Error-Correcting WOM Codes: Concatenation and Joint Design

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Abstract—We construct error-correcting WOM (write-once memory) codes that guarantee correction of any specified number of errors in $q$-level memories. The constructions use suitably designed short $q$-ary WOM codes and concatenate them with outer error-correcting codes over different alphabets, using suitably designed mappings. With a new storage-efficiency measure we call EC-rate, we show that for common error types the codes save redundancy and implementation complexity over straightforward concatenation. In addition to constructions for guaranteed error correction, we extend the error-correcting WOM scheme to binary multi-level coding for random errors, and toward soft-decision decoding provide an efficient way to extract reliability information without using higher-precision readout.

Index Terms—WOM codes, error-correcting WOM codes, codes for memories, concatenated codes, multi-level coding, coded modulation.

I. INTRODUCTION

Flash-based non-volatile memories (NVM) are the storage media of choice in most modern information applications, thanks to their fast access and growing densities. However, the Flash technology suffers from the major impediment of not being able to update data in-place. Because removing charges from memory cells cannot be done at a fine granularity, it is not possible to update written data without first erasing a very large data unit. As a result, write performance is degraded and the wear of cells is accelerated. It has been demonstrated and recognized [20], [24] that WOM (write-once memory) codes [21] hold promise to mitigate this access limitation by allowing to update the logical data multiple times without need to physically remove charges from the cells. While WOM codes can support a more flexible write access to Flash, a concern is raised about their effect on data reliability. With WOM codes, cells are written multiple times between erases, and are accessed in a less predictable order than when written only once without WOM. These two effects may increase the severity of inter-cell interference, and degrade reliability if not properly addressed.

Our objective in this paper is to improve the reliability of WOM codes by making them resilient to errors from noise and interference. Our method is to combine WOM codes with error-correcting (EC) codes to get guaranteed error correction with flexible parameters. In general combining WOM codes with EC codes is a non-trivial task because the EC encoder may not respect the WOM constraints, and the WOM decoder may cause error propagation affecting a large number of EC symbols. While there have been some works addressing EC+WOM combination [26], [23], [17], the problem is yet to be adequately solved for practical deployment in non-volatile storage. In particular, prior to our work no scheme provides guaranteed correction of $\tau$ errors, for an arbitrary $\tau$. One exception is a construction in [23] that lifts triple-error correcting WOM codes to general $\tau$, but due to its use of codeword replication has sum-rates degrading as the reciprocal of $\tau$. In comparison, the sum-rates of our constructions degrade much slower: linearly with $\tau$. A combination of EC and WOM codes is also the subject of [16], however, in that scheme the EC code is used as a building block for the WOM codes’ re-write capabilities. The results of [16] also include the basic EC-WOM concatenation we use as a baseline to compare to our constructions.

Our contribution in this paper is a new scheme to combine WOM with EC codes. We use short $(n=2)$ WOM codes over the memory’s $q$-ary alphabet ($q$ is the number of levels supported by the physical device), and concatenate them with outer EC codes of any length $N$. Our contributions divide into two parts: the first part (Section IV) addresses code constructions for guaranteed error correction with worst-case errors, while the second part (Section V) develops additional tools for random error channels. In the first part we propose multiple code constructions for guaranteed $\tau$-error correction (for any $\tau$), which improve storage rates over straightforward concatenation for errors common in Flash (e.g. magnitude-1 or asymmetric magnitude-1). Our construction method is to jointly design the EC and WOM codes, such that the WOM redundancy expended for rewriting is designed to also contribute to error correction. By that, our work develops concrete formal constructions building on the notion that WOM codes can have intrinsic error resilience, previously observed in [19] and further developed in [14]. The main observation leading to the constructions’ good correction capabilities is that the WOM alphabet can be decomposed and mapped to multiple smaller alphabets, and on each alphabet different EC capabilities can be prescribed for the outer EC code. Each of the constructions is specified with four components: 1) the WOM code, 2) the alphabet decomposition and mapping, 3) the EC code, and 4) the construction combining 1-3. To evaluate and compare our constructions, we define the EC-rate as a new rate measure that captures the residual redundancy added to a WOM code for error correction. Code families with concrete parameters are obtained by evaluating the EC-rate on parameters of the ubiquitously used BCH codes. In addition to concrete advantage in EC-rates, the lower-alphabet EC codes used in our constructions reduce the implementation complex-
ity over straightforward concatenation. In the second part we complement the code constructions with additional coding-theoretic tools that can improve error-correction performance with random errors. Firstly, we develop a multi-level coding framework for which WOM codes are constructed to work with binary-only EC codes (for low implementation complexity). The WOM alphabet is mapped hierarchically to multiple bits, and information from the WOM decoding function is used to feed an error+erasure multi-stage decoder. Analysis using decoding-error probability bound shows the good performance of the WOM codes constructed especially for multi-level coding. In addition, it shows the significant advantage in decoding performance of decoders that know the current write number of the WOM code. Secondly, toward extending the EC-WOM framework to soft-decision decoding, we show how the soft reliability inputs can come not from costly higher-precision readout, but from the reliability information provided by the inner WOM decoder. We show that for additive Gaussian noise channels the ordering of symbol reliabilities based on the WOM decoding function has a nice structure.

Our constructions of EC-WOM codes provide code families that are infinite in two ways: we can grow the block length in return to more information or larger τ, and we can grow the alphabet size in return to more writes. The constructions use different inner WOM codes with \( n = 2 \), which we found to offer a good balance between error controllability and re-write efficiency. We believe that our approach can be extended to using longer inner \( q \)-ary codes, such as those in [8], [2], [11], [12], [18].

II. Preliminaries

We start by including definitions regarding WOM codes; then we add definitions addressing error correction by WOM codes.

A. WOM codes

A WOM code is our basic object of study in this paper, and we focus on the generalization of the WOM coding model to \( q \)-ary alphabets, with \( q > 2 \). We give the general definition of a WOM code up to restricting it to be fixed-rate, that is, the same number of bits is written in each of its writes.

Definition 1. A \((\ell, q, t, B)\) (fixed-rate) WOM code is a code applied to a size \( \ell \) block of \( q \)-ary cells, and guaranteeing \( t \) writes of input size \( B \) each, while the cell levels do not decrease.

A WOM code is specified through a pair of functions: the decoding and update functions.

