Optimal Las Vegas reduction from one-way set reconciliation to error correction

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Abstract
Suppose we have two players $A$ and $C$, where player $A$ has a string $s[0..u-1]$ and player $C$ has a string $t[0..u-1]$ and none of the two players knows the other’s string. Assume that $s$ and $t$ are both over an integer alphabet $[\sigma]$, where the first string contains $n$ non-zero entries. We would wish to answer to the following basic question. Assuming that $s$ and $t$ differ in at most $k$ positions, how many bits does player $A$ need to send to player $C$ so that he can recover $s$ with certainty? Further, how much time does player $A$ need to spend to compute the sent bits and how much time does player $C$ need to recover the string $s$?

This problem has a certain number of applications, for example in databases, where each of the two parties possesses a set of $n$ key-value pairs, where keys are from the universe $[u]$ and values are from $[\sigma]$ and usually $n \ll u$. In this paper, we show a time and message-size optimal Las Vegas reduction from this problem to the problem of systematic error correction of $k$ errors for strings of length $\Theta(n)$ over an alphabet of size $2^{\Theta(\log \sigma + \log (u/n))}$. The additional running time incurred by the reduction is linear randomized for player $A$ and linear deterministic for player $B$, but the correction works with certainty. When using the popular Reed-Solomon codes, the reduction gives a protocol that transmits $O(k(\log u + \log \sigma))$ bits and runs in time $O(n \cdot \text{polylog}(n)(\log u + \log \sigma))$ for all values of $k$. The time is randomized for player $A$ (encoding time) and deterministic for player $C$ (decoding time). The space is optimal whenever $k \leq (u\sigma)^{1-\Omega(1)}$.

1 Introduction
Suppose we have two strings $s$ and $t$ of equal length $u$ over an integer alphabet $[\sigma]$, where the number of non-zeros in the first string equals $n$ and the Hamming distance between the two strings is at most $k$ (the number of positions $i$ such that $s[i] \neq t[i]$)

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Moreover suppose that we have two players $A$ and $C$, where player $A$ knows $s$, but does not know $t$ and player $C$ knows $t$ but does not know $s$. That is, the only information a player has about the other’s string is the upper bound $k$ on the Hamming distance between the two strings.

Then we would like to design a one-way protocol in which player $A$ computes some data on his string $s$ and sends it to player $C$, so that the latter can recover the string $s$. How good can such a protocol be, in terms of the size of the sent data and in terms of time complexity?

This problem can be considered as a variant of the document exchange problem, which is defined as follows. Player $A$ has a document $s$ and player $C$ has a document $t$ and the only knowledge that a player has about the other player’s document is an upper bound on some distance between the two documents (the distance can be edit distance, Hamming distance or some other distance). Then the two players must exchange some data so that each of the two can recover the document that is held by the other. Other variants of the problem have already been studied in the literature, most notably document exchange under the edit distance (see paper [15] and references therein).

The problem is also known in the literature as the set reconciliation problem [19, 14]. In the database context, the problem can be considered as that of synchronizing two sets of $n$ key-value pairs, where the keys are from universe $[u]$ and values are from the set $[\sigma]$.

The problem can be solved with optimal message size (up to constant factor) using traditional error correcting codes, like Reed-Solomon codes 1. However, those solutions would require time $\Omega(u)$. In many cases $n \ll u$ and time $\Omega(u)$ is unacceptable (for example in database context, we could have $u = 2^{64}$ or $u = 2^{128}$ with $n$ ranging from $2^{10}$ to $2^{30}$).

It is thus much more preferable to have a time that depends on $n$ (the number of non-zero elements) rather than $u$. The best one could wish for is a time linear in $n$, since even reading or writing a representation of the strings $s$ or $t$ (even in compressed form) requires $\Theta(n)$ time in the worst-case.

There already exist solutions to the problem with time bounds that depend only on $n$ and $k$ (possibly with logarithmic dependence on $u$). We review some of those solutions. In all cases an alphabet size $\sigma = 2$ is assumed, but the solutions can easily be extended to support an arbitrary alphabet by replacing $u$ by $u\sigma$ 2.

The deterministic solution proposed in [19] uses polynomials over finite fields. The protocol transmits $O(k \log u)$ bits and has time complexity $O(nk + k^3 \log u)$. The term $nk$ is needed for evaluating a polynomial at $k$ given points. We note that the term can be improved to $O(n \cdot \text{polylog}(u))$ by using faster (and more sophisticated) operations on polynomials (e.g. using the discrete Fourier transform [13]). The term $k^3 \log u$ comes from the rational function interpolation and the factoring of two polynomials of degree $k$. The best al-

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1 An optimal message-size requires a code with $\Theta(k \log(\sigma/k))$ bits of redundancy. Since Reed-Solomon codes have $O(k \log(\sigma))$ bits of redundancy they will lead to optimal message size $O(k \log(\sigma/k))$ bits only if $k \leq (\sigma)^{1-\epsilon}$ for some constant $\epsilon > 0$. Notice however, there provably exist less time efficient (exponential time) codes that achieve the required redundancy [13, 27] for all values of $k$.

