y' \) satisfying \( X' \neq Y' \), \( X' \in \mathbb{F}_q^{n-t-k}, Y' \in \mathbb{F}_q^n \), and \( d_e(X', Y') = t - k + 2\ell \). We have \( X' = aX \) and \( Y' = bY \), with \( a, b \in \mathbb{F}_q \). In the case \( a = b \), from (9),

\[ |I_k(aX) \cap I_l(bY)| = |I_k(aX) \cap I_l(aY)| = |I_k(X) \cap I_l(Y)| + (q-1)|I_{k-1}(aX) \cap I_{l-1}(aY)|. \]

(10)

We note that since \( d_e(X', Y') = t - k + 2\ell \) and \( X' = aX \), \( Y' = bY \), and \( X, Y \) and \( X', Y' \) must be at the same distance as \( X' \) and \( Y' \), so that \( d_e(X, Y) = t - k + 2\ell \). Thus, \( X \) is of length \( n-1 \) plus \( t - k \), \( Y \) is of length \( n-1 \) plus \( t - k \), and \( Y' \) is of length \( n-1 \) plus \( t - k \). Similarly, we have that \( |I_{k-1}(aX) \cap I_{l-1}(aY)| \leq N_q^+ (n-1, t-1, k-1, \ell) \).

We call this step argument matching, since we are computing the length, insertion, and distance arguments in order to produce the correct \( N_q^+ \) bound. Putting the two bounds into (10), gives

\[ |I_k(aX) \cap I_l(bY)| \leq N_q^+ (n-1, t, k, \ell) + (q-1)N_q^+ (n, t-1, k-1, \ell). \]

The next case is \( a \neq b \). Then, from (7),

\[ |I_k(aX) \cap I_l(bY)| = |I_k(X) \cap I_l(bY)| + |I_{k-1}(aX) \cap I_l(Y)| + (q-2)|I_{k-1}(aX) \cap I_{l-1}(bY)|. \]

(11)

Again, we bound the terms in (11) with formulas of the type \( N_q^+ (n, t, k, \ell) \). In this case, the argument matching is slightly more complicated. For the first term in (11), \( bY \) has length \( n \) while the insertion arguments are clearly \( t - 1 \) and \( k \). It remains to find the distance argument \( \ell' \). Recall that \( d_e(aX, bY) = t - k + 2\ell \). We have, from Claim 2

\[ d_e(aX, bY) \in \{ t - k + 2\ell - 1, t - k + 2\ell + 1 \} \]

and

\[ d_e(aX, Y) \in \{ t - k + 2\ell - 1, t - k + 2\ell + 1 \}. \]

We must have \( d_e(X, bY) = (t - k - 2\ell') \), so that \( \ell' = \frac{1}{2} (d_e(X, bY) + k - (t-1)) \in \{ \ell, \ell + 1 \} \). Thus, the possible argument tuples for the first term are \( \{ (n, t-1, k, \ell), (n, t-1, k, \ell + 1) \} \). Next, for the second term in (11), \( Y \) has length \( n \) and the insertion arguments are \( t \) and \( k - 1 \). The distance argument \( \ell' \) satisfies \( d_e(aX, Y) = t - (k-1) + 2\ell' \in \{ t - k + 2\ell - 1, t - k + 2\ell + 1 \} \) by Claim 2.

Thus, the possible argument tuples for this term are \( \{ (n-1, t, k-1, \ell - 1), (n-1, t, k-1, \ell) \} \). Finally, the arguments for the last term must be \( (n, t-1, k-1, \ell) \). Applying Claim 1, we have that

\[ |I_k(X) \cap I_{l-1}(bY)| \leq \max\{ N_q^+ (n-1, t-1, k, \ell), N_q^+ (n, t-1, k-1, \ell) \} = N_q^+ (n-1, t-1, k-1, \ell), \]

and, finally,

\[ |I_{k-1}(aX) \cap I_l(Y)| \leq \max\{ N_q^+ (n-1, t-1, k-1, \ell - 1), N_q^+ (n-1, t-1, k-1, \ell) \} = N_q^+ (n-1, t-1, k-1, \ell - 1), \]

and we are done.

We proceed with the proof of Theorem 5.

**Proof:** We first use induction on \( n + t + k \) to show that

\[ N_q^+ (n, t, k, \ell) \leq N_q^+ (n, t, k, \ell). \]

(12)

The base cases are \( n + t + k \in \{ 1, 2 \} \). First we consider \( n \in \{ 1, 2 \} \) and \( t = k = \ell = 0 \). Since \( I_0(X') \cap I_0(Y') \subseteq \{ X' \} \), we have that \( |I_0(X') \cap I_0(Y')| \leq 1 \). The right hand side of (12) evaluates to \( \left( \binom{0}{0} \right) \left( \binom{n}{0} \right)^2 = 1 \), as we wished. The other possibility is \( n = 1, t = 1, k = 0, \ell = 0 \). We have that \( I_0(X') \cap I_1(Y') \subseteq \{ X' \} \), and indeed, \( N_q^+ (1, 1, 0, 0) = \left( \binom{1}{1} \right)^2 = 1 \).

Assume that the claim in (12) holds for all \( n + t + k < m \). We prove that it is true for \( n + t + k = m \). Take sequences \( X', Y' \), where \( X' \neq Y' \), \( X' \in \mathbb{F}_q^{n+t-k}, Y' \in \mathbb{F}_q^n \) and \( d_e(X', Y') = t - k + 2\ell \), and \( n + t + k = m \). Write \( X' = aX \) and \( Y' = bY \), with \( a, b \in \mathbb{F}_q \).

We examine the first symbol. The first case is that \( X' \) and \( Y' \) start with the same symbol, so that \( a = b \). Then, using Lemma 6, the induction hypothesis, and the first recursion in Lemma 7, we write

\[ |I_k(aX) \cap I_l(bY)| \leq N_q^+ (n-1, t, k, \ell) + (q-1)N_q^+ (n, t-1, k-1, \ell) \]

and

\[ N_q^+ (n-1, t, k, \ell) + (q-1)N_q^+ (n, t-1, k-1, \ell) \]

= \[ N_q^+ (n-1, t, k, \ell). \]

The other case is \( a \neq b \), so that \( X' \) and \( Y' \) start with different symbols. Now we apply the second result in Lemma 7.
induction hypothesis, and the second recursion in Lemma 6 yielding
\[ |I_k(aX) \cap I_l(bY)| \]
\[ \leq N_q^+(n, t-1, k, \ell) + N_q^+(n-1, t, k-1, \ell - 1) + (q-2)N_q^+(n, t-1, k-1, \ell) \]
\[ \leq N_q^+(n, t-1, k, \ell) + N_q^+(n-1, t, k-1, \ell - 1) + (q-2)N_q^+(n, t-1, k-1, \ell) \]
\[ = N_q^+(n, t, k, \ell). \]

Thus, \( N_q^+(n, t, k, \ell) \leq N_q^+(n, t, k, \ell). \) Now we show that there exist sequences \( X', Y' \) that attain the maximum; that is, we find \( X', Y' \) with \( |I_k(X') \cap I_l(Y')| = N_q^+(n, t, k, \ell). \) This allows us to conclude that \( N_q^+(n, t, k, \ell) = N_q^+(n, t, k, \ell), \) completing the proof.

We select the sequences
\[ X' = 00\ldots0 \quad \text{and} \quad Y' = 11\ldots100\ldots0 \]
Note of course that we could have selected any sequences with the structure \( X' = \overline{a}\overline{a}\ldots\overline{a} \) and \( Y' = \overline{b}\overline{b}\ldots\overline{b} \) for \( a \neq b \) and \( a, b \in \mathbb{F}_q. \)

It is hard to give a direct proof that \( |I_k(X') \cap I_l(Y')| = N_q^+(n, t, k, \ell); \) instead, we proceed inductively. As we will see, the induction is identical to the previous proof, replacing inequalities with equalities. As before, the induction is on \( n + t + k. \) The base cases are \( n + t + k \in \{1, 2\}. \) If \( n \in \{1, 2\} \) and \( t = k = \ell = 0, X' = 00 \) and \( Y' = 00 \) or \( X' = 0 \) and \( Y' = 0. \) Thus, \( |I_0(X') \cap I_0(Y')| = 1 = N_q^+(n, 0, 0, 0). \) as desired. If \( n = 1 \) and \( k = \ell = 0, X' = 00 \) and \( Y' = 00. \) Then \( |I_0(X') \cap I_1(Y')| = 1 = N_q^+(1, 1, 0, 0). \)