Definition 2. The decoding function is defined as \( \psi : \{0, \ldots, q-1\}^\ell \rightarrow \{0, \ldots, B-1\} \), which maps the current levels of the \( \ell \) cells to the data input in the most recent write. The update function is defined as \( \mu : \{0, \ldots, q-1\}^\ell \times \{0, \ldots, B-1\} \rightarrow \{0, \ldots, q-1\}^\ell \), which specifies the new cell levels as a function of the current cell levels and the new data value at the input. By the WOM requirement, the \( i \)-th cell level output by \( \mu \) cannot be lower than the \( i \)-th cell level in the input, for each \( i \).

Definition 2 of the WOM-code functions assumes the special case that the function outputs do not depend on the current write number. Later in the paper (Definition 10), we generalize the definition of the decoding function \( \psi \) to also depend on the write number.

Definition 3. The code’s physical state is defined as the \( \ell q \)-ary levels to which the cells are currently programmed. The code’s logical state is the data element from \( \{0, \ldots, B-1\} \) returned by \( \psi \) on the current physical state.

A convenient way to specify the WOM decoding function is through an \( \ell \)-dimensional matrix of physical states \( \{0, \ldots, q-1\}^\ell \), where at each position the matrix holds the logical state in \( \{0, \ldots, B-1\} \) corresponding to the output of the decoding function for that physical state. The update function specifies the target physical state given the current physical state and the input logical value.

Example 1. Let us consider two WOM-code examples. In Figure 1a we have the decoding function of a \((\ell = 2, q = 7, t = 3, B = 8)\) WOM code constructed by Construction 3 in [5]. This code is applied on a pair of \( q = 7 \)-level memory cells, enabling \( t = 3 \) guaranteed writes of size \( B = 8 \) \((\log_2 B = 3 \text{ bits})\) each. In Figure 1b we have a different \((\ell = 2, q = 7, t = 3, B = 8)\) code, offering the exact same parameters. Let us assume that for the \( t = 3 \) writes we want to write the logical values 7, 6 and 2 using the WOM code of Figure 1a. For the first write the logical state is 7 and the physical state is \((2, 1)\). When updating the logical state to 6, the physical state becomes \((2, 4)\). For the third write of 2, the physical state becomes \((2, 6)\). The codes of Figure 1a,b are equivalent in terms of their specified re-write performance (the same guaranteed \( t \) and \( B \) using the same coding resources \( \ell \) and \( q \)). However, the codes may differ significantly in performance beyond the specified parameters. For example, the code of Figure 1a can be extended to 4 writes when moving from \( q = 7 \) to \( q = 8 \), while the code of Figure 1b cannot.

In the remainder of the paper, we focus on WOM constructions that maximize the error-correction performance within their

![Figure 1: Sample \( \ell = 2 \) WOM constructions (from [5]). (a) Decoding function \( \psi \) for a \((2, 7, 3, 8)\) code. (b) Decoding function \( \psi \) for another \((2, 7, 3, 8)\) code. Physical states are represented by \((c_1, c_2)\) and logical states are labeled inside each matrix position.](image-url)
class of \((\ell, q, t, B)\) codes. The error-correction performance of a WOM code is now defined.

**Definition 4.** A WOM code with parameters \((\ell, q, t, B)\) is called \(\tau\)-error correcting if it can correct any error combination in up to \(\tau\) (out of the \(\ell\)) \(q\)-ary code symbols.

### III. EC-WOM Concatenation

The route to error-correcting WOM codes we pursue in this paper is through concatenation. We call such codes EC-WOM codes. Let \(W\) be a \((n, q, t, M)\) WOM code. Denote an EC code \(C\) by \(C_A[N, N - r]\) if it is defined over an alphabet of size \(A\) (power of a prime), it has length \(N\), and \(r\) redundant \(A\)-ary symbols. A pictorial illustration of concatenating the EC code \(C\) with the WOM code \(W\) is given in Figure 2.

![EC-WOM concatenation](image)

**Figure 2:** EC-WOM concatenation. Note that \(u_i, v_j \in GF(A)\), whereas \(w_i \in \{0, ..., q - 1\}^n\)

We start with the simplest construction using an inner WOM code and an outer EC code over the same alphabet as the WOM-code input [16].

**Construction 1.** Given a \((n, q, t, M)\) WOM code \(W\) and a \(C_M[N, N - r]\) EC code \(C\), the concatenated code is the \((nN, q, t, M^{N - r})\) WOM code obtained by taking \(N\) copies of \(W\) and inputting to each a symbol of \(C\). In addition, if \(C\) can correct \(\tau\) \(M\)-ary errors, then the concatenated code is a \(\tau\)-error correcting WOM code.

The concatenation of Construction 1 is extremely simple, but it requires to use an EC code over an alphabet as large as the input size of the WOM code. In subsequent sections we generalize to constructions that concatenate outer EC codes with alphabets smaller than \(M\), which will be shown to offer savings in both redundancy and complexity over Construction 1, for error models that are not \(q\)-ary symmetric.

**A. EC codes and BCH redundancy**

We now define a measure of redundancy that is suitable for evaluating the concatenated codes. To compare between different EC-WOM concatenated constructions we use the EC-rate of the code, defined next.

**Definition 5.** Let \(CW\) be a \((nN, q, t, M^{N - r_W})\) WOM code with some error-correction capability, which is constructed with a \((n, q, t, M)\) inner WOM code. We define the **EC-rate** of \(CW\) to be

\[
1 - \frac{r_{\text{eff}}}{N}.
\]

\(r_{\text{eff}}\) is the effective redundancy of the EC-code in units of \(M\)-ary symbols.

Note that the usual WOM sum-rate of the same code is \((t(N - \log_q M)/n)\), which equals the EC-rate multiplied by the factor \((t \log_q M)/n\) depending only on the (non error-correcting) WOM code used in the concatenation. Thus for notational compactness we use the EC-rate in the sequel to compare EC-WOM constructions that concatenate WOM codes with the same parameters \((n, q, t, M)\).

The constructions we provide in the paper work with general EC codes, but for convenience we present them with BCH EC codes [3] [15]. Beyond their well-known effectiveness in a variety of applications (including non-volatile memories), BCH codes are especially useful for this study because they can be defined over different alphabet sizes, and there are simple expressions for their redundancy that can capture and compare their performance.