2 That is, given $s$ (and $t$) one can construct a string of length $u\sigma$, by expanding every element $s[i]$ (resp. $t[i]$) into a unitary binary vector of length $\sigma$ that contains a single 1 at position $s[i]$ (resp. $t[i]$).

3 In this paper we define $\log x = \lfloor \log_2(\max(x, 2)) \rfloor$. 2
algorithm currently known for factoring a polynomial of degree $k$ runs in time $\Omega(k^{3/2})$ \[17\]. Thus the total time will be $n^{1+\Omega(1)}$ when $k$ is at least $n^{2/3+\Omega(1)}$.

A randomized Monte-Carlo solution achieving $O(n \log u)$ running time and $O(k \log u)$ transmitted bits was achieved in \[23\]. That solution only guarantees a successful reconciliation with high probability. Finally approaches based on Invertible Bloom filters \[13, 14, 10, 20\] achieve $O(n \log u)$ time with $O(k \log u)$ bits, but again working with only probabilistic guarantees\footnote{The analysis of those solutions is not fully rigorous though, as the solutions do not specify the used hash functions.}.

The main result of this paper is to show a time and message-size optimal Las-Vegas reduction from our problem to the problem of systematic error correction of $k$ errors for strings of length $\Theta(n)$ over an alphabet of size $2^{\Theta(\log \sigma + \log(u/n))}$.

The high level idea is to use the two-level hashing scheme of \[11\] in which the second level is made deterministic using the deterministic hashing results from \[24\] and \[12\].

This first step takes $O(n)$ expected time in total. Then player $A$ applies a systematic error correction on the resulting data structure built on his string $s$ and sends the redundancy to player $C$. Finally player $C$ uses the redundancy he receives from player $A$ along with the knowledge of his own string $t$ to recover in (deterministic) linear time the two level hashing data structure that was built by player $A$ \footnote{For this step to run in deterministic time, it it crucial that the second level of the hashing is deterministic.}. The latter allows player $C$ to recover the whole string $s$.

As a corollary of our main result, we show the first (to the best of our knowledge) Las-Vegas solution with $O(k(\log u + \log \sigma))$ transmitted bits and time $O(n \cdot \text{polylog}(n)(\log u + \log \sigma))$ independent of $k$. The time is randomized for player $A$ and worst-case for player $C$.

## 2 Preliminaries

### 2.1 Model and Notation

We assume a word RAM model with word size $w = \Omega(\log u + \log \sigma)$, and in which all standard arithmetic and logic operations (including multiplication but not division) are assumed to take constant time. All the considered alphabets are integer alphabets $[\beta]$, where $\beta$ is called the alphabet size and $\log \beta$ is called the bitlength of the alphabet. We use $[\alpha]$ to denote the set $[0..\alpha - 1]$. We denote the string of player $A$ by $s$ and the string of player $C$ by $t$. The length of the first string is denoted by $n$ and the maximal Hamming distance between the two strings by $k$. Throughout it is implicitly assumed that the strings are given in the following compressed form. The string $s$ is represented by a list of $n$ pairs $(p, s[p])$ where $s[p] \neq 0$. Similarly $t$ is represented by a list of pairs that represents all non-zero positions in $t$ along with their values. Both lists can be in arbitrary order.

### 2.2 Tools

We now present the main tools that will be used in our constructions. Given a length $n$ and an error threshold $k$, an error correcting code encodes any string $s$ of length $n$ into a longer string $s'$ such that $s$ can be recovered from $s'$, even if up
Lemma 1 [16, 22] There exists a systematic error correcting code that encodes a string $s$ of length $n$ over an alphabet $[n]$ into a string $s'$ of length $n + \Theta(k)$ over an alphabet of size $\Theta(n)$ and that allows to correct up to $k$ errors in $s'$. Moreover the encoding and decoding can be done in $O(n \cdot \text{polylog}(n))$

Some of our results will make use of two special-purpose instructions defined as follows:

Definition 1 Given a number $xx_1x_2...x_r \in [2^r]$, we define the instruction MSB$(x)$ as the one that returns the that isolates the most significant set bit. That is, given MSB$(x) = y$ if and only if $x = x_1x_2...x_{y-1}10^{r-y}$

Definition 2 that given an integer $x = x_1x_2...x_r \in [2^r]$ and mask $M \in [2^r]$. We define the instruction PACK$(x,M)$ as the function returning the number $x' = x_{i_1}x_{i_2}...x_{i_t}0^{r-t}$, where $i_1 < i_2 < ... < i_t$ are the positions of bits set to 1 in $M$.