Assume that the induction hypothesis holds for \( n + t + k < m; \) we show it is true for \( n + t + k = m. \) If \( \ell \geq 1, \) we apply (7) to write
\[ |I_k(X') \cap I_l(Y')| = |I_k(X) \cap I_{l-1}(Y')| + |I_{k-1}(X') \cap I_l(Y')| + (q-2)|I_{k-1}(X') \cap I_{l-1}(Y')|, \]
where \( X = 00\ldots0 \) and \( Y = 11\ldots100\ldots0. \) Note that to produce \( \ell 1's \) and insert \( t + k + n + 1 - (n - \ell) = \ell + t + k + 1 \) 0's. Thus, \( d_{e}(X, Y') = t + k + 2\ell + 1. \) Now we again match arguments: in the first term in (13), \( Y' \) has length \( n, \) the insertion arguments are \( t - 1 \) and \( k \) and \( d_{e}(X, Y') = t + k + 2\ell + 1. \) Thus, the distance argument satisfies \( \ell' = \frac{1}{2}(d_{e}(X, Y') + k - (t - 1)) = \frac{1}{2}(t + k + 2\ell + 1 + k - (t - 1)) = \ell. \)
Applying the induction hypothesis, we may thus write \( |I_k(X) \cap I_{l-1}(Y')| = N_q^+(n, t, k, \ell) \) similarly, for the second term in (13), \( Y \) has length \( n - 1, \) the insertion arguments are \( t - 1 \) and \( k - 1 \) while \( d_{e}(X', Y) = t + k + 2\ell - 1. \) The distance argument satisfies \( \ell' = \frac{1}{2}(d_{e}(X', Y) + (k - 1) - t) = \frac{1}{2}(t + k + 2\ell - 1 + (k - 1) - t) = \ell. \) Apply the induction hypothesis to write \( |I_{k-1}(X') \cap I_l(Y)| = N_q^+(n-1, t, k-1, \ell - 1). \) For the

last term, the arguments are clearly \( (n, t - 1, k - 1, \ell). \) Using the induction hypothesis once more, \( |I_{k-1}(X') \cap I_{l-1}(Y')| = N_q^+(n, t - 1, k - 1, \ell). \) Now, using Lemma 6 we get that
\[ |I_k(X') \cap I_l(Y')| = N_q^+(n, t-1, k, \ell) + N_q^+(n-1, t, k-1, \ell - 1) + (q-2)N_q^+(n, t-1, k-1, \ell) = N_q^+(n, t, k, \ell). \]

If \( \ell = 0, X' \) and \( Y' \) both start with 0 so we apply (9) to write
\[ |I_k(X') \cap I_l(Y')| = |I_k(X) \cap I_l(Y)| + (q-1)|I_{k-1}(X') \cap I_{l-1}(Y')|. \] In this case too we apply the induction hypothesis and Lemma 6 giving
\[ |I_k(X') \cap I_l(Y')| = N_q^+(n-1, t, k, \ell) + (q-1)N_q^+(n, t - 1, k, - 1) = N_q^+(n, t, k, \ell), \]
thus completing the proof. To retrieve the formula given by (9), take \( t = k \) in (8).

**B. Corollaries**

Specific cases of Theorem 5 yield a variety of interesting results. Before we present these results, we require two auxiliary binomial identities. The purpose of these identities is to help simplify the more complex formulas in (8) and (9) for special cases. The proofs are found in the appendix.

**Lemma 8.**

1. For \( m \geq 0, \)
\[ \sum_{j=0}^{m} \binom{2j}{j} \binom{m+j}{2j} (-1)^{m+j} = 1. \]

2. For \( n, m, t, j \geq 0 \) and \( t + j \geq m, \)
\[ \sum_{j=0}^{m} \binom{t+j-i}{t+j-m} \binom{n+t}{t} (-1)^{m-i} = \binom{n+m-j-1}{m}. \]

We are now ready to proceed with our corollaries. We begin by showing that it is possible to recover Levenshtein’s formula for \( N_q^+(n, t) = \max_{X \in \mathbb{F}_q^n} |I_k(X) \cap I_l(Y)| \) by taking \( \ell = 1 \) in (9). In other words, the maximum number of supersequences is attained by taking sequences at the smallest possible \( d_e = 2 \) non-zero distance.

**Corollary 9 (Levenshtein’s result for \( N_q^+(n, t, \ell = 1) \))**

For \( n \) a positive integer and \( t \) a non-negative integer,
\[ N_q^+(n, t, 1) = N_q^+(n, t), \]
\[ = \sum_{i=0}^{t-1} \binom{n+t}{i} (-1)^{(q-1)i} (-1)^{t-i}. \]

**Proof:** Note first that \( N_q^+(n, t, 1) = N_q^+(n, t) \) follows from Claim 1. From the first identity in Lemma 8, we have that \( \sum_{j=1}^{m} \binom{m+j}{2j} (-1)^{m+j} = 1 - (-1)^{m}. \) Taking \( m = t - i \) yields
\[ \sum_{j=1}^{t-i} \binom{2j}{j} (-1)^{t-i+j} = 1 - (-1)^{t-i}. \]
If we exchange the sums in $i$ and $j$, we can rewrite the $\ell = 1$ case in (9) as
\[
N_q^+ (n, t, 1) = \sum_{i=1}^{t-1} \sum_{j=1}^{i} \binom{2j}{j} \binom{n+t}{i} (q-1)^i (1 - (1)^{j+i-i}).
\]
Applying (15), we have that
\[
N_q^+ (n, t, 1) = \sum_{i=0}^{t-1} \binom{n+t}{i} (q-1)^i (1 - (1)^{i+j-i}),
\]
as desired. 

Note that we did not require the distance parameter $\ell$ to be positive in Theorem 5. In fact, $\ell = 0$ implies $d_t(X, Y) = 0$, or $X = Y$. In other words, all supersequences of $X$ are “common” supersequences (of $X$ and $X$), so we expect $N_q^+ (n, t, 0)$ to reduce to the formula for the number of supersequences $I_q(n, t)$. Happily, this is the case:

\textbf{Corollary 10 ($\ell = 0$ case)} For $n$ a positive integer and $t$ a non-negative integer,
\[
N_q^+ (n, t, 0) = I_q(n, t) = \sum_{i=0}^{t-1} \binom{n+t}{i} (q-1)^i.
\]
\textbf{Proof:} We exchange the sums for $i$ and $j$ in (9) with $\ell = 0$. This gives
\[
N_q^+ (n, t, 0) = \sum_{i=0}^{t-1} \sum_{j=0}^{i} \binom{2j}{j} \binom{n+t}{i} (q-1)^i (1 - (1)^{j+i-i}).
\]
Now set $m = t - i$ in the first part of Lemma 8 and apply the result to (17). We get
\[
N_q^+ (n, t, 0) = \sum_{i=0}^{t-1} \binom{n+t}{i} (q-1)^i,
\]
which is just $I_q(n, t)$. 

The case $\ell = 0$ represents the minimum distance criterion. We also consider the maximum criterion. Recall that in Theorem 5 we required that $k \geq \ell$. What happens if $\ell > k$? In this case, the number of common supersequences is always 0. If a common supersequence $Z$ existed for $X \in \mathbb{F}_q^{n+t-k}$ and $Y \in \mathbb{F}_q^n$ with $d_t(X, Y) = 2\ell + t - k$, we could then produce $Y$ from $X$ with $t + k$ insertions and deletions. However, since $\ell > k$, $t + k < 2\ell + t - k$, a contradiction. This is especially easy to see for equal-length sequences ($t = k$).

The maximum distance with a non-zero number of common supersequences is for $\ell = k$. In that case, the formula in Theorem 5 reduces to $\binom{t+k}{k}$. Here, we can even see a direct combinatorial interpretation of the formula. Consider for example $X = 00\ldots0$ and $Y = 11\ldots1$, where $X$ is made up of $t$ 0’s and $Y$ is made up of $k$ 1’s. Then, any common supersequence (formed by $t$ insertions into $Y$ and $k$ insertions into $X$) is a sequence with $t$ 0’s and $k$ 1’s. There are clearly $\binom{t+k}{k}$ such sequences.

Finally, we introduce a binary simplification of the result in Theorem 5.

\textbf{Corollary 11 (Binary Case)} Let $n$ be a positive integer and $t, k, \ell$ be non-negative integers such that $t \geq k \geq \ell$ and $n \geq \ell$. Then, we have the formula
\[
N_2^+ (n, t, k, \ell) = \sum_{j=0}^{k-j} \binom{t+j}{j} \binom{n+k-(2j+1)}{k-j}.
\]
\textbf{Proof:} We take $m = k-j$ in the second identity of Lemma 8 giving
\[
\sum_{j=0}^{k-j} \binom{t+j}{j} \binom{n+k-(2j+1)}{k-j} = \binom{n+k-2j+1}{k-j}.
\]
Now, applying this result, we have
\[
N_2^+ (n, t, k, \ell) = \sum_{j=0}^{k-j} \binom{t+j}{j} \binom{n+k-(2j+1)}{k-j} = \binom{n+k-2j+1}{k-j},
\]
as desired. 

\section{IV. Reconstruction in Varshamov-Tenengolts (VT) Codes}

Thus far, we have examined the number of traces required to distinguish between (sufficiently distant) sequences. We answered this question by deriving the $N_q^+ (n, t, k, \ell)$ formula. However, this expression represents the worst case. We may wonder whether in some particular code the situation is better. That is, we may ask whether the codewords of a code require fewer than $N_q^+ (n, t, k, \ell) + 1$ traces for reconstruction.