**Definition 6.** For alphabet size \(A\), code length \(N\), and designed minimum distance \(d\), define the **BCH (approximate) redundancy** as

\[
\frac{A - 1}{A} \cdot (d - 2) \cdot \log_A N \quad [\text{\(A\)-ary symbols}].
\]

The BCH redundancy is the approximate number of parity symbols in the codeword, given in units of \(A\)-ary symbols. This definition of redundancy comes from a known approximation of the redundancy of primitive \(A\)-ary BCH codes for large \(A\), when \(A\) is a power of a prime [25]. We can see that fixing \(N\) and \(d\) the BCH redundancy as defined decreases with \(A\) as \((A - 1)/(A \log A)\), but note that when fixing the units (e.g. to bits) the amount of actual redundancy in the code grows with \(A\) as \((A - 1)/A\).

When using Construction 1 with a BCH code we get the following result combining Definitions 6 and 5.

**Example 2.** Take a \((n = 2, q = 8, t = 4, M = 8)\) WOM code \(W\) and concatenate it with an outer \(\tau\)-error correcting (design minimum distance \(d = 2\tau + 1\)) BCH code \(C = C_M[N, N - r]\) using Construction 1. Then we get a \((2N, 8, 4, 8^{N - \tau})\) WOM code that is \(\tau\)-error correcting.

The outer code \(C\) used in the construction of Example 2 has BCH redundancy

\[
\frac{7}{8} \cdot (2\tau - 1) \cdot \log_8 N \quad [\text{\(8\)-ary symbols}].
\]

**Proposition 1.** A \(\tau\)-error correcting WOM code obtained by Construction 1 with an \(M\)-ary BCH code \(C\) has EC-rate

\[
1 - \frac{M - 1}{M} \cdot \left(2\tau - 1\right) \cdot \log_M N,
\]

and one may recognize the BCH redundancy of the \(M\)-ary BCH code as \(r_{\text{eff}}\) at the numerator of the right term in (3). Specifically for the parameters of Example 2 we get that Construction 1 attains the EC-rate of

\[
1 - \frac{7}{8} \cdot \left(2\tau - 1\right) \cdot \log_8 N.
\]

Equation (3) reveals the advantage of the concatenation-based constructions of this paper: offering WOM sum-rates that scale with the number of errors as \(\rho(1 - \beta \tau)\) (\(\rho, \beta\) some constants independent of \(\tau\), whereas the sum-rates of the best prior constructions [23] decay to zero much faster, as \(\rho'/\tau\), due to their use of codeword replication.
B. Construction parameters

In the rest of the paper we present concatenated constructions using inner WOM codes, and we design these inner codes to provide error correction in addition to re-writing capabilities. For ease of presentation, our constructions are given with inner codes of specific WOM parameters: \((n = 2, q = 8, t = 4, M = 8)\), but we explain how each construction can be extended to an infinite family of codes with any \(q\) and \(t = \lfloor 4(q-1)/7 \rfloor\). The advantage of this parameter family is that \(M = 8\) is an integer power of \(2\) (thus offering easy mapping of information bits), and \(n = 2\) is small enough to control error propagation and large enough to get good re-write efficiency (for enough to control error propagation and large enough to get easy mapping of information bits), and

IV. EC-WOM Concatenated Constructions

In this section we construct concatenated EC-WOM codes that improve over the basic Construction 1 from the previous section. Construction 1 offers EC-WOM codes by using standard \(M\)-ary error-correcting codes for symmetric errors. The main drawback of this construction is that symmetric error correction of the \(M\)-ary information symbols is costly in redundancy and complexity, while not necessary when the physical errors are restricted in magnitude and/or direction. In the constructions of this section we show that for the error types common in non-volatile memories, guaranteed error correction can be attained with less redundancy and lower decoding complexity. In particular, the error-correcting codes used in the concatenation will be over alphabets smaller than \(M\), and often the binary alphabet. The key method is to carefully design the inner WOM code such that guaranteed error correction is attained with a more efficient outer error-correcting code.

The error model considered in Definition 4 covers any possible error affecting a code symbol. In non-volatile memories with relatively large number of levels \(q\) (e.g. \(q = 8\)), the dominant errors are often more restricted in magnitude and/or direction [6]. Instead, specific dominant error types are addressed, which are more limited in the symbol transitions they cover. We follow the same wisdom here, and refine the error-correction characterization for EC-WOM codes. We consider the following error types defined over the alphabet \(\{0, \ldots, q-1\}\).

Definition 7. Let \(c \in \{0, \ldots, q-1\}\) be the symbol written to a memory cell. The cell suffers a mag-1 error if the read symbol is \(c' \in \{0, \ldots, q-1\}\) such that \(|c' - c| = 1\).

Mag-1 errors allow transitions of one level upward or downward from the correct symbol. We similarly define the asymmetric version Amag-1.

Definition 8. Let \(c \in \{0, \ldots, q-1\}\) be the symbol written to a memory cell. The cell suffers an Amag-1 error if the read symbol is \(c' \in \{0, \ldots, q-1\}\) such that \(|c' - c| = 1\).

Amag-1 errors allow transitions of one level in the upward direction only. Amag-1 errors are common in non-volatile memories such as Flash due to inter-cell interference (ICI): adding charges to a cell due to writing to a neighboring cell.

We can plug in the mag-1 and Amag-1 error types into the definition of \(\tau\)-error correction (Definition 4) as follows.

Definition 9. A WOM code with parameters \((t, q, s, B)\) is called \(\tau\) mag-1 error correcting if it can correct any combination of mag-1 errors, respectively, up to \(\tau\) (out of the \(t\) \(q\)-ary code symbols).

Note that in particular a \(\tau\) mag-1 error correcting code is also \(\tau\) Amag-1 error correcting, and a \(\tau\)-error correcting code (from Definition 4) is both \(\tau\) mag-1 error correcting and \(\tau\) Amag-1 error correcting.

To efficiently correct mag-1 and Amag-1 errors in WOM codes, in the sequel we specify the constructions in three steps: 1) specification of the inner WOM code, 2) mapping the \(M\)-ary logical alphabet of the WOM code \(\{0, \ldots, M-1\}\) to a more structured alphabet, and 3) prescribing EC-codes over the structured alphabet with parameters that guarantee correcting \(\tau\) errors. This construction method has the flavor of generalized concatenation (GC) and coded modulation (CM) [4] found very useful in previous applications\(^1\). Unlike the simple Construction 1, in the improved constructions that follow we will need to go beyond specifying just the parameters of the inner WOM code, to designing new WOM codes with extra properties that are needed for the concatenation to work.