We will use the following two results about randomized hashing.

Lemma 2 [6, consequence of Theorem 3] Given two numbers $u$ and $n$ with $n < u$ and a set $S \subseteq [u]$ with $|S| = n$, we can find in expected $O(n)$ time two numbers $a, b \in [u \cdot n]^2$ such that $f(x) \neq f(y)$ for all $x, y \in S$ and $x \neq y$ where $f$ is a function from $[u]$ into $[n^2]$ such that $f(x) = ((x \cdot a + b) \text{ div } (u \cdot n^2)) \text{ div } u$.

The lemma derives from the fact that the family of hash functions $f$ parametrized by $a$ and $b$ is 2-wise independent ( [6, Theorem 3]).

Lemma 3 For an arbitrary given prime $P$, a number $n$ with $n^2 < P$ and a set $S \subseteq [P]$ with $|S| = n$ one can in expected time $O(n)$ find three integers $a, b, c \in [P]$ such that for the function $f(x) = (ax^2 + bx + c) \text{ mod } P$ the following holds:

1. $(f(x) \text{ mod } n^2) \neq (f(y) \text{ mod } n^2)$ for all pairs $(x, y) \in S^2$ with $x \neq y$.
2. $\sum_{i=0}^{n-1} |S_i|^3 = \alpha \cdot n$, where $S_i = \{x | x \in S, (f(x) \text{ mod } n) = i\}$.

The lemma derives easily from the fact that the family of hash functions $f$ parametrized by $a$, $b$ and $c$ is 3-wise independent [28]. That is, by 2-wise independence, the probability of not fulfilling first condition is at most $1/2$ and by 3-wise independence the probability of not fulfilling second condition can
be made at most 1/4 by appropriate choice of $\alpha$. Thus, by union bound the probability for both conditions to be false is at most 3/4, and hence appropriate values for parameters $a, b, c$ can be be found after an expected constant number of trials, each time checking whether the condition are fulfilled in time $O(n)$. Thus the expected time to find appropriate values for $a, b, c$ is $O(n)$.

We will also use the following two (well-known) results about deterministic hashing:

**Lemma 4** [24, consequence of Lemma 6]

Given two numbers $r$ and $s$ with $s < r$ and a set $S \subset [2^r]$ with $|S| = n \leq 2^{r/2}$, then it is possible to find in deterministic $O(n^2r)$ time two numbers $a, b \in [2^r]$, such that $f(x) \neq f(y)$ for all $x, y \in S$ and $x \neq y$ with $f(x) = ((x \cdot a + b) \mod 2^r) \div 2^{r-s}$.

**Lemma 5**

Given two integers $r$ and $n$ with $n \leq r \leq w$, there exists a family $F$ of hash functions from $[2^r]$ into $[2^{n-1}]$ such that a member $f$ of the family can be described using $O(r)$ bits, $f(x)$ can be evaluated in constant time and moreover given any set $S \subset [2^r]$ of size $n$, one can in deterministic $O(n \log n)$ time find a member $f$ of the family such that $f(x) \neq f(y)$ for any $x, y \in S$ with $x \neq y$. The construction time assumes the availability of an instruction $\text{MSB}$ that can be evaluated in constant time.

The evaluation time assumes the availability of an instruction $\text{PACK}$ that can be evaluated in constant time.

The construction is done as follows. The perfect hash function is simply the concatenation of (at most) $n-1$ selected bits out of the $r$ bits. The hash function description consists in a word $A$ considered as a bitvector that marks (with a 1) the selected bits. The description is built as follows. Initially all bits in $A$ are set to zero.

We then sort the elements of the set $S$ in time $O(n \log n)$. We let that order be $x_1, x_2, \ldots, x_n$. We then isolate the most significant set bit in $(x_i \text{ XOR } x_{i+1})$ for all $i \in [1, n-1]$ (using the MSB instruction) and OR the result with $A$. The end result is a word $A$ that contains $n' \leq n-1$ ones. Then evaluating $f(x)$ amounts to computing the word $x'$ which contains in its least $n'$ significant bits the bits in $x$ whose corresponding positions are set to one in $A$ (using the PACK instruction). The function $f$ will be perfect since every pair of keys will differ in at least one of the selected $n'$ positions [12].