One major challenge when dealing with such a question is that there are not very many explicit codes correcting a fixed number of insertions and deletions. We will examine the most famous such code, the binary single insertion/deletion-correcting Varshamov-Tenengolts (VT) code [28, 29]:
\[
C_{VT}(n, a) := \left\{ X = (x_1, \ldots, x_n) \in \mathbb{F}_2^n : \sum_{i=1}^{n} i \cdot x_i \equiv a \mod n + 1 \right\}.
\]

We may take any $0 \leq a \leq n$ to form a different code. The VT codes partition the space $\{0, 1\}^n$. Since the VT codes correct a single insertion or deletion and have equal-length codewords, for any $E, F \in C_{VT}(n, a)$ ($E \neq F$), we have that $d_0(E, F) \geq 4$. This represents the case of $\ell = 2$ in our notation. Thus we seek to find out whether there exist (unordered) pairs of codewords $\{E, F\}$ with $E, F \in C_{VT}(n, a)$ for some particular $a$ such that $|I_{\ell}(E) \cap I_{\ell}(F)| = N_2^+ (n, t, 2)$. If such pairs exist for each $a$, we can conclude that the VT codes require the worst-case $N_2^+ (n, t, 2) + 1$ traces for reconstruction. We are motivated to seek codes with similar properties to the VT codes but with fewer traces needed for reconstruction.
In this section, we prove an even stronger result. Not only is there always a pair of codewords achieving the worst case, but there are exponentially many such pairs for each VT code. In particular, we prove that for any $n \geq 7$ and any $a$ with $0 \leq a \leq n$, there exists a set $S_a$ of unordered pairs of elements, where for any element $\{E, F\} \in S_a$, we have $E, F \in C_{VT}(n, a)$, $E \neq F$, $N_2^+(n, t, 2) = |I_1(E) \cap I_2(F)|$ and $|S_a| \geq 2^n - \lceil \log_2(n) \rceil - 3$.

We first establish some simple claims. We make use of the following notation. Let $V = (v_1, \ldots, v_{n-2}) \in \mathbb{F}_2^{n-2}$ be a sequence of length $n - 2$. We write

$$X(n, p, V) := (v_1, \ldots, v_{p-1}, 1, v_p, v_{p+1}, \ldots, v_{n-2}) \in \mathbb{F}_2^n,$$

i.e., $X(n, p, V)$ is a sequence whose components are equal to $V$ in the first $p - 1$ positions, followed by 1,1, then by the remaining $n - p - 1$ bits in $V$. Similarly, let

$$Z(n, p, V) = (v_1, \ldots, v_{p-1}, 0, 0, v_p, \ldots, v_{n-2}) \in \mathbb{F}_2^n,$$

$Y(n, p, V) = (v_1, \ldots, v_{p-1}, 0, v_p, 0, v_{p+1}, \ldots, v_n) \in \mathbb{F}_2^n,$

and

$$W(n, p, V) = (v_1, \ldots, v_{p-1}, 1, v_p, 1, v_{p+1}, \ldots, v_n) \in \mathbb{F}_2^n.$$

Before continuing we provide a small example that illustrates the main ideas.

**Example 1.** We take $V = 100100 \in \mathbb{F}_2^6$. Then, we have that $X(8, 4, V) = 10011100$, $Z(8, 4, V) = 10001001$. Note that $X(8, 4, V)$ and $Z(8, 4, V)$ are in the same VT code, $C_{VT}(8, 7)$, since $\sum_{i=1}^8 i x_i = 1 + 4 + 5 + 6 \equiv 7 \mod 9$ and $\sum_{i=1}^8 i z_i = 1 + 6 \equiv 7 \mod 9$.

Note that we always have that

$$d_c(X(n, p, V), Z(n, p, V)) = 4,$$

since the two sequences are the same except for the 0,0 and 1,1 bits in the middle. Similarly, we have that

$$d_c(Y(n, p, V), W(n, p, V)) = 4.$$

Next we show that if we let $p$ be the central position in some $V$, then the resulting sequences $X$ and $Z$ satisfy $|I_1(X) \cap I_2(Z)| = N_2^+(n, t, \ell = 2)$ and similarly $|I_1(Y) \cap I_2(W)| = N_2^+(n, t, \ell = 2)$.

**Lemma 12.** For any $n \geq 4$, $t \geq 2$, and $V \in \mathbb{F}_2^{n-2}$,

$$N_2^+(n, t, 2) = |I_1(X(n, \lceil \frac{n}{2} \rceil, V)) \cap I_2(Z(n, \lfloor \frac{n}{2} \rfloor, V))| = |I_1(Y(n, \lceil \frac{n}{2} \rceil, V)) \cap I_2(W(n, \lfloor \frac{n}{2} \rfloor, V))|.$$

**Proof:** We show that $N_2^+(n, t, 2) = |I_1(X(n, \lfloor \frac{n}{2} \rfloor, V)) \cap I_2(Z(n, \lceil \frac{n}{2} \rceil, V))|$. Just as in our earlier proofs, we use induction; this time on $n + t$. Since $n \geq 4$ and $t \geq 2$, we have $n + t = 6$ as our base case. It can be verified exhaustively that for any $V \in \mathbb{F}_2^3$, $N_2^+(4, 2, 2) = |I_2(X(4, 2, V)) \cap I_2(Z(4, 2, V))| = \sum_{j=1}^{t} (\binom{n}{j} (n-t-2) + 1) = 10^5 (2-5) = 6$, as desired.

Suppose that $N_2^+(n, t, 2) = |I_1(X(n, \lceil \frac{n}{2} \rceil, V)) \cap I_2(Z(n, \lfloor \frac{n}{2} \rfloor, V))|$ for all $n + t < m$ and consider the case where $n + t = m$. Suppose that $n$ is even. The case where $n$ is odd can be proven using similar arguments. Let $V' \in \mathbb{F}_2^{n-3}$ be the sequence obtained by deleting the first bit from $V$. Since $X(n, \lceil \frac{n}{2} \rceil, V)$ and $Z(n, \lfloor \frac{n}{2} \rfloor, V')$ start with the same bit, we can use (5) in Claim (5) to write

$$|I_1(X(n, \lceil \frac{n}{2} \rceil, V')) \cap I_2(Z(n, \lfloor \frac{n}{2} \rfloor, V))| = |I_1(X(n-1, \lceil \frac{n-1}{2} \rceil, V')) \cap I_2(Z(n-1, \lfloor \frac{n-1}{2} \rfloor, V'))| + |I_1-1(X(n, \lceil \frac{n}{2} \rceil, V')) \cap I_2(Z(n, \lfloor \frac{n}{2} \rfloor, V))|.$$
sequence is in a particular VT code. The remaining positions, however, form a basis, (that is, a linear combination of them produces any \( a \) with \( 0 \leq a \leq n \) modulo \( n + 1 \)) so that we can control their output in the checksum to fix the \( a \) VT code parameter.

The algorithm enables us to find the following result. We defer the proof to the appendix.

**Lemma 13.** For any integer \( n \geq 7 \), and integer \( 0 \leq m \leq n \), there exists a subset \( T' \subseteq FP(n) \) where \( \sum_{i \in T'} i = m \). In addition, \( \lfloor \log_2(n) \rfloor \leq |FP(n)| \leq \lfloor \log_2(n) \rfloor + 1 \). Furthermore, if \( n \) is even, we have \( \{ \lfloor \frac{n}{2} \rfloor, \lfloor \frac{n}{2} \rfloor + 1 \} \not\subseteq FP(n) \) and if \( n \) is odd, then \( \{ \lfloor \frac{n}{2} \rfloor, \lfloor \frac{n}{2} \rfloor + 2 \} \not\subseteq FP(n) \).

The idea in Lemma 13 is that a linear combination of at most \( \lfloor \log_2(n) \rfloor \) numbers in \([n]\) can generate any \( m \) for \( 0 \leq m \leq n \). These numbers are precisely those returned by the function \( FP(n) \). Therefore, we can force a sequence of length \( n \) to be in any particular VT code by controlling the sequence components at these positions. The subset \( T' \) represents those positions where we will place a 1 in the codeword, while the positions given by \( FP(n) \setminus T' \) will be set to 0. The idea is explained in further detail in the following proof of the main result of this section.

**Theorem 14.** For any \( n \geq 7 \) and \( a \in \mathbb{Z}_{n+1} \), there exists a set of unordered pairs \( S_a \) where for any pair \( \{ E, F \} \in S_a \), we have \( E, F \in CVT(n, a) \), \( E \neq F \), \( N_2^+ = |I_t(E) \cap I_t(F)| \), and \( \sum_{i \in S_a} i = 2^{n-\lfloor \log_2(n) \rfloor} - 3 \).

**Proof:** We start by counting the number of different ways to form a pair \( \{ E, F \} \in S_a \). First, consider the case where \( n \) is even. Let \( U = FP(n) \cup \left\{ \frac{n}{2}, \frac{n}{2} + 1 \right\} \) be a set of positions in our codewords. We can select the remaining positions, that is, \( [n] \setminus U \), freely and still form a codeword. Therefore, we have \( 2^n - |FP(n)| - 3 \) choices for these positions. Next, let us say that \( \sum_{k \in [n]} k e_k \equiv c \mod n + 1 \). Then, using Lemma 13, we fix the components of \( E \) at positions in the set \( FP(n) \) so that \( \sum_{k \in FP(n)} k e_k \equiv a - c \mod n + 1 \). Finally, let \( (e_1, e_{\frac{n}{2}}) = (0, 0) \). Thus, we may write \( E = Z(n, \frac{n}{2}, E') \) for some vector \( E' \in \mathbb{F}_2^{\frac{n}{2}} \). Notice that \( E \in CVT(n, a) \) since \( \sum_{k \in [n]} k e_k = \sum_{k \in [n]} k e_k + \sum_{k \in FP(n)} k e_k + \frac{n}{2} e_{\frac{n}{2}} + \left( \frac{n}{2} + 1 \right) e_{\frac{n}{2} + 1} \equiv a \mod (n + 1) \).