A. EC-WOM construction for mag-1 errors

We present the construction for the specific parameters \(q = 8\) and \(t = 4\), but the same construction can be extended to any \(q\) and \(t = \lfloor 4(q-1)/7 \rfloor\), from the periodicity of the tiling construction (proved in [5]), and of the mapping below.

1) Tiling WOM code: The inner WOM code we use in this construction is the known tiling-based \((n = 2, q = 8, t = 4, M = 8)\) code from [5], depicted in Figure 3. The figure shows the decoding function of the code, and we omit here the specification of the update function because in this case it is immaterial for proving the error-correction properties (this is not true in general, and in subsequent constructions we will rely on the update function for error correction).

2) Mapping 8-ary to a product of 4-ary and binary: The concatenation requires introducing structure to the logical \(M = 8\)-ary alphabet as follows. We map each number \(s \in \{0, \ldots, 7\}\) to its mixed radix representation \((\bar{s}, s)\), where the upper symbol \(\bar{s} \in \{0, 1, 2, 3\}\) and the lower symbol \(s \in \{0, 1\}\) satisfy \(2\bar{s} + s = s\), and are unique. In Figure 4 we apply this mapping to the tiling WOM code of Figure 3, where for clarity we map \(\{0, 1, 2, 3\}\) to \(\{a, b, c, d\}\). It can be seen in Figure 4b that tiling the space with the tile of Figure 4a gives the property that physical states adjacent horizontally or vertically have the opposite binary value in their lower symbol. This property will be used by the construction to combat mag-1 errors, which correspond to horizontal and vertical transitions between adjacent physical states.

\(^1\)GC fits when seeing the WOM as a code, and CM is appropriate when thinking about WOM as modulation.
Fig. 3: Decoding function of the tiling-based \( \mathcal{W}(2, 8, 4, 8) \) code taken from [5].

Fig. 4: Mapping the 8-ary WOM logical symbols to 4-ary+binary symbols. (a): a single tile, and (b): the entire WOM code.

3) Outer 4-ary and binary EC codes: For the 4-ary upper symbols we take the \( C_4[N, N - r_1] \) BCH code with design distance \( \tau + 1 \); for the binary lower symbols we take the \( C_2[N, N - r_2] \) BCH code with design distance \( 2\tau + 1 \). These choices will allow us to correct up to \( \tau \) errors in the binary lower symbols and up to \( e' \) errors and \( e \) erasures in the 4-ary upper symbols, for \( 2e' + e \leq \tau \).

4) The construction:

Construction 2. Given the \( (2, 8, 4, 8) \) WOM code \( \mathcal{W} \) specified with its logical mapping in Fig. 4b and EC codes \( C_4[N, N - r_1], C_2[N, N - r_2] \) specified in Section IV-A.3, the concatenated code is the \( (2N, 8, 4, 4^{N-r_1} \cdot 2^{N-r_2}) \) WOM code obtained by taking \( N \) copies of \( \mathcal{W} \) and inputting to each a symbol of \( C_4 \) to its upper symbol and a symbol of \( C_2 \) to its lower symbol.

The essential properties of Construction 2 are given in the following proposition. The proof includes the decoding algorithm.

Proposition 2. For any \( N \) and \( \tau \), a \( (2N, 8, 4, 4^{N-r_1} \cdot 2^{N-r_2}) \) code obtained by Construction 2 is \( \tau \) mag-1 error correcting. Moreover, the code has EC-rate

\[
1 - \left[ \frac{3}{4}(\tau - 1) + \frac{1}{2}(2\tau - 1) \right] \log_8 N. \tag{5}
\]

Proof: To see that the code from Construction 2 is \( \tau \) mag-1 error correcting, define by \( \tau_1 \) the number of copies of \( \mathcal{W} \) that suffered a mag-1 error in exactly one of their two symbols. Similarly, define by \( \tau_2 \) the number of copies of \( \mathcal{W} \) that suffered mag-1 errors in both their symbols. We show that any combination with \( \tau_1 + 2\tau_2 \leq \tau \) is correctable. We first decode the binary code \( C_2 \). From the mapping seen in Figure 4, there will be bit errors in all positions corresponding to the \( \tau_1 \) mag-1 errors affecting one symbol out of a \( \mathcal{W} \) pair. Since \( \tau_1 \leq \tau \), the decoder of \( C_2 \) will locate and correct all these errors. Next we decode the 4-ary code \( C_4 \), where every position with error in \( C_2 \) is erased in the input of \( C_4 \)’s decoder. The decoder sees \( \tau_1 \) erasures and \( \tau_2 \) errors, and can correct them because of the restriction \( \tau_1 + 2\tau_2 \leq \tau \) (recall that \( C_4 \) has design minimum distance \( \tau + 1 \)). At this point all up to \( \tau \) symbols in error are recovered by \( C_2 \) and \( C_4 \) jointly.

To prove the EC-rate, we observe that

\[
4^{N-r_1} \cdot 2^{N-r_2} = 8^{N-r_1-r_2} = 8^{N-r_1-\frac{3}{4}r_2}. \tag{6}
\]

Hence in Definition 5 we have \( r_{\text{eff}} = \frac{3}{4}r_1 + \frac{1}{4}r_2 \). Substituting in \( r_{\text{eff}} \) the BCH redundancies from (2)

\[
r_1 = \frac{3}{4}(\tau - 1) \log_4 N, \quad r_2 = \frac{1}{2}(2\tau - 1) \log_2 N,
\]

we get (5) for the EC-rate. Comparing to Construction 1, we see that Construction 2 gives EC-rate higher by \( \frac{1}{4} \log_2 N \). This advantage is possible for the model of mag-1 errors, in which case Construction 2 is able to reduce the EC-code alphabet, and thanks to (2) saves redundancy for the same number of errors. Another important advantage of Construction 2 is in its much simpler decoding. While Construction 1 requires a BCH decoder for \( \tau \) errors over the finite field \( \text{GF}(8) \), Construction 2 only needs to correct \( \tau \) binary errors in \( C_2 \), and in the worst case only \( \tau/2 \) errors over \( \text{GF}(4) \) in \( C_4 \). Moreover, in typical decoding instances, the code \( C_4 \) will mostly need to deal with erasures, because in most error patterns very few copies of \( \mathcal{W} \) suffer errors in both symbols.