### 3 The reduction

We establish a general reduction from the set reconciliation on strings of length $u$ over an alphabet $[\sigma]$, where one of the two strings has $n$ non-zeros and the two strings differ in at most $k$ positions to three instances of systematic error encoding that correct $k$ errors on strings of length $\Theta(n)$ over alphabets of bitlength (at most) $\Theta(\log(u/n) + \log \sigma)$. The reduction is error-free and uses $O(n)$ randomized time (i.e. the reduction is Las-Vegas) for the sender and deterministic $O(n)$ time for the reader.

The main idea of the reduction is to use the framework of two-level hashing introduced in [11] and known as FKS with a careful choice of hash functions and combining it with systematic error correction encoding.
More precisely the sender will produce a two level hashing. The first level consists in a global hash function that maps the n non-zero positions of its string into an array of n buckets. The second level is an array of n hash functions where each hash function is specific to a bucket and will map the m keys of the bucket injectively into the range \([m^2]\). By a careful choice of first level and second level hash functions, we can ensure that the total construction time is linear and that the sum of the range sizes of all buckets will also be linear.

The second level of the hashing scheme will provide an injective mapping from positions into a range of size \(n' = \Theta(n)\) and the final step will consist in storing in a table of size \(n'\) the non-zero positions and their value (character). The sender then applies systematic error correcting codes on three tables: a table that stores the bucket hash functions, a table that stores the number of keys assigned to each bucket and the table of pairs of non-zero positions and their values. Finally the sender sends a description of the first level hash function and the redundancies built on the three tables. The receiver will then rebuild a two-level hash function scheme on its own keys, but taking care of doing it progressively in three steps where at each step the output is corrected using the received redundancy. That is at first step, the receiver simply uses the global hash function he received. He then corrects the table that stores the number of keys assigned to each bucket. He then builds the hash functions specific to each bucket, but ignoring the buckets that have a different size from the sender. After that he corrects the table of bucket hash functions with the received redundancies and applies the hash function on its own keys to build the table of positions and values and corrects it using the corresponding redundancy of the sender. Finally reconstructing the string of the sender can be done by scanning the corrected table of positions and values. An attentive reader might wonder whether the following simple approach would work: build the whole FKS scheme on both the sender and receiver side and reconcile them using an error correcting code built on the concatenation of the FKS components of the sender. We argue that this approach will potentially have much more \(O(k)\) errors. For example, if one uses the different global hash functions, then the resulting buckets would be completely different, even if the set of keys is the same. Likewise using different bucket hash functions will result in a table of pairs and values containing potentially much more than \(k\) mismatches. Thus synchronizing the components one-by-one is really necessary to ensure that the number of errors at next component is upper bounded by \(k\).

### 3.1 Sender protocol

The reduction uses the two-level hashing scheme of [11]. Player A builds the two-level hashing scheme on its set of \(n\) keys over a universe of size \(u\), where the keys are the positions of the non-zeros in \(s[0..u-1]\).

The first level consists in a hash function \(g_1\) computed using Lemma 2 and a hash function \(g_2\) computed using Lemma 3 (hash function \(f\)). The hash function \(g_1\) is computed on the set \(S\) mapping \(S\) injectively into \(S' \subset [n^2]\) and then \(g_2\) is computed on \(S'\).

In order to apply Lemma 3, we first find a prime \(P\) in \([n^2, 2n^2 - 1]\) using a method described in [8]. This is done in time \(o(n)\) by repeatedly selecting a random number in \([n^2, 2n^2 - 1]\) and testing whether it is prime or not using a deterministic primality testing algorithm in [2]. Since the density of the primes
in the interval \([n^2, 2n^2 - 1]\) is \(\Theta(1 / \log n^2) = \Theta(1 / \log n)\) and the primality testing algorithm runs in time \(O(\text{polylog}(n))\), we conclude that the time to find a prime is \(O(\text{polylog}(n))\).

We let \(f\) be such that \(f(x) = g_2(g_1(x))\). The hash function description occupies \(O(\log u + \log P) = O(\log u)\) bits. We additionally define the functions \(f_1\) and \(f_2\) as follows. We let \(f_1(x) = (f(x) \mod n)\) and \(f_2(x) = (f(x) \mod n^2)\). A key \(x\) will be mapped to bucket number \(i\) if and only if \(f_1(x) = i\). Denote by \(b_i\) the number of keys mapped to bucket \(i\). Then by Lemma 5 we will have

\[
\sum_{i=0}^{n-1} (b_i)^3 = O(n).
\]

The lemma also guarantees that the function \(f_2\) is injective. We could have used the lemma directly on the set \(S\), and we would have obtained the same result. The only reason we did not do so, is to avoid the \(O(\text{polylog}(u))\) cost associated with the search for a prime number in \([u, 2u - 1]\). The second level of the scheme is implemented over each bucket. Each bucket will use a local hash function \(h_i\) determined using lemmata 3 and 4.