Let \( F = (e_1, \ldots, e_{\frac{n}{2} - 1}, 1, e_{\frac{n}{2} + 1}, \ldots, e_n) = (f_1, \ldots, f_n) \), so that \( F = X(n, \frac{n}{2}, E) \). Then, \( F \in CVT(n, a) \), since \( \sum_{k \in [n]} k f_k = \sum_{k \in [n]} k e_k + \frac{n}{2} + \left( \frac{n}{2} + 1 \right) \equiv a \mod (n + 1) \). Thus, from Lemma 12, \( |I_t(E) \cap I_t(F)| = N_2^+(n, t, 2) \). Thus, \( |S_a| \geq 2^n - |FP(n)| - 2 \geq 2^n - \lfloor \log_2(n) \rfloor - 3 \).

Next we examine the case of \( n \) odd. We proceed in the same manner as before except that we first select the values of \( E \) except in positions from the set \( U = FP(n) \cup \left\{ \frac{n}{2}, \frac{n}{2} + 1 \right\} \). Afterwards we assign values to components in \( E \) whose indices belong to set \( FP(n) \) in such a way that \( E \in CVT(n, a) \). Finally, \( e_{\lfloor \frac{n}{2} \rfloor + 1} = 2 \) are set to zero so that \( E = Y(n, \lfloor \frac{n}{2} \rfloor, E') \). Next, \( F = (f_1, \ldots, f_n) \) is set to equal \( E \) except that \( f_{\lfloor \frac{n}{2} \rfloor} = f_{\lfloor \frac{n}{2} \rfloor + 2} = 1 \). Thus \( F = W(n, \lfloor \frac{n}{2} \rfloor, E') \). Using the same arguments as in the previous paragraph, it can be shown that \( E, F \in CVT(n, a) \). Furthermore, from Lemma 12, \( |I_t(E) \cap I_t(F)| = N_2^+(n, t, 2) \) and thus \( |S_a| \geq 2^n - \lfloor \log_2(n) \rfloor - 3 \).

Next, we consider other channels and applications of our main results.

**V. OTHER CHANNELS AND APPLICATIONS**

Thus far we have only been concerned with insertion channels. It is reasonable to ask what occurs in the cases of deletion or mixed insertion/deletion channels. It is not too surprising that finding expressions for similar problems for these channels is much harder: the deletion channel is much less symmetric compared to the insertion channel, and the insertion/deletion channel deals with the challenges of both. We also apply the results (derived for an adversarial channel) to a probabilistic insertion channel.

**A. DELETION CHANNEL**

Levenshtein examined exact reconstruction for deletion channels in [1]. He defined \( N_q^- (n, t) := \max_{X, Z \in \mathbb{F}_2^n, X \neq Z} |D_t(X) \cap D_t(Z)| \) and showed that

\[
N_q^- (n, t) = \left( \sum_{i=1}^{a-1} D_q(n - i - 1, t - i) \right) + D_q(n - 2, t - 1).
\]

Here, \( D_q(n, t) \) is the maximum size of the deletion ball \( D_t(X) \) for some \( X \in \mathbb{F}_2^n \). It is known that \( D_q(n, t) \) satisfies the recursion \( D_q(n, t) = \sum_{i=0}^{t-1} D_q(n-i, t-i) \), where \( D_q(n, t) = 1 \) if \( n \geq t \geq 0 \) and \( D_q(n, t) = 0 \) otherwise [30]. It is not hard to see that \( D_2(n, t) = \sum_{i=0}^{t-1} n^{-i} \). This enables us to write that the maximum number of common subsequences in the binary case is given by

\[
N_q^- (n, t) = 2 \sum_{i=0}^{t-1} \left( n - t - 1 \right).
\]

Just as before, we may ask what happens to the number of sequences required for reconstruction if we select the original sequences from an insertion/deletion-correcting code. We can analogously define

\[
N_q^- (n, t, \ell) = \max_{X, Z \in \mathbb{F}_2^n, d_t(X, Z) \geq 2\ell} |D_t(X) \cap D_t(Z)|.
\]

Few results are known regarding \( N_q^- (n, t, \ell) \). The work [31] is dedicated to the \( \ell = 2, q = 2 \) case (corresponding to VT codes). The authors showed that for \( t \leq n/2 \),

\[
N_q^- (n, t, 2) = 2D_2(n - 4, t - 2) + 2D_2(n - 5, t - 2) + 2D_2(n - 7, t - 2) + 2D_2(n - 6, t - 3) + 2D_2(n - 7, t - 3).
\]

Our contribution consists of removing the reliance on recursions from this formula, yielding the exact expression

\[
N_q^- (n, t, 2) = 2D_2(n - 2, t - 1) - 2 \left( \frac{n - t - 3}{t - 1} \right) - \left( \frac{n - t - 4}{t - 3} \right).
\]
Let us slightly abuse notation as follows: given a set $S \subseteq \mathbb{F}_2^n$, we write $I_t(S)$ and $D_t(S)$ for $\cup_{X \in S} I_t(X)$ and $\cup_{X \in S} D_t(X)$, respectively. Then, the $t$-insertion/$t$-deletion ball centered at $X$ may be written $B_t(X) := I_t(D_t(X))$.

Since the general version of the problem seems intractable, in this subsection we focus on providing a lower bound on the number of distinct distorted sequences (resulting from an insertion/deletion channel) required to reconstruct a binary sequence $X$. Specifically, we are interested in a lower bound on $N_H(\mathbb{F}_2^n, 2t)$, defined as in \cite{2}, where $H$ is the set of single symbol insertions and deletions. Note that here, we specifically require the $2t$ argument to imply $t$ insertions and $t$ deletions. We can write, in general, that

$$N_H(\mathbb{F}_2^n, 2t) = \max_{X, Z \in \mathbb{F}_2^n; X \neq Z} |I_t(D_t(X)) \cap I_t(D_t(Z))|.$$  

We provide a lower bound on $N_H(\mathbb{F}_2^n, 2t)$ by computing the number of common distorted sequences in one particular (and non-trivial) case. This is the case of the so-called binary circular string $C_n = 0101\ldots$ (or $1010\ldots$). This string is particularly interesting; in \cite{30} it is shown that $C_n = \arg \max_{X \in \mathbb{F}_2^n} |D_t(X)|$.

We begin by evaluating the size of the ball centered at $C_n$, $B_t(C_n) = I_t(D_t(C_n))$.

**Theorem 15.** The size of the $t$-ball around the binary circular string $C_n \in \mathbb{F}_2^n$ is given by

$$|B_t(C_n)| = |I_t(D_t(C_n))| = \sum_{i=0}^{2t} \binom{n}{i}.$$  

Before we proceed with the proof of Theorem 15, we comment on this result. Since $C_n$ is known to maximize the deletion ball size $|D_t(X)|$, we may ask whether the string $C_n$ also maximizes the insertion/deletion ball size. Surprisingly, this is not the case. Although $|B_t(C_n)|$ is quite large and in certain cases is in fact maximal, the string $X = 00110011\ldots$ generally yields a larger ball size. More details on which strings maximize such ball sizes can be found in \cite{32}.

We use the following lemma in the proof of Theorem 15:

Lemma 16. Let $n, t$ be positive integers with $n \geq 2t$. Let $C_{n-2t}$ be the substring formed by the first $n-2t$ bits of the circular string $C_n$. Then, the $t$-deletion ball centered at $C_n$ is exactly the $t$-insertion ball centered at $C_{n-2t}$. That is,

$$D_t(C_n) = I_t(C_{n-2t}).$$  

**Proof:** First, observe that $C_{n-2t}$ begins and ends with the same bit as $C_n$. We will show the result by induction on $n$.

The base case is $n = 2t$. Here, $C_{n-2t}$ is the empty string, and the right hand side in (19) is just $\mathbb{F}_2$, the set of all binary sequences of length $t$. It is easy to see that this set is equal to $D_t(C_n) = D_t(C_{2t}) = D_t(01\ldots01)$ (or $D_t(10\ldots10)$). We may delete either the 0 or the 1 in all of the $t$ consecutive 0 (or 10) pairs in order to produce any sequence of length $t$. This establishes the base case.

Now, we assume that $D_{t'}(C_m) = I_{t'}(C_{m-2t})$ for all $t' \leq t$ and $m$ satisfying $2t \leq m \leq n$. The cases of $t' < t$ follow from the $t$ case by deleting and inserting identical elements. We examine $D_t(C_{n+1})$ with the goal of showing that it is identical to $I_t(C_{n+1-2t})$. We take the last bit of $C_{n+1}$ (and thus, of $C_{n+1-2t}$) to be 1, without loss of generality. Consider some $X \in D_t(C_{n+1})$ so that $X$ ends in exactly $k$ consecutive 0s, with $0 \leq k \leq t$. We show that $X \in I_t(C_{n+1-2t})$.

If $k = 0$, $X$ ends in 1, like $C_{n+1}$. In this case, $X = Y1$ where $Y$ has length $n-t$. Then, $Y$ may be produced by $t$ deletions in the string $C_n$, which is itself the first $n$ bits of $C_{n+1}$. Thus, $Y \in D_t(C_n)$. By the induction hypothesis, $D_t(C_{n+1}) = I_t(C_{n+1-2t})$, so $Y \in I_t(C_{n+1-2t})$. Then, $Y$ can be produced by $t$ insertions to $C_{n-2t}$, so, since $C_{n-2t}$ ends in 1, indeed $X \in I_t(C_{n+1-2t})$.