B. Improved EC-WOM construction for mag-1 errors

While Construction 2 offered an improvement of the EC-rate over the basic concatenation, the improvement is quite small, and in particular the difference \( \frac{1}{4} \log_2 N \) does not grow with \( \tau \). In this sub-section we aim at improving the EC-rate further, and more significantly. To get the desired improvement in EC-rate we need two ingredients: 1) a new suitably designed \( (2, 8, 4, 8) \) WOM code, and 2) a slight refinement of the error
model to distinguish between mag-1 errors not in the same \( W \) pair (more common) and those falling in the same pair (less common). Also, from now on we will rely on the fact that the decoder knows the current write count, that is, how many times (out of the \( t \)) the WOM has been written so far. This assumption is standard in the WOM literature, and does not incur redundancy or complexity costs when \( N \) is at least moderately large (storing the write counter is amortized over many parallel WOM blocks). The following construction can be extended from \( q = 8, t = 4 \) to any \( q, t = 4\lfloor(q - 1)/7\rfloor \) by stacking more copies of the WOM code (Figure 5) diagonally with period 7. Extending the code properties below to additional parameters is possible, but for terseness we omit these extensions here.

1) 3-Manhattan WOM: In order to improve the EC-rate, we now show a more suitably designed \((n = 2, q = 8, t = 4, M = 8)\) WOM code. The decoding function of the WOM code is shown in Figure 5. The physical states are colored according to which write(s) reach them: magenta, green, red, blue for writes 1-4, respectively. The physical states with mixed colors are shared between two adjacent writes; for example physical state \((0, 7)\) is shared between writes 3 and 4. With the colors of the states provided, the update function can be specified in a straightforward manner. It can be verified that from any physical state in one write’s color it is possible to reach any logical state in the next write’s color. This property of restricting the physical state to belong to the corresponding write’s color will be useful for later proving the error-correction properties of the construction.

2) Mapping 8-ary to a product of 4-ary and binary: We use an 8-ary to a product of 4-ary and binary mapping, like in Construction 2 (Section IV-A.2) to map the 8-ary alphabet to a product of 4-ary and binary alphabets. This mapping gives the WOM the following property: two physical states in the same write that have the same 4-ary symbol and opposite binary bit are at Manhattan distance at least 3, where the Manhattan distance between two dimension-\( n \) vectors \( x, y \) is \( \sum_{i=1}^{n} |x_i - y_i| \). This property can be seen in Figure 6 showing the WOM code with the product alphabet. For example, in write 2 physical state \((3, 1)\) with value \( \text{b0} \) and physical state \((5, 0)\) with value \( \text{b1} \) are at Manhattan distance 3, and no \( \text{b1} \) state is closer to \((3, 1)\) in write 2 (there is a closer \( \text{b1} \) state in \((1, 1)\), but it does not belong to write 2).

3) Outer 4-ary and binary EC codes: For the 4-ary symbols we take the \( C_4[N, N - r] \) BCH code with design distance \( 2r + 1 \); the binary bits are left uncoded (i.e., we use the trivial code \( C_2[N, N] \)). These choices will allow us to correct up to \( \tau \) errors in the 4-ary symbols and recover the binary bit.

4) The construction:

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{figure5.png}
\caption{A more suitably designed WOM code: \( \mathcal{W}_2(2.8, 4.8) \). The numbers as usual specify the decoding function and the colors represent at which write(s) the physical state can be reached. Solid color represents a single write and mixed color represents two consecutive writes.}
\end{figure}

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{figure6.png}
\caption{The code of Figure 5 after mapping 8-ary to 4-ary+binary. The code+mapping have the 3-Manhattan property used for error correction.}
\end{figure}

\begin{construction}
Given the \((2,8,4,8)\) WOM code \( \mathcal{W}_2 \) specified with its logical mapping in Figure 6 and EC codes \( C_4[N, N - r] \), \( C_2[N, N] \) specified in Section IV-B, the concatenated code is the \((2N, 8, 4^{N-r} \cdot 2^N)\) WOM code obtained by taking \( N \) copies of \( \mathcal{W}_2 \) and inputting to each a symbol of \( C_4 \) to its upper symbol and a symbol of \( C_2 \) to its lower symbol.

The concatenated EC-WOM code specified in Construction 3 has the following properties. The proof includes the decoding algorithm.

\begin{proposition}
For any \( N \) and \( \tau \), a \((2N, 8, 4, 4^{N-r} \cdot 2^N)\) the code obtained by Construction 3 is \( \tau \) mag-1 error correcting.
\end{proposition}
assuming no copy of $W_2$ suffers two mag-1 errors. Moreover, the code has EC-rate

$$1 - \frac{3}{4} \cdot \frac{(2\tau - 1) \cdot \log_8 N}{N}. \quad (7)$$

**Proof:** Suppose $\tau$ copies of $W_2$ suffered a mag-1 error in exactly one of their two cells. Assuming the decoder knows the current write number, a mag-1 error resulting in a physical state outside this write’s color can be detected by the decoder, and mapped to a 4-ary erasure in $C_4$. Other mag-1 errors map to 4-ary errors in $C_4$, because adjacent physical states within the same write’s color have different 4-ary symbols (check in Figure 6). Thus all $\tau$ 4-ary symbols in the erroneous cells can be recovered since $C_4$ is designed with distance $2\tau + 1$. To recover the lower bits of the cells with mag-1 errors we use the Manhattan-distance-3 property shown in Section IV-B.2 that guarantees that not both $x0$ and $x1$ states are adjacent to the read cell state, where $x$ is the 4-ary symbol recovered by $C_4$.

To prove the EC-rate, we note that $4^{N-r} \cdot 2N = 8^{N-2\tau}$, hence in Definition 5 we need to substitute $r_{\text{eff}} = \frac{3}{4} \tau$. We calculate the BCH redundancy $r$ for $C_4$ from Definition 6 and get $r = \frac{3}{4} (2\tau - 1) \log_4 N$. Substituting into $r_{\text{eff}}$ we obtain (7).