The hash function \(h_i\) maps the \(b_i\) keys to \(4(b_i)^2\) cells (the factor 4 comes from the fact that we round \(b_i\) to the nearest power of two immediately above prior to using Lemma 4) so that each key is mapped to a distinct cell. This hash function occupies \(O(\log n)\) bits in total and is determined in \(O(\log n)\) time.

The hash function is determined as follows: for every bucket \(i\) to which the set of mapped keys is \(S_i\), we let \(S'_i \subseteq [n^2]\) be the image of \(S_i\) under the function \(f_2\) (note that \(f_2\) is injective on \(S_i\)). Whenever \(b_i > \log n\) (we call such buckets large and the others small), use Lemma 3 and \(h_i\) will be the resulting hash function that maps every \(x'\) in \(S_i\) to a unique number in the range \(\lfloor 4(b_i)^2 \rfloor\). If \(b_i \leq \log n\), we apply Lemma 5 generating a hash function \(h_{i,1}\) that injectively maps \(S'_i\) to a set \(S''_i \subseteq \lfloor 2^{b_i - 1} \rfloor\). We can easily simulate the instructions MSB and PACK using the four russians technique [4] after a preprocessing phase of \(O(n)\) time. Since the arguments are of length \(2\log n\) bits, we can use a table to simulate MSB on 4 chunks of length \(\lfloor \log n/2 \rfloor\) and another table to simulate the PACK instruction on 8 pairs of arguments of length \(\lfloor \log n/4 \rfloor\). We then use Lemma 4 on the set \(S''_i\) to generate a hash function \(h_{i,2}\). We let \(h_i\) be such that \(h_i(x) = h_{i,2}(h_{i,1}(x))\). Analysing the time to generate \(h_i\) on a set \(S'_i\), we can see that the computation in Lemma 4 takes \(O(b_i^2 \log n)\) if \(i\)th bucket is large and \(O(b_i^2 \log(2^{b_i - 1})) = O(b_i^2)\) otherwise, and the computation in Lemma 5 takes \(O(b_i \log b_i)\) in \(O(b_i^2)\). Each bucket stores a representation that consists in the number \(b_i\) and the description of the hash function \(h_i\) using \(O(\log n)\) bits. We store an array \(b[0..n-1]\) in which \(b[i] = b_i\). We also store an array \(B'[0..n-1]\) such that \(B'[i]\) stores the description of \(h_i\). If the set \(S_i\) is empty, then \(B'[i]\) will store a special value \(\text{null}\). The two arrays \(b\) and \(B\) together form the bucket representation. Note that the time to determine the bucket representation is \(O(n)\) since \(O(\sum_{i=0}^{n-1} (b_i)^3) = O(n)\).  

Finally the keys are stored in their associated cells as follows. We use a table \(\beta[0..N-1]\) of \(N = \sum_{i=0}^{n-1} 4(b_i)^2 = O(n)\) cells that store pairs \((x, c)\), where \(x \in [n]\) is a position in \(s\) and \(c \in [\sigma]\) is a character. The cells are assigned as follows: For each position \(x\) such that \(s[x] \neq 0\) and \(y = f_1(x)\), we store the pair \((x, s[x])\) in \(\beta[h_y(x) + j]\), where \(j = \sum_{i=0}^{y-1} 4(b_i)^2\). For every position \(k\) of \(\beta\) for which no pair was assigned, we assign the special value \(\text{null}\) to \(\beta[k]\). We call  

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\*We note that a hash function of the same form as that used in Lemma 4 can be constructed directly in time \(O((b_i)^2 \log b)\) time [20], and thus avoiding the need to use the hash function of Lemma 5. However the algorithm of [20] and its analysis seem more complicated than those used in the lemmata.
the resulting table $\beta$, the cell representation. Note that the resulting table has $O(n)$ cells each needing $\log \sigma + \log u$ bits.

In the next phase player $A$ uses a systematic error correcting code on top of each of the arrays $b$, $B$ and $\beta$ capable of correcting $k$ errors. The redundancies will be respectively $r(n, k, 2^{\Theta(u)})$, $r(n, k, 2^{\Theta(u)})$ and $r(n, k, 2^{\Theta(u+\log \sigma)})$.

Finally player $A$ sends the description of the hash function $f$ and the redundancies built on top of the arrays $b$, $B$ and $\beta$ for a total of $\log u + \Theta(n, k, 2^{\Theta(u+\log \sigma)})$ bits. Note that the time spent by player $A$ is randomized $O(n)$ in addition to the time needed to build the redundancies of the error correcting codes.