If $0 < k \leq t$, $X$ ends with the substring $100\ldots0$. In fact, we may write $X = Y00\ldots0$ for some string $Y$ of length $(n+1-t-k)$, $X$ results from the deletion of the last $k$ 1’s from $C_{n+1}$, and the deletion of an additional $t-k$ elements from the first $n+1-2k$ bits of $C_{n+1}$, which themselves form $C_{n+1-2k}$. That is, $Y \in D_{t-k}(C_{n+1-2k})$. Applying the induction hypothesis, $Y$ is in the set $I_{t-k}(C_{n+1-2k}) = I_t(C_{n+1-2t})$. Then, clearly $X \in I_t(C_{n+1-2t})$, as we may use the remaining $k$ insertions to add $k$ 0s to the end of $Y$ to produce $X$. We conclude that $D_t(C_{n+1}) \subseteq I_t(C_{n+1-2t})$.

The other direction is essentially identical. Take $Z \in I_t(C_{n+1-2t})$. If $Z$ ends in 1, then $Z = Y1$ where $Y$ may be formed by $t$ insertions into $C_{n-2t}$. By the induction hypothesis, $Y \in D_t(C_{n+1})$, and since $C_{n+1}$ ends in 1, we have that $Z \in D_t(C_{n+1})$. If $Z$ ends in exactly $k$ 0s, $(1 \leq k \leq t)$ then $Z = Y00\ldots0$ for some $Y$ of length $n+1-t-k$.

Then, $Y$ can be formed by $t-k$ insertions into $C_{n+1-2t}$. By the induction hypothesis, $Y \in D_{t-k}(C_{n+1-2t}) = D_{t-k}(C_{n+1-2k})$. Then, $Z \in D_t(C_{n+1})$, since we may use the remaining $k$ deletions to delete the last $k$ 1s in $C_{n+1}$. With this, $I_t(C_{n-2t}) \subseteq D_t(C_{n+1})$.

Thus, $D_t(C_{n+1}) = I_t(C_{n-2t+1})$, and we are done.

**Theorem 15** follows almost immediately from Lemma 16.
Proof: Let \( n \geq 2t \). According to Lemma 16, \( D_t(C_n) = I_t(C_{n-2t}) \). Then, we have that

\[
|B_t(C_n)| = |\bigcup_{X \in D_t(C_n)} I_t(X)| = |\bigcup_{X \in D_t(C_{n-2t})} I_t(X)| = |I_{2t}(C_{n-2t})| = \sum_{i=0}^{2t} \binom{n}{i},
\]

where in the last step, we used the formula (4) for the number of supersequences formed by \( 2t \) insertions. The remaining cases for \( n < 2t \) are identical to the base case \( n = 2t \) in the proof of Lemma 16. Here too, \( D_t(C_n) = F_2^{n-t} \), so that \( \bigcup_{X \in D_t(C_n)} I_t(X) = \bigcup_{X \in F_2^{n-t}} I_t(X) \), implying that \( |B_t(C_n)| = 2^n \).

Theorem 15 is interesting, as in general it is very difficult to compute the exact ball size \( |B_t(X)| \) for any non-trivial \( X \) (such as any sequence that is not made up of all 0’s or all 1’s) or \( t > 1 \). The underlying symmetries for the circular string enable us to give this exact expression. We remark that Lemma 16 also yields an alternative way to compute the size of \( D_t(C_n) \) [30].

Now we return to the problem of common distorted sequences. Recall that we are interested in computing \( |I_t(D_t(X)) \cap I_t(D_t(Z))| \) for at least some non-trivial \( X, Z \in F_2^n \). Let us take \( X = C_n = 101010 \ldots \) and \( Z = C'_n = 010101 \ldots \). Note that \( d_t(C_n, C'_n) = 2 \), since we need only take the leading 1 in \( C_n \) and move it to the end to reproduce \( Z \). Now, we have that

\[
|I_t(D_t(C_n)) \cap I_t(D_t(C'_n))| = |I_t(I_t(C_{n-2t})) \cap I_t(I_t(C'_{n-2t}))| = |I_{2t}(C_{n-2t}) \cap I_{2t}(C'_{n-2t})| = N_t^2 \binom{n-2t}{2t,1} = 2^{2t-1} \sum_{i=0}^{2t-1} \binom{n}{i}.
\]

The equality (rather than inequality) in the third step is easy to check. Therefore, we have our desired bound on \( N_{H}(F_2^n, 2t) \):

\[
N_{H}(F_2^n, 2t) \geq 2^{2t-1} \sum_{i=0}^{2t-1} \binom{n}{i}.
\]

The important idea here is to replace deletions in our insertion/deletion channel with insertions. This idea is often useful when computing sizes of insertion/deletion balls, since deletions are much more difficult to deal with. Note that we can use a similar idea to compute the number of common distorted sequences for some other cases. For example, if we let \( Z = C_n \), but take \( X = 00 \ldots 0 \), we have that \( I_t(D_t(0 \ldots 0)) = I_t(0 \ldots 0) = I_t(n-t, t) \), which yields \( |I_t(D_t(0 \ldots 0)) \cap I_t(D_t(C_n))| = |I_t(0 \ldots 0) \cap I_{2t}(C_{n-2t})| \leq N_t^+ \binom{n-2t}{2t,1} \). A number of other similar expressions can be computed.

Having examined the deletion and insertion/deletion channels, we also consider applying our results to non-adversarial channels.

C. Application to Probabilistic Channel

Thus far, we have entirely focused on adversarial channels. However, there are many practical communication scenarios that can be modeled with probabilistic synchronization channels (e.g., underwater communication [33] and file synchronization [34, 35]). Our results can be applied to such channels as well. We focus on the standard probabilistic insertion channel shown in Fig. 3. Let \( x \) be a sequence in \( \{0,1,\ldots,q-1\}^n \) being transmitted through the channel. The insertion process is modeled as a Markov chain. For each symbol \( x_j \) for \( 1 \leq j \leq n \), there is an associated state which we call \( s_j \). The initial state is \( s_1 \). At state \( s_j \), \( 1 \leq j \leq n \), there is an option to insert a symbol in \( \{0,1,\ldots,q-1\} \) and return to state \( s_j \), with some probability \( p \leq 1 \), or to transmit symbol \( x_j \), with probability \( 1-p \). After transmitting \( x_j \) (from the original sequence \( x \) ), the process continues to state \( s_{j+1} \) if \( j < n \) or terminates if \( j = n \).

Let \( C \subseteq \{0,1,\ldots,q-1\}^n \) be a code with minimum edit distance \( 2\ell \) and length \( n \) and suppose \( x \in C \). In the probabilistic scenario, we repeatedly transmit \( x \) through the channel and accumulate outputs until we can decode/reconstruct \( x \). The outputs from repeated transmissions of \( x \) will be of varying lengths. Any channel output of length in \( \{n,n+1,\ldots,n+\ell-1\} \) is decodable (with code \( C \)’s decoder). If \( N_q(n,v,\ell) \) outputs of length \( n+v \) (where \( v \geq \ell \)) are received by repeated transmissions before an output with length in \( \{n,n+1,\ldots,n+\ell-1\} \) is received, we will uniquely reconstruct (rather than decode) \( x \). This scenario will occur if the probability of insertion \( p \) is sufficiently high. The forthcoming analysis will show such cases where reconstruction occurs before decoding. This idea shows the applicability of our result to probabilistic channels.

We let \( p_i \) be the probability that a particular channel output has been affected by exactly \( i \) insertions for \( i \geq 0 \). Let \( T_i \) be a threshold that counts the required number of length \( (n+i) \) outputs for decoding/reconstruction of \( x \). We seek to answer the following question: what is the average required number of transmissions to collect \( T_i \) channel outputs of length \( (n+i) \)? In particular, for what value of \( i \) is this quantity minimized?

First we compute a formula for the \( T_i \)’s. The code \( C \) has minimum distance \( 2\ell \), so that it can correct up to \( \ell-1 \) insertions. As a result, a single output of length \( n,n+1,\ldots,n+\ell-1 \) is sufficient to decode. If the outputs have length \( n+v \) for \( v \geq \ell \), we need \( N_q(n,v,\ell) + 1 \) outputs for reconstruction. Thus,

\[
T_i = \begin{cases} 1, & \text{for } 0 \leq i \leq \ell-1, \\ N_q(n,v,\ell) + 1, & \text{for } \ell \leq i. \end{cases}
\]
Next we can compute $p_i$ as follows. There are $i$ insertions. Since the channel allows for multiple insertions to take place at each state, we select $r$ ($1 \leq r \leq n$) states for the $i$ insertions. The total number of ways to distribute $i$ objects into $r$ buckets with at least one object per bucket is $\binom{n}{r-1}$. There are $\binom{n}{r}(i-1)$ ways to perform the insertions. Since there are $i$ insertions, we have a factor of $p^i$. From the channel procedure, to move on from each state, we need a factor of $(1-p)$, for a total of $(1-p)^n$ for $n$ states. Putting it all together, $p_0 = (1-p)^n$, while for $i \geq 1$,

$$p_i = \frac{\min(n, i)}{\sum_{r=1}^{\min(n, i)} \binom{n}{r} (i-1) (r-1) p^i (1-p)^n}.$$