Examining the EC-rate attained for the improved Construction 3, we see a significant improvement over both Construction 1 and Construction 2. The new EC-rate is higher than Construction 1 by $\frac{3}{4} \cdot \frac{(2\tau - 1) \cdot \log_8 N}{N}$, now an improvement that grows with $\tau$. This advantage comes at the cost of excluding error patterns with both cells in error at the same copy of $W_2$ (most errors of this type are still correctable, but not all; see Figure 6 c1 in state (3,3) changing to d1 in state (4,4); the decoder will miscorrect to c0 in state (5,4)). The uncorrectable error combinations are unlikely in a random error pattern, because when $\tau \ll N$ hitting two cells of the same $W_2$ copy has a very small probability. That said, we next extend Construction 3 to also guarantee a specified number of double mag-1 errors in the same $W_2$ copy. The extension is done in 3’ and 4’ below by adding an EC-code to the binary symbols as well, instead of the trivial $C_2[N,N]$ used previously in Section IV-B.3.

3’) **Refined outer 4-ary and binary EC codes:** For correcting $\tau_1$ single mag-1 errors (one in a $W_2$ copy) and $\tau_2$ double mag-1 errors (both in the $W_2$ copy), use for the 4-ary upper symbols the code $C_4[N,N-r_{\text{1}}]$ with design distance $2(\tau_1 + \tau_2) + 1$. For the lower binary symbols use the code $C_2[N,N-r_{\text{2}}]$ designed with distance $2\tau_2 + 1$.

4’) **The refined construction:**

**Construction 4.** This construction is the same as Construction 3, but with its EC-codes changed to $C_4[N,N-r_{\text{1}}], C_2[N,N-r_{\text{2}}]$ specified in Section IV-B.3’.

Now with the refined EC-codes we have the following correction capabilities. The proof includes the decoding algorithm modified accordingly.

**Proposition 4.** For any $N$ and $\tau_1, \tau_2$, a $(2N, 8, 4, 4^{N-r_{\text{1}}}, 2^{N-r_{\text{2}}})$ the code obtained by Construction 4 corrects any error combination where at most $\tau_2$ $W_2$ copies suffered double mag-1 errors and at most $\tau_1 + \tau_2$ $W_2$ copies suffered mag-1 errors (double or single). Moreover, the code has EC-rate

$$1 - \frac{3}{4} \cdot \frac{(2\tau_1 + \frac{10}{3} \tau_2 - \frac{5}{3}) \cdot \log_8 N}{N}. \quad (8)$$

**Proof:** As in Proposition 3, $C_4$ will correct all the 4-ary symbols in $W_2$ copies that suffered mag-1 errors, single or double. After setting the bits in the binary symbols according to the closest physical states to the read outputs, the decoder of $C_2$ will be able to correct the at most $\tau_2$ bit errors from double-error $W_2$ copies.

To prove the EC-rate, we have $r_{\text{eff}} = \frac{2}{3} \tau_1 + \frac{1}{4} \tau_2$, and substituting the BCH redundancy for $\tau_1 + \tau_2$ in $C_4$ and $\tau_2$ in $C_2$ we get

$$r_1 = \frac{3}{4} (2\tau_1 + 2\tau_2 - 1) \log_4 N, \quad r_2 = \frac{1}{2} (2\tau_2 - 1) \log_2 N.$$

From this and some rearrangement (8) follows.

Construction 4 offers the nice compromise of correcting a number of double errors needed for good error coverage, while keeping the EC-rate well above that of Construction 2 when this number is not too high. Furthermore, it can be shown that the refined Construction 3 can correct $\tau_2$ arbitrary errors in $W_2$ copies, not just double mag-1 errors.

In an effort to further increase the EC-rate and reduce complexity, we next construct EC-WOM codes that correct the weaker but similarly motivated Amag-1 errors.

**C. EC-WOM construction for Amag-1 errors**

In the following construction we endow a WOM code with guaranteed error correction while *exclusively using binary codes* in the concatenation. This is a major implementation advantage in practice, as binary codes (in particular BCH codes) are much easier to implement. We present the construction for the specific parameters $q = 8$ and $t = 4$, but the same construction can be extended to any $q$ and $t = \lfloor 4(q - 1)/7 \rfloor$, from the periodicity of the tiling construction and of the mapping below.

1) **Tiling WOM code:** The inner WOM code we use in this construction is the known tiling-based $(n = 2, q = 8, t = 4, M = 8)$ code from [5], same as in Section IV-A. We use a different mapping as specified next.

2) **Mapping 8-ary to three bits:** We map the 8-ary symbols to a 3-bit binary representation as specified in Figure 7 showing the decoding function. Note that this representation is not the standard 8-ary-to-binary mapping of Figure 3. We order the bits’ significance from right to left (the LSB is the right bit), and note the following properties of this mapping. 1) Between two physical states adjacent horizontally or vertically, exactly one of the two higher bits is flipped, and between two physical states adjacent on the main diagonal (bottom-left to top-right), both high bits are flipped. 2) The LSB flips along the secondary diagonal (top-left to bottom-right). These properties will be used later on for decoding.

3) **Outer binary EC codes:** We use the code $C_2[2N, 2N-r_{\text{1}}]$ with design distance $2\tau + 1$ for the two upper bits, and for the lower bit we use the code $C_2[N, N-r_{\text{2}}]$ designed with distance $\tau + 1$. 


All other lower bits are correct. Since the decoder sees at most Amag-1 error we input an erasure to the lower bit's decoder. C can correct all these errors. Next we decode \( \tau \) errors. Since \( \tau \) errors in both their cells. We show that any combination with \( \tau \) suffered an Amag-1 error in exactly one of their two cells, and 2 upper bits, thus the seen in Figure 7, each Amag-1 error results in an error in the \( (\text{Section IV-C.3, the concatenated code is the with the logical mapping in Figure 7 and EC codes specified in} \) Section IV-C.3, the concatenated code is the \( (2N,8,4,2^{2N-r_1},2^{N-r_2}) \) WOM code obtained by taking \( N \) copies of \( W \) and inputting to each two bits of \( C_2[2N,2N-r_1] \) as the higher bits and a bit of \( C_2[N,N-r_2] \) to its lower bit.

The properties of Construction 5 result in the following proposition. The proof includes the decoding algorithm.