3.2 Receiver protocol

We now describe the recovery done by player $C$. Player $C$ starts by applying the hash function $f$ (which he got from player $A$) on the set $S'$ of keys consisting in positions of non-zeros in $t[0..u-1]$. He then assigns to each bucket $i$ the keys that are mapped to value $i$ by the hash function, and builds an array $b'[0..n-1]$ such that $b'[i]$ stores the number of keys assigned to the bucket number $i$. Then player $C$ uses the redundancy that player $A$ has built on top of $b$ to correct the array $b'$ into the array $b$. In the next step, for every bucket $i$ such that $b'[i] = b[i]$, player $C$ determines the hash function $h_i$ based on the keys assigned $S'_i$ to the bucket $i$ using exactly the same algorithm as that used by player $A$. He builds an array $B'[0..n-1]$ as follows. If $b'[i] \neq b[i]$, then $B'[i]$ will contain an arbitrary value. Otherwise, it will contain the description of the hash function $h_i$ (if $b[i] = 0$ then $B'[i]$ will contain null). Player $C$ can now recover the array $B$ from the array $B'$ using the redundancy that player $A$ has built on the array $B$. Next player $C$ builds a cell representation $\beta'$ based on the bucket representation of player $A$. That is, for every bucket $i$, apply the hash function $h_i$ on every key $x$ that is assigned to bucket $i$ and map it to position $\beta'[h_i(x) + j]$ where $j = 2 \sum_{i=0}^{t-1} (b[i])^2$. Note also that there could be collisions between the keys into each cell of player $C$. If more than one key is mapped to a cell then choose one of them arbitrarily and assign it to the cell. If a cell $i$ has no key assigned to it, then player $C$ sets $\beta'[i] = \text{null}$. Then player $C$ can correct its own cell representation $\beta'$ using the redundancy that has been previously built by player $A$ on its own cell representation $\beta[0..N-1]$.

3.3 Analysis

We argue that the correction will work because each of the pairs of arrays $(b, b')$, $(B, B')$, $(\beta, \beta')$ will differ in at most $k$ positions. To see why, notice that the (at least) $n-k$ keys that are in the intersection of $S'$ and $S$ are mapped to the same buckets. That means that at most $k$ buckets of player $C$ could get a different set of keys from the corresponding buckets of player $A$. This implies that the correction will work correctly for the array $b'$. Moreover, the fact that the determination of the functions $h_i$ is fully deterministic implies that the same $h_i$ will be built for every bucket $i$ that has the same set of assigned keys for both players. Therefore the correction will also work out for the array $B'$. We argue that the cell representations $\beta$ and $\beta'$ also differ in at most $k$ positions. Given a bucket $i$, denote by $k_i$ the number of keys that differ between bucket number $i$ of player $A$ and bucket number $i$ of player $C$. Then at least $b_i - k_i$ keys will be assigned to the same cell for both players. Thus at most $k_i$ cells of bucket
i will differ and overall the content of at most \( k = \sum_{i=0}^{i=n-1} (k_i) \) cells will differ between the two representations.

Note that the time spent by player C is deterministic \( O(n) \) in addition to the time needed to apply the error corrections. The analysis of the running time is trivial except for the step where player C builds the array \( B'[0..n-1] \). In that case, note that player C searches a function \( h_i \) only if \( b'_i = b_i \), spending exactly the same time spent by player A. Thus the total time player C spends on building the array \( B' \) is no larger than the time spent by player A on building the array \( B \).

**Theorem 6** We can solve the one-way set reconciliation problem with a protocol that transmits \( O(\log u + r(n, k, 2^{\Theta(\log \sigma + \log n)}) \) bits and that runs in randomized \( O(n + t_c(n, k, 2^{\Theta(\log \sigma + \log n)}) \) time for the sender side and deterministic time \( O(n + t_d(n, k, 2^{\Theta(\log \sigma + \log n)}) \) for the receiver side (the time becomes randomized if \( t_d \) is randomized).

To get an efficient scheme we will use Reed-Solomon codes as our underlying error correcting code. We divide each character to be represented into \( \Theta(\log u + \log \sigma) \) chunks of \( \log n \) bits each. Then we build \( \Theta(\log u + \log \sigma) \) different strings from the original string. The alphabet of the new strings will be of size \( n \) and the character number \( j \) of the string number \( i \) will receive the chunk number \( i \) of the character number \( j \) in the original string.

The redundancies will occupy a total of \( \Theta(k(\log u + \log \sigma)) \) bits of space.