The number of transmissions required to collect $T_i$ channel outputs of length $(n+i)$ follows a negative binomial distribution. For simplicity, we ignore the possibility of identical channel outputs of the same length. The mean number of trials to generate $f$ successes is given by $\frac{f}{p}$. Thus, we have that the expected number of transmissions is

$$E_i = \frac{T_i}{p_i},$$

or

$$E_i = \begin{cases} \frac{\min(n, i)}{\sum_{r=1}^{\min(n, i)} \binom{n}{r} (i-1) (r-1) p^i (1-p)^n}, & \text{for } 0 \leq i \leq \ell - 1, \\ \frac{\min(n, i)}{\sum_{r=1}^{\min(n, i)} \binom{n}{r} (i-1) (r-1) p^i (1-p)^n} + N_q^{i}(n, t, \ell) + 1, & \text{for } \ell \leq i. \end{cases}$$

It is possible to numerically compute $n + \arg \min_i E_i$, the expected length of the output sequence that will first allow us to decode/reconstruct. Indeed, it can be the case that reconstruction will be performed sooner than decoding. Consider, for example, $C$ a VT code of length $n = 500$ and distance $2\ell = 4$. Set $p = 0.3$. Then, $E_2$ transmissions are required on average for the $\binom{500}{2} + 1 = 7$ outputs of length $n + 2$ needed for reconstruction. We need $E_1$ or $E_0$ expected transmissions to decode. In this case, $E_1/E_2 = 10.74$, while $E_0/E_2 = 1.61 \times 10^3$. Thus, we need at least ten times fewer expected transmissions to reconstruct. In this case, $E_3/E_2 = 8.6$ (and the $E_i$ are increasing for $i \geq 3$), so on average, outputs of length $n + 2$ will be used for reconstruction. More generally, for $\ell$ small compared to $n$, if $p(n - \ell + 1) > \ell \times \binom{n}{2}$, $E_i$ is smaller than $E_i$ for $i < \ell$, and reconstruction will be performed prior to decoding.

VI. CONCLUSION

In this work, we examined the exact reconstruction of sequences that are codewords of synchronization (insertion/deletion-correcting) codes from traces that are the result of an insertion channel. We provided exact formulas for the number of traces necessary for the binary and non-binary cases of this problem. These formulas resolve a problem left open by Levenshtein, who derived the first expressions for the uncoded case. We also examined traces produced by other channels, such as the insertion and deletion channel, and explored the application of the results to a probabilistic insertion channel.

The expressions we found represent the worst-case number of traces needed when performing reconstruction in any code with the required minimum edit distance. We asked whether selecting a particular code allows us to reconstruct with fewer traces compared to the worst-case. We showed that for the popular single insertion/deletion-correcting Varshamov-Tenengolts codes, there are always many codeword pairs that require the worst-case number of traces for reconstruction. This inspires us to ask whether we can construct new codes that have similar properties to the VT codes, but better (smaller) requirements for reconstruction.

Our results can be viewed as a promising first step towards a more general theory for coded data reconstruction. This is a rich area with many interesting further questions. It would be interesting to allow the traces to be affected by other types of errors, such as deletions, substitutions, burst errors, transpositions, and others. It would be especially interesting to derive the maximal number of traces for the cases of deletion channels (for $\ell > 2$) and combined insertions/deletions/substitutions channels, which accurately model real-life data reconstruction scenarios. Equally intriguing is a study of efficient algorithms for reconstruction given the necessary number of traces: for example, given $N_2^{\ell}(n, t, k, \ell) + 1$ traces of $X$, what is the most efficient algorithm to reproduce $X$?

REFERENCES

[1] V. I. Levenshtein, “Efficient reconstruction of sequences,” IEEE Trans. Inf. Theory, vol. 47, no. 1, pp. 2-22, Jan. 2001.

[2] V. I. Levenshtein, “Efficient reconstruction of sequences from their subsequences or supersequences,” J. Combinatorial Theory, vol. 93, no. 2, pp. 310-332, Feb. 2001.

[3] T. Batu, S. Kannan, S. Khanna, and A. McGregor, “Reconstructing strings from random traces,” in Proc. ACM-SIAM Symp. Discrete Algorithms (SODA), New Orleans, LA, Jan. 2004, pp. 910-918.

[4] S. Kannan and A. McGregor, “More on reconstructing strings from random traces: insertions and deletions,” in Proc. IEEE Int. Symp. Inf. Theory (ISIT), Adelaide, Australia, Jun. 2005.

[5] K. Viswanathan and R. Swaminathan, “Improved string reconstruction over insertion-deletion channels,” in Proc. ACM-SIAM Symp. Discrete Algorithms (SODA), San Francisco, CA, Jan. 2008, pp. 399-408.

[6] T. Holenstein, M. Mitzenmacher, R. Panigrahy, and U. Wieder, “Trace reconstruction with constant deletion probability and related results,” in Proc. ACM-SIAM Symp. Discrete Algorithms (SODA), San Francisco, CA, Jan. 2008, pp. 389-398.

[7] E. Yarlobi and J. Bruck, “On the uncertainty of information retrieval in associative memories,” in Proc. IEEE Int. Symp. Inf. Theory (ISIT), Cambridge, MA, Jul. 2012, pp. 106-110.

[8] F. Sala, C. Schoeny, R. Gabrys, and L. Dolecek, “Three novel combinatorial theorems for the insertion/deletion channel,” in Proc. IEEE Int. Symp. Inf. Theory (ISIT), Hong Kong, China, Jun. 2015, pp. 2702-2706.

[9] F. Sala, R. Gabrys, C. Schoeny, K. Mazooji, and L. Dolecek, “Exact sequence reconstruction for insertion-correcting codes,” in Proc. IEEE Int. Symp. Inf. Theory (ISIT), Barcelona, Spain, Jul. 2016, pp. 615-619.

[10] V. Junnila and T. Laihonen, “Codes for information retrieval with small uncertainty,” IEEE Trans. Inf. Theory, vol. 60, no. 2, pp. 976-985, Feb. 2014.

[11] V. Junnila and T. Laihonen, “Information retrieval with varying number of input clues,” IEEE Trans. Inf. Theory, vol. 62, no. 2, pp. 625-638, Feb. 2016.

[12] F. Farnoud, M. Schwartz, and J. Bruck, “Estimating mutation rates and sequence age under a stochastic model for tandem duplication and point mutation,” available, https://dl.dropboxusercontent.com/u/2041685/website_docs/papers/2015-Estimating%20Mutation%20Rates%20and%20Sequence%20Age.pdf.
A. Proof of Lemma 6

We present a proof of Lemma 6 which we restate below:

**Lemma 6.** For $n \geq 1, q \geq 2$ and $t, k, \ell \geq 1$ with $t \geq k \geq \ell$, $N_q^+(n, t, k, \ell)$ satisfies the recursions

\[
N_q^+(n, t, k, \ell) = N_q^+(n - 1, t, k, \ell) + (q - 1)N_q^+(n, t - 1, k, \ell),
\]

and

\[
N_q^+(n, t, k, \ell) = N_q^+(n - 1, t - 1, k, \ell) + (q - 2)N_q^+(n, t - 1, k - 1, \ell).
\]

**Proof:** The proofs of these formulas only use standard sum manipulations and binomial identities. We first show the series of equalities and then describe the steps. For the first recursion, we have

\[
N_q^+(n - 1, t, k, \ell) + (q - 1)N_q^+(n, t - 1, k, \ell)
\]

\[
= \sum_{j=\ell}^{k-1} \binom{n-1}{j} \binom{t-1}{j} \binom{k-1}{j} (q-1)^j (-1)^{k-j-i}.
\]

**VII. APPENDIX**

[13] G. Benson and L. Dong, “Reconstructing the duplication history of a tandem repeat,” in *Proc. Int. Conf. Intell. Syst. Mol. Bio. (ISMB)*, Heidelberg, Germany, Aug. 1999, pp. 44-53.

[14] S. Jain and F. Farnoud and M. Schwartz, and J. Bruck, “Duplication-correcting codes for data storage in the DNA of living organisms,” in *Proc. IEEE Int. Symp. Inf. Theory (ISIT)*, Barcelona, Spain, Jul. 2016, pp. 1029-1032.

[15] F. Farnoud, M. Schwartz, and J. Bruck, “The capacity of string-duplication systems,” in *IEEE Trans. Inf. Theory*, vol. 62, no. 2, pp. 811-824, Feb. 2016.

[16] R. Gabrys, E. Yaakobi, and O. Milenkovic, “Codes in the Damerau distance for DNA storage,” in *Proc. IEEE Int. Symp. Inf. Theory (ISIT)*, Barcelona, Spain, Jul. 2016, pp. 2644-2648.

[17] S. M. Yazdi, H. M. Kiah, E. R. Garcia, J. Ma, H. Zhao, and O. Milenkovic, “DNA-based storage: trends and methods,” in *IEEE Trans. Mol. Biol. Multi-Scale Commun.* vol. 1, no. 3, pp. 230-248, Sep. 2015.

[18] R. Gabrys, H. M. Kiah, and O. Milenkovic, “Asymmetric Lee distance codes for DNA-based storage,” in *Proc. IEEE Int. Symp. Inf. Theory (ISIT)*, Hong Kong, China, Jun. 2015, pp. 909-913.

[19] H. M. Kiah, G. J. Puleo, and O. Milenkovic, “Codes for DNA storage channels,” in *Proc. IEEE Inf. Theory Workshop (ITW)*, Jerusalem, Israel, Apr. 2015, pp. 5.