**Proposition 5.** For any \( N \) and \( \tau \), a \( (2N,8,4,2^{3N-r_1-r_2}) \) code obtained by Construction 5 is \( \tau \) Amag-1 error correcting. Moreover, the code has EC-rate

\[
1 - \frac{3}{4} \left( \frac{2\tau - \frac{3}{2}}{N} \right) \log_2 N + \frac{4}{5} \frac{\tau}{N} - \frac{3}{5}.
\]

**Proof:** Let \( r_1 \) denote the number of copies of \( W \) that suffered an Amag-1 error in exactly one of their two cells, and \( r_2 \) denote the number of copies of \( W \) that suffered Amag-1 errors in both their cells. We show that any combination with \( r_1 + 2r_2 \leq \tau \) is correctable. We first decode the binary code \( C_2[2N,2N-r_1] \) of the two higher bits. From the mapping seen in Figure 7, each Amag-1 error results in an error in the two upper bits, thus the 2N upper bits suffer at most \( r_1 + 2r_2 \) errors. Since \( r_1 + 2r_2 \leq \tau \), the decoder of \( C_2[2N,2N-r_1] \) can correct all these errors. Next we decode \( C_2[N,N-r_2] \) of the lower bit. For each copy of \( W \) that suffered a single Amag-1 error we input an erasure to the lower bit’s decoder. All other lower bits are correct. Since the decoder sees at most \( r_1 \leq \tau \) erasures, it can recover the lower bits successfully.

To prove the EC-rate, we observe that

\[
2^{3N-r_1-r_2} = 8^{N - \frac{1}{2}(r_1 + r_2)}.
\]

Hence in Definition 5 we have \( r_{\text{eff}} = \frac{1}{3}(r_1 + r_2) \) 8-ary symbols. Substituting in \( r_{\text{eff}} \) the binary BCH redundancies from (2)

\[
r_1 = \frac{1}{2}(2\tau - 1) \log_2 2N, \quad r_2 = \frac{1}{2}(\tau - 1) \log_2 N,
\]

and after some rearrangement we get (9) for the EC-rate. ■

As \( N \) grows, \( r_{\text{eff}} \) of Construction 5 tends to \( \left( \frac{2}{3} \tau - 1 \right) \log_2 N \) (see (9)), which is the lowest redundancy among all the constructions given so far in the paper. The closest redundancy to this is achieved by Construction 3, but without guarantee to correct double errors in the same WOM copy. Another benefit of using Construction 5 in the presence of Amag-1 errors is that both codes are binary codes, which implies simpler implementation with BCH codes, and also greater flexibility to use alternative lower complexity binary codes like LDPC codes. A summary and comparison of the constructions in this section is given in Table I. In the right column we write the EC-rate difference (gap) between each construction and the basic Construction 1. Construction 2 has a gap that is constant in \( \tau \), while the improvement is more significant with Construction 3 as it grows with \( \tau \), albeit with a slight restriction of the error model. Construction 4 has a less restrictive error model than Construction 3, and still offers EC-rate gap that grows with \( \tau = r_1 + r_2 \), so long that \( r_2 < \tau/4 \). Finally, Construction 5 offers the largest EC-rate advantage (for Amag-1 errors) when \( N \) is large enough to make the term \(-8\tau\) negligible.

**V. EC-WOM CODES FOR RANDOM ERRORS**

After constructing EC-WOM codes for guaranteed worst-case errors in Section IV, we now move to study EC-WOM codes for random errors. Our objective is to improve and analyze the WOM error correction over realistic memory channel models. To keep implementation complexity low, we only use binary EC-codes throughout the section, and use multiple codes through the concept of multi-level coding [22]. To not confuse the levels in the multi-level coding hierarchy with cell levels, we call the former "bit-levels" instead of just "levels". As in the previous parts of the paper, in its remainder too we assume a basic readout of cells returning a "hard" estimate in \( \{0,\ldots,q-1\} \) for each cell level. Soft decoding of our constructions is possible without finer readout by extracting reliability information from the decoder of the inner WOM code.

**A. Multi-level coding**

In multi-level coding, one maps each high-order symbol (e.g., a modulation symbol in communications, a \( q \)-ary cell level in standard Flash, or a \( M \)-ary WOM logical symbol in this paper) to multiple bits, each coded by a separate binary EC-code. Each individual EC-code is designed for the channel induced on its bit-level by the high-order channel. Multi-level coding orders the bit-levels in a hierarchy, and applies multi-stage decoding, whereby the decoder outputs of a bit-level are supplied to the decoders of the bit-levels above it in the hierarchy. We now define the specific variant of multi-stage decoding we use in this section, first without taking into account the WOM code feeding the multi-stage decoder.
The multi-stage decoder: The inputs to the multi-stage decoder are $m$ bit-levels ordered $1, \ldots, m$, each having $N$ inputs from the alphabet $\{0, 1, \perp\}$. Each bit-level code with some minimum distance $d_i$ is decoded by an error-erasure (hard-decision) bounded-distance decoder. In bit-level 1, the inputs to the decoder are equal to the inputs to the multi-stage decoder in coordinates that have 0 or 1, and are erasures in coordinates that have $\perp$. For each bit level $i > 1$, we in addition input erasures in coordinates that have 0 or 1 and at least one bit-level in $\{1, \ldots, i - 1\}$ corrected an error in this coordinate.

The most distinctive feature of the multi-stage decoder here compared to prior multi-stage decoder definitions is that it gets $\perp$ (erase) inputs despite using only hard estimates from the physical readout. Previously (e.g. in M-PSK/QAM communications) erase inputs were only possible with soft channel outputs, but here the $\perp$s coming from the inner WOM code offer a significant decoding-performance boost even with hard channel outputs. The multi-stage decoder used here is a hybrid between the “hard” and “soft” decision variants [22], in that it uses hard-decision decoding but propagates bit errors upwards as a soft indication of an unreliable coordinate. Our choice to erase bits with a corrected error in any lower bit-level is a simple special case of a more complex erasing problem [7]. We justify this simple choice in the next sub-section discussing the WOM constructions for multi-level coding. We now move to introduce the WOM codes into the multi-level coding framework. For this we first revisit and refine the definition of the WOM decoding function from Section II.