**Theorem 7** We can solve the one-way set reconciliation problem with a protocol that transmits \( \Theta(k(\log u + \log \sigma)) \) bits and succeeds with certainty. The protocol runs in time \( O(n \cdot \text{polylog}(n) \cdot (\log u + \log \sigma)) \) for both parties. The time is randomized for the sender and deterministic for the receiver.

### 4 Improved reduction

We will now present a better reduction (actually optimal) with reduced alphabet size. Since the reduction above is optimal whenever \( u = n^{1+\Omega(1)} \) (i.e., the reduction implies a message size \( O(k(\log u + \log \sigma)) \) bits which is optimal up to constant factors), we will focus here only on the case \( u \leq n^{3/2} \).

**Theorem 8** We can solve the one-way set reconciliation problem when \( n \geq u^{2/3} \) with a protocol that transmits \( O(\log u + r(n, k, 2^{\Theta(\log \sigma + \log(u/n)}) \) bits and that runs in randomized \( O(n + t_c(n, k, 2^{\Theta(\log \sigma + \log(u/n)}) \) time for the sender side and deterministic time \( O(n + t_d(n, k, 2^{\Theta(\log \sigma + \log(u/n)}) \) for the receiver side (the time becomes randomized if \( t_d \) is randomized).

The main ingredient we will use to reduce the term \( \log u \) to \( \log(u/n) \), is a technique known as quotienting [21]. More in detail, we modify the previous scheme as follows. The first modification is to use only Lemma 3 to generate the function \( f \) directly on the set \( S \). Since \( S \subset [u] \subset \subseteq [n^{3/2}] \), we do not need to use Lemma 2 to first reduce the keys to universe \([n^{3/2}]\).

The second modification is to use a different finite field in Lemma 3. Instead of using the finite field \( \mathbb{F}_P \) for a prime \( P \geq u \), we will use the finite field
whether such codes admit attaining the Gilbert-Varshamov bound \( [13, 27] \). Unfortunately, we do not know otherwise, is a bit computed as follows. We solve the equations which determines two solutions \( X \). Then if \( X_0 = x \), we set \( I = 0 \), otherwise, \( X_1 = x \) and we set \( I = 1 \). Now, player \( C \) can recover a key \( x \) that is mapped to a bucket \( i \) and a cell containing the pair \((I, j)\) by solving the equation \( aX^2 + bX + c = f(x) \), which involves additions, subtractions, multiplication, inversion and square root operations. All the operations on \( F_q^2 \) can be implemented using a constant number of operations from the same set of operations on the base field \( F_q \), even square root and inversion \([1]\). All operations on \( F_q \) are trivially supported in constant time, except for square root and inversion which are implemented using lookup tables that use \( \gamma \) computed modulo the polynomial \( q \). That is array \( b[0..n-1] \) will now use \( \log(u/n) \) bits per entry since any bucket can get at most \( u/n \) keys assigned to it. Now the set \( S' \) is the image of \( S \) under the function \( f_2 \), and the hash function \( h_1 \) occupies \( O(\log(u/n)) \) bits. For the cell representation, we will replace every key \( x \) by pair \((I, f_2(x))\), where \( I \) is a bit computed as follows. We solve the equations \( aX^2 + bX + c = f(x) \), which determines two solutions \( X_0 < X_1 \). Then if \( X_0 = x \), we set \( I = 0 \), otherwise, \( X_1 = x \) and we set \( I = 1 \). Now, player \( C \) can recover a key \( x \) that is mapped to a bucket \( i \) and a cell containing the pair \((I, j)\) by solving the equation \( aX^2 + bX + c = j \cdot n + i \), obtaining two numbers \( X_0 \) and \( X_1 \) and \( x = X_1 \). Solving the equation involves additions, subtractions, multiplications, inversion and square root operations. All the operations on \( F_q^2 \) can be implemented using a constant number of operations from the same set of operations on the base field \( F_q \), even square root and inversion \([1]\). All operations on \( F_q \) are trivially supported in constant time, except for square root and inversion which are implemented using lookup tables that use \( O(n^{3/4} \log n) \) bits of space and are computed in time \( O(n^{3/4}) \). By combining theorems \([\delta]\) and \([\delta]\) we can state the following formal reduction:

**Theorem 9** We can solve the one-way set reconciliation problem with a protocol that transmits \( O(\log u + \pi(n, k, 2^{\Theta(\log \sigma + \log(u/n))})) \) bits and that runs in randomized \( O(n + t_d(n, k, 2^{\Theta(\log \sigma + \log(u/n))})) \) time for the sender side and deterministic time \( O(n + t_d(n, k, 2^{\Theta(\log \sigma + \log(u/n))})) \) for the receiver side (the time becomes randomized if \( t_d \) is randomized).