[20] S. M. Yazdi, Y. Yuan, J. Ma, H. Zhao and O. Milenkovic, “A rewritable, random-access DNA-based storage system,” *Nature Scientific Reports*, vol. 5, no. 14138, 2015.

[21] R. N. Grass, R. Heckel, M. Puddu, D. Paunescu, and W. J. Stark, “Robust chemical preservation of digital information on DNA with error-correcting codes,” *Angewandte Chemie International Edition*, vol. 54, no. 8, Feb. 2015, pp. 2552-2555.

[22] J. Bornholt, R. Lopez, D. M. Carmean, L. Ceze, G. Seelig, and R. Strauss, “A DNA-based archival storage system,” in *Proc. Int. Conf. Architectural Support for Programming Languages and Operating Syst. (ASPLOS)*, Atlanta, GA, Apr. 2016.

[23] Y. Liron and M. Langberg, “A characterization of the number of subsequences obtained via the deletion channel,” in *IEEE Trans. Inf. Theory*, vol. 61, no. 5, pp. 2300-2312, May 2015.

[24] A. A. Kulkarni and N. Kiyavash, “Non-asymptotic upper bounds for single-deletion correcting codes,” in *IEEE Trans. Inf. Theory*, vol. 59, no. 8, pp. 5115-5130, Aug. 2013.

[25] D. Cullina, N. Kiyavash, and A. A. Kulkarni, “Restricted composition deletion correcting codes,” in *IEEE Trans. Inf. Theory*, vol. 62, no. 9, pp. 4819-4832, Sept. 2016.

[26] T. Jiang and A. Vardy, “Asymptotic improvement of the Gilbert-Varshamov bound on the size of binary codes,” *IEEE Trans. Inf. Theory*, vol. 50, no. 8, pp. 1655-1664, Aug. 2004.

[27] V. Vu and L. Wu, “Improving the Gilbert-Varshamov bound for $q$-ary codes,” *IEEE Trans. Inf. Theory*, vol. 51, no. 9, pp. 3200-3208, Sep. 2005.

[28] V.I. Levenshtein, “Binary codes capable of correcting deletions, insertions, and reversals,” *Soviet Physics Doklady*, vol. 10, no. 8, 1966.

[29] R. R. Varshamov and G. M. Tenengolts, “Codes which correct single asymmetric errors,” *Avtom. i Telemeh.*, vol. 26, no. 2, pp. 288-292, 1965.

[30] L. Calabi and W.E. Hartnett, “Some general results of coding theory with applications to the study of codes for the correction of synchronization errors,” *Inf. Control*, vol. 15, no. 3, 1969.

[31] R. Gabrys and E. Yaakobi, “Sequence reconstruction over the deletion channel,” in *Proc. IEEE Int. Symp. Inf. Theory (ISIT)*, Barcelona, Spain, Jul. 2016, pp. 1590-1596.

[32] D. Cullina, A. Kulkarni, and N. Kiyavash, “A coloring approach to constructing deletion correcting codes from constant weight subgraphs,” in *Proc. IEEE Int. Symp. Inf. Theory (ISIT)*, Cambridge, MA, Jul. 2012, pp. 513-517.

[33] S. Mason, C. Berger, S. Zhou, and P. Willett, “Detection, synchronization, and Doppler scale estimation with multicarrier waveforms in underwater acoustic communication,” *IEEE J. Sel. Areas Commun.*, vol. 26, no. 9, pp. 1638-1649, Dec. 2008.

[34] R. Venkataramanan, V. Swamy, and K. Ramchandran, “Low-complexity interactive algorithms for synchronization from deletions, insertions, and substitutions,” *IEEE Trans. Inf. Theory*, vol. 61, no. 10, pp. 5670-5689, Oct. 2015.

[35] F. Salá, C. Schoeny, N. Bitouzé, and L. Dolecek, “Synchronizing files from a large number of insertions and deletions,” *IEEE Trans. Commun.*, vol. 64, no. 6, pp. 2258-2273, Jun. 2016.

[36] H. S. Wilf, *Generatingfunctionology*. San Diego, CA: Academic Press, 1990.
Step (a) follows from the definition of $\mathcal{N}$. In step (c), we changed the range of summation for $i$ from $[0, k-j-1]$ to $[1, k-j]$ for the second term. In (d), we broke up the sum for the first term, removing the components with $i = 0$ in the inner sum. In step (e), we note that there is no inner sum for $j = k$, so we change the limit of the outer sum to $k-1$. We then combined terms. In step (f) we applied the identity $(n+1-i)+ (n+1-i) = (n+i)$. All other steps are immediate rearrangements of terms.

Next, for the second recursion, we have that

$$\mathcal{N}^+_q(n, t, k, \ell) = \mathcal{N}^+_q(n-1, t, k-1, \ell-1)$$

$$+ (q-2) \mathcal{N}^+_q(n, t-1, k-1, \ell)$$

(a)$$\sum_{j=1}^{k-j} \sum_{i=0}^{k-j-1} \left( t - k + 2j \right) \left( t + j - i \right) \left( t - k - 2j \right) \left( t - j - i \right)$$

$$\left( \sum_{j=1}^{k-j-1} \sum_{i=0}^{k-j-1} \left( t - k + 2j \right) \left( t - j - i \right) \right)$$

+ $\sum_{j=t}^{k-j-1} \sum_{i=0}^{k-j-1} \left( t - k + 2j \right) \left( t - j - i \right) \left( t - k - 2j \right) \left( t - j - i \right)$

$$= \mathcal{N}^+_q(n, t, k, \ell).$$

$$+ \sum_{j=t}^{k-j-1} \sum_{i=0}^{k-j-1} \left( t - k + 2j \right) \left( t - j - i \right) \left( t - k - 2j \right) \left( t - j - i \right)$$

$$\left( \sum_{j=t}^{k-j-1} \sum_{i=0}^{k-j-1} \left( t - k + 2j \right) \left( t - j - i \right) \right)$$

+ $\sum_{j=t}^{k-j-1} \sum_{i=0}^{k-j-1} \left( t - k + 2j \right) \left( t - j - i \right) \left( t - k - 2j \right) \left( t - j - i \right)$

$$= \mathcal{N}^+_q(n, t, k, \ell).$$

$$+ \sum_{j=t}^{k-j-1} \sum_{i=0}^{k-j-1} \left( t - k + 2j \right) \left( t - j - i \right) \left( t - k - 2j \right) \left( t - j - i \right)$$

$$\left( \sum_{j=t}^{k-j-1} \sum_{i=0}^{k-j-1} \left( t - k + 2j \right) \left( t - j - i \right) \right)$$

+ $\sum_{j=t}^{k-j-1} \sum_{i=0}^{k-j-1} \left( t - k + 2j \right) \left( t - j - i \right) \left( t - k - 2j \right) \left( t - j - i \right)$

$$= \mathcal{N}^+_q(n, t, k, \ell).$$

$$+ \sum_{j=t}^{k-j-1} \sum_{i=0}^{k-j-1} \left( t - k + 2j \right) \left( t - j - i \right) \left( t - k - 2j \right) \left( t - j - i \right)$$

$$\left( \sum_{j=t}^{k-j-1} \sum_{i=0}^{k-j-1} \left( t - k + 2j \right) \left( t - j - i \right) \right)$$

+ $\sum_{j=t}^{k-j-1} \sum_{i=0}^{k-j-1} \left( t - k + 2j \right) \left( t - j - i \right) \left( t - k - 2j \right) \left( t - j - i \right)$

$$= \mathcal{N}^+_q(n, t, k, \ell).$$
\[
\left[\left(\frac{t + j - i - 1}{t - k + 2 j - 1} + \frac{t + j - i - 1}{t - k + 2 j}\right) \times \left(\frac{n + t - 1}{i}\right)(q - 1)^i (-1)^{k+j-i}\right]
+ \sum_{j=\ell}^k \left(\frac{t - k + 2 j}{j}\right) \left(\frac{t + j - (k - j) - 1}{t - k + 2 j - 1}\right) \times \left(\frac{n + t - 1}{k - j}\right)(q - 1)^{k-j} (-1)^{k+j-(k-j)}
+ \sum_{j=\ell}^{k-1} \sum_{i=0}^{k-j-1} \left(\frac{t - k + 2 j}{j}\right) \left(\frac{t + j - i - 1}{t - k + 2 j}\right) \times \left(\frac{n + t - 1}{i}\right)(q - 1)^i (-1)^{k+j-i}
+ \sum_{j=\ell}^k \left(\frac{t - k + 2 j}{j}\right) \left(\frac{t + j - i}{t - k + 2 j}\right) \times \left(\frac{n + t - 1}{i}\right)(q - 1)^i (-1)^{k+j-i}
\]

The steps we used are the following. In (a) we applied the definition of \(\mathcal{N}\). In (b), we changed the range of summation for \(j\) in the middle term from \([\ell - 1, k - 1]\) to \([\ell, k]\). In (d), we used the identity \(\left(\frac{t - k + 2 j}{j}\right) + \left(\frac{t + j - k - 1}{j}\right) = \left(\frac{t - k + 2 j}{j}\right)\). In (e), we broke up the second term from (d), which is multiplied by a factor of \((q - 2)\) into two terms, one multiplied by a factor of \((q - 1)\) and the other by \((-1)\). In (f), we combined the first two terms from (e). In (g), we used the identity \(\left(\frac{t + j - i - 1}{t - k + 2 j - 1}\right) + \left(\frac{t + j - i}{t - k + 2 j}\right) = \left(\frac{t + j - i}{t - k + 2 j}\right)\). In (h), we changed the range of summation for \(i\) in the second term from \([0, k - j - 1]\) to \([1, k - j]\). In (i), we combined terms and again applied the identity \((\frac{n + t - 1}{i}) + (\frac{n + t - 1}{i - 1}) = (\frac{n + t}{i})\). We also combined the last two summands, using the identity \((\frac{n + t - 1}{k - j - 1}) + (\frac{n + t - 1}{k - j}) = (\frac{n + t}{k - j})\). In (k) we combined all remaining terms.

**B. Proof of Lemma 8**

We present the proof of the two auxiliary combinatorial identities.

**Lemma 8.**

1. For \(m \geq 0\),
\[
\sum_{j=0}^m \left(\frac{2j}{j}\right) \left(\frac{m + j}{2 j}\right)(-1)^{m+j} = 1.
\]

2. For \(n, m, t, j \geq 0\) and \(t + j \geq m\),
\[
\sum_{i=0}^m \left(\frac{t + j - i}{t + j - m}\right) \left(\frac{n + t}{i}\right)(-1)^{m-i} = \begin{pmatrix} n + m - j - 1 \end{pmatrix}_m.
\]

**Proof:** Both identities will be proved by a generating function approach. This strategy is described as the “snake oil method” in [26]. The idea is that the right-hand side of each identity has an easily-derived generating function, while we will perform more complex manipulations to derive an identical generating function for the left-hand side. For the first identity, the generating function \(F(x)\) for the left-hand side is written as
\[
F(x) = \sum_{m \geq 0} \sum_{j=0}^m \left(\frac{2j}{j}\right) \left(\frac{m + j}{2 j}\right)(-1)^{m+j}
= \sum_{j=0}^\infty \sum_{m \geq j} x^m \left(\frac{2j}{j}\right) \left(\frac{m + j}{2 j}\right)(-1)^{m+j}
\]
Thus we conclude that \( m \) is the fifth and seventh steps, we applied the binomial theorem.

The fifth step has generating function

\[ \sum_{j=0}^{\infty} \binom{2j}{j} x^{-j} \sum_{m \geq j} \binom{m+j}{2j} (-x)^{m+j} \]

\[ = \sum_{j=0}^{\infty} \binom{2j}{j} x^{-j} \sum_{r \geq j} \binom{r}{2j} (-x)^r \]

\[ = \sum_{j=0}^{\infty} \binom{2j}{j} x^{-j} \frac{(-x)^{2j}}{(1+x)^{2j+1}} \]

\[ = \frac{1}{1+x} \sum_{j=0}^{\infty} \binom{2j}{j} \frac{x}{(1+x)^2} \]

\[ = \frac{1}{1+x} \frac{1}{\sqrt{1-4x}} \]

\[ = \frac{1}{1+x} \frac{1}{\sqrt{1-4x}} \]

\[ = \frac{1}{1+x} \frac{1}{\sqrt{1-4x}} \]

In the fourth step, we replace \( m + j \) with \( r \). We can start the sum at \( r' = 0 \) since the binomial term \( \binom{r'}{j} \) evaluates to 0 for all \( m < j \). Next, in the fifth step, we use the series \( \sum_{r \geq j} \binom{r}{2j} (-x)^r \), which is the only condition for this identity is \( 2j \geq 0 \). Next, in the seventh step, we applied the generating function for the central binomial coefficients

\[ \sum_{j=0}^{\infty} \binom{2j}{j} x^j = \frac{1}{\sqrt{1-4x}}. \]

Thus we conclude that \( F(x) = 1 + x + x^2 + \ldots \), so indeed \( \sum_{j=0}^{\infty} \binom{2j}{j} (-x)^{j+m+j} = 1 \).

We use the same approach for the second identity. The right-hand side of the identity counts the number of ways to distribute \( m \) items in \( n-j \) buckets. It is easy to see that this quantity has, with respect to \( m \), the generating function

\[ (1 + x + x^2 + \ldots)^{n-j} = (1-x)^{-(n-j)}. \]

The left-hand side has generating function

\[ F(x) = \sum_{m \geq 0} \sum_{i=0}^{\infty} x^m \sum_{t+j=i} \binom{t+j-i}{t+j-m} \binom{n+t}{i} (-1)^{m-i} \]

\[ = \sum_{i=0}^{\infty} \binom{n+t}{i} x^i \sum_{m \geq i} \binom{t+j-i}{t+j-m} (-1)^{m-i} \]

\[ = \sum_{i=0}^{\infty} \binom{n+t}{i} x^i \sum_{r \geq i} \binom{r}{t+j-r+i} (-1)^r \]

\[ = \sum_{i=0}^{\infty} \binom{n+t}{i} x^i \sum_{r \geq i} \binom{r}{t+j-r+i} (-1)^r \]

\[ = \sum_{i=0}^{\infty} \binom{n+t}{i} x^i (1-x)^{t+j-i} \]

\[ = (1-x)^{t+j+i} \sum_{i=0}^{\infty} \binom{n+t}{i} \frac{x}{1-x} \]

\[ = (1-x)^{t+j+i} \left( 1 + \frac{x}{1-x} \right) n+t \]

\[ = (1-x)^{t+j+i}(1-x)^{-(n+t)} \]

\[ = (1-x)^{j-n} \]

and we are done. In the third step, we write \( m - i = r \). In the fifth and seventh steps, we applied the binomial theorem.

C. Proof of Lemma [13]

**Lemma 13.** For any integer \( n \geq 7 \), and integer \( 0 \leq m \leq n \), there exists a subset \( T' \subseteq FP(n) \) where \( \sum_{t \in T'} i = m \). In addition, \( \log_2(n) \leq |FP(n)| \leq \log_2(n) + 1 \). Furthermore, if \( n \) is even, we have \( \frac{m}{2} + 1 \notin FP(n) \) and if \( n \) is odd, then \( \frac{m}{2} + 2 \notin FP(n) \).

**Proof:** The third sentence in Lemma [13] is trivially true based on the steps in the algorithm. There are four possible cases of \( FP(n) \), dependent on \( n \). These cases follow directly from step 2 of the algorithm. We will show that each of these cases satisfies the conditions of Lemma [13]. In particular, for each case we will show that there exists a subset \( T' \subseteq FP(n) \) where \( \sum_{t \in T'} i = m \) (for \( 0 \leq m \leq n \)), and that \( \log_2(n) \leq |FP(n)| \leq \log_2(n) + 1 \). In the following analysis, we make extensive use of the fact that there exists a subset of the set \( \{2^0, 2^1, \ldots, 2^k\} \) that sums to every positive integer up to \( 2^{k+1} - 1 \). For the following cases, assume \( n \geq 7 \) and \( k \) is an integer.

**Case 1:** \( n = 2^k \) or \( n = 2^k + 1 \). In this case, we have \( FP(n) = \{2^0, 2^1, \ldots, 2^{k-2}, 2^{k-1}, 1, 2^k - 1\} \). We have that \( |FP(n)| = k + 1 \), which satisfies the size condition in Lemma [13]. Note that the first \( k - 1 \) elements can sum to any integer up to \( 2^{k-1} - 1 \). Including the next element in \( FP(n) \), \( 2^{k-1} - 1 \), we can now sum up to any integer \( 2^{(2^{k-1} - 1)} = 2^k - 2 \). Including the last element in the set, we see that there are no gaps and there exists a subset of \( FP(n) \) that sums up to any integer \( 0 \leq m \leq n \).

**Case 2:** \( n = 2^k - 2 \). In this case, we have \( FP(n) = \{2^0, 2^1, \ldots, 2^{k-2}, 2^{k-1} - 2, 2^k - 2\} \). We have that \( |FP(n)| = k + 1 = \log_2(n) + 1 \), which satisfies the size condition in Lemma [13]. Again, note that the first \( k - 1 \) elements can sum to any integer up to \( 2^{k-1} - 1 \). Including the next element in \( FP(n) \), \( 2^{k-1} - 2 \), we can now sum up to any integer \( 2^{(2^{k-1} - 2)} = 2^k - 3 \). The final element in \( FP(n) = 2^k - 2 \), which is equal to \( n \) itself.

**Case 3:** \( n = 2^k - 3 \). In this case, we have \( FP(n) = \{2^0, 2^1, \ldots, 2^{k-2}, 2^{k-1} - 1\} \). We have that \( |FP(n)| = k = \log_2(n) \), which satisfies the size condition in Lemma [13]. Again, note that the first \( k - 1 \) elements can sum to any integer up to \( 2^{k-1} - 1 \). Including the next element in \( FP(n) \), \( 2^{k-1} - 1 \), we can now sum up to any integer \( 2^{(2^{k-1} - 1)} = 2^k - 2 \). The final element in \( FP(n) = 2^k - 3 \), which is equal to \( n \) itself.

**Case 4:** All other \( n \geq 7 \). In this case, we simply have \( FP(n) = \{2^0, 2^1, \ldots, 2^{\log_2(n)}\} \). We have that \( |FP(n)| = k + 1 = \log_2(n) \). Additionally, since this set is simply all the powers of 2 less than \( n \), we know that there exists a subset of \( FP(n) \) that can sum to any integer \( 0 \leq m \leq n \).

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