The reduction is optimal (up to constant factors), since, on the one hand, even if player \( A \) knows player \( C \)'s string, he needs to send the positions of the mismatches and the mismatching characters using \( \Omega(k(\log(u/k) + \log \sigma)) \) bits, and on the other hand the protocol sends \( \Theta(\log u) = 2^{\Theta(\log \sigma + \log(u/n))} \leq \Theta(k(\log(u/k)) \) bits in addition to the redundancies of the error correcting codes which can be \( \Theta(k(\log(u/k) + \log \sigma)) \) bits in the best case. The latter follows from the existence of error correcting codes that achieve optimal redundancy \( \Theta(k(\log(n/k) + \log \sigma')) = \Theta(k(\log(n/k) + \log(u/n) + \log \sigma')) \), in particular those attaining the Gilbert-Varshamov bound \([13, 27]\). Unfortunately, we do not know whether such codes admit \( O(n \cdot \text{polylog}(n)) \) decoding time for all values of \( k \) and \( n \).

Footnote: The main reason for the modification is that we will now need to solve equations over the finite field, which will require efficient support for inversions and square root operations. These operations can be supported using lookup tables of \( O(P \log P) \) and \( O(q \log q) \) bits respectively on \( F_P \) and \( F_q^2 \). Since \( |u| \) can be as big as \( n^{3/2} \), the lookup tables of \( F_P \) will occupy \( O(n^{3/2} \log n) \) bits while those for \( F_q^2 \) will occupy only \( o(n) \) bits of space and can be computed in \( o(n) \) additional time.
n, and thus we can not use Theorem 9 to state a message-size optimal solution that would improve on Theorem 7.

5 Concluding remarks

It was implicit that the numbers $\log \sigma$, $\log u$ and $\log n$ are known to the algorithm. If that was not the case, then we need to have an implementation of the MSB operation. The operation can be simulated in constant time, using a constant number of standard instructions (including multiplications) \[^5\]. Our reduction will also need to efficiently simulate divisions by $u$ and $n$. Division by $u$ is used in Lemma 2 and division by $n$ is used in Lemma 5. If both $u$ and $n$ are powers of two, then the division is just a right shift. Otherwise, the division (and modulo) can be simulated using the multiplication by the inverse. The latter can be precomputed, respectively, in $O(\log \log u)$ and $O(\log \log n)$ time (using the Newton-Raphson method). The term $O(\log \log n)$ is clearly within our time bound, but this is not necessarily the case for the term $O(\log \log u)$. We can avoid the problem by rounding $u$ to the nearest power of two immediately above (using the MSB operation). This will only increase the final size of the message by $O(k)$ bits.

In the proof of Theorem 8 we can replace the polynomial from Lemma 3 by a 3-wise independent permutation, computed using the Möbius transform \[^3\] over the projective line $\mathbb{F} \cup \{\infty\}$, where $\mathbb{F}$ is any finite field. The projection is given by the formula $y = \frac{ax + b}{cx + d}$, where $a, b, c, d$ are constants such that $ad - bc = 1$. Its reverse can be computed through the formula $x = \frac{dy - b - cy}{a}$. We can thus use $\mathbb{F} = \mathbb{F}_q^2$ and simulate the transform and its inverse in constant time (in particular simulating division in constant time over $\mathbb{F}_q^2$ by using operations on $\mathbb{F}_q$). Then in a cell, we will only need to store a value $f_2(x)$ instead of pair $(I, f_2(x))$, since we can now recover $x$ directly from $f(x) = f_2(x) \cdot n + f_1(x)$.

Our modification of the two-level hashing scheme of \[^11\] may be of independent interest. It uses only $O(\log u)$ bits of randomness and only in the first level. This is not the first result of the kind. In \[^7\], a variant of the scheme of \[^11\] is presented which uses only $O(\log n + \log \log u)$ bits of randomness. We could use such a scheme as an alternative in our reduction. The sender would build the structure, and then send the random bits that he uses. The main drawback of that approach is that the second level is randomized too, in particular, the construction time of the functions of the buckets depends on the keys mapped to the bucket (and not only their number as in our case). As a consequence, there is no guarantee that the receiver will spend a total deterministic linear time for building the second level (the time is only expected linear). We could also have slightly simplified our scheme by only implementing set reconciliation with alphabet $\{0, 1\}$ and simulating any alphabet \[^9\]. The drawback of doing so is an increase in the message size, since now the log $u$ term is replaced by $\log u + \log \sigma$ in the description of the global hash function and in the redundancy associated with array $B$. A future research direction is to reduce the constants associated with our scheme and to make the reduction run in deterministic time for the sender side.
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