Definition 10. The WOM decoding function is redefined as $\psi : \{0, \ldots, q - 1\}^n \times \{1, \ldots, t\} \rightarrow \{0, \ldots, M - 1\} \cup \perp$, where the second argument $k \in \{1, \ldots, t\}$ is the write number, and the output $\perp$ represents a hole (invalid state). In addition, let $m = \log_2 M$ and a $M$-ary to binary mapping be given. Define the WOM binary decoding function as $\psi_b : \{0, \ldots, q - 1\}^n \times \{1, \ldots, t\} \rightarrow \{0, 1\}^m \cup \perp^m$.

Note that the addition of the $k \in \{1, \ldots, t\}$ argument to the decoding function is done for the purpose of error detection, so as usual in WOM any physical state is mapped to at most one (non $\perp$) logical state. As an example for Definition 10 we look at the WOM code of Figure 5 and the physical state $x = (2, 0)$. We get $\psi(x, 1) = 5$, and $\psi_b(x, 2) = 101$. Alternatively, $\psi_b(x, 2) = \perp$ and $\psi_b(x, 2) = \perp \perp \perp$, because $x$ is not possible in write number 2. We also use the convention of $\psi(x, k) = \perp \forall x \not\in \{0, \ldots, q - 1\}^n$ (in case the physical readout returns a value outside the $q$ legal levels).

Our focus in this paper is on WOM codes with logical alphabet size $M = 8$, translating to $\log_2 8 = 3$ bit-levels in the multi-level coding hierarchy (but the same techniques can be extended to more bit-levels for higher $M$ values). We start our discussion of WOM multi-level coding by listing the concatenation components, as done in each sub-section of Section IV.

1) Low error-propagation WOM codes: The WOM codes used in this section are suitably designed to minimize the propagation of errors from the $q$-ary memory alphabet to the binary EC-code alphabet. That is, memory read/write noise induces wrong $q$-ary values in the read physical states, and we wish these errors to be expressed in few bit errors seen by the binary EC decoders. Lower error propagation can be achieved by introducing invalid physical states in the decoding function, which are physical states not corresponding to any logical state (and thus will never be reached by the update function). These invalid states are called in the sequel holes. To ease the design of such WOM codes, we later define the WOM code’s propagation index as a single-letter measure of its error-propagation performance.

2) Mapping 8-ary to three bits: The 8-ary logical alphabet of the WOM code is mapped to 3 bits with the following property: physical states adjacent horizontally or vertically have the opposite binary value in their lowest bit (LSB). Thanks to this property, a successful decoding of the LSB will result in decoding the upper two bits from physical states separated by Manhattan distance at least 2 (see Figure 8b for LSB=0). The upper two bits are mapped such that physical states adjacent diagonally have on average a small number of bits difference. Manhattan distance 1 and diagonal errors are, respectively, the first and second dominant errors in memory channels such as those with additive Gaussian-distributed noise. An example of such a mapping is shown in Figure 8 for the case of the tiling WOM code: (a) shows the tiling-based (2, 8, 4, 8) WOM decoding function and its mapping, while (b) shows the subset of the physical states with LSB equal to 0. Assuming that the LSB was correctly decoded to 0 by bit-level 1’s decoder, (b) shows that a physical error with Manhattan distance 2 is necessary to introduce an error in any of the upper two bits (recall that according to the specification of the multi-stage decoder, a Manhattan distance 1 error will be found by bit-level 1 and result in erasures and not errors in the upper bits.)
Figure 8: Multi-level coding WOM decoding function and mapping (tiling WOM). (a) The full decoding function and mapping, and (b) the residual decoding function after the LSB was decoded to 0 by bit-level 1.

Figure 9: The MLcode1 construction.

Figure 10: The MLcode2 construction.

Figure 11: The MLcode3 construction.

In addition to the tiling WOM code and its mapping shown in Figure 8, we construct three additional WOM codes – all with parameters \((n,q,t,M) = (2,8,4,8)\) – to work better in the multi-level coding setting. The codes are specified in Figures 9, 10, and 11. As before, the decoding functions are given explicitly, while the update functions can be readily inferred from the colors representing the regions of the \(t = 4\) writes. The codes are specified with their binary mappings. We name the constructions MLcode1 (Figure 9), MLcode2 (Figure 10), and MLcode3 (Figure 11). MLcode1 is based on [14] but with injected holes, and MLcode3 is based on Construction 3 but with more holes. Examining the WOM constructions in Figures 8,9,10,11 justifies the specification of the multi-stage decoder in Section V-A: any bit corrected \(0 \rightarrow 1\) or \(1 \rightarrow 0\) by bit-level \(i\) implies that the readout is mostly useless for bit-levels \(j > i\), because we can find in most cases two physical states at the same Manhattan distance from the read physical state with opposite binary values at bit-level \(j\). For example, in Figure 9 if we read physical state \((4,4)\)
we assume the basic “WOM-independent” multi-stage decoder

While this may offer improvement in practice, in the analysis
bit, and thus may not be erased by the decoder of bit-level
(1
states at Manhattan distance 1 from
0
corrects the LSB
in Figure 11 if we read physical state
(1
of the WOM code to decide what bits to erase. For example,
can be obtained if the multi-stage decoder uses the knowledge
decoder is a justified decision. Better correction performance
j > i
two bits. Hence erasing bit-levels
1
and bit-level 1
and bit-level 1 corrects the LSb
0 → 1, then in Manhattan
distance 1 from (4, 4) we have all possibilities for the upper
two bits. Hence erasing bit-levels
j > i
by the multi-stage
decoder is a justified decision. Better correction performance
can be obtained if the multi-stage decoder uses the knowledge
of the WOM code to decide what bits to erase. For example,
in Figure 11 if we read physical state
(1, 1) and bit-level 1
corrects the LSB
0 → 1, then the only two valid physical
states at Manhattan distance 1 from
(1, 1) have
0 in the second
bit, and thus may not be erased by the decoder of bit-level 2.
While this may offer improvement in practice, in the analysis
we assume the basic “WOM-independent” multi-stage decoder
defined in Section V-A.

Our goal is to analyze the error-correcting performance of
the different WOM constructions under multi-level coding with
multi-stage decoding. Such an analysis tool will help in choosing
the best WOM code and in designing the EC codes used
in each bit-level.

C. Bounding the decoding-error probability

Our main tool to analyze multi-level coding of WOM codes
is through the calculation of the decoding-error probability. In
multi-level coding, a decoding-error event happens when at least
one of the bit-level decoders either returns the wrong
.codeword or declares failure. We denote the decoding-error
probability by
P_E.
In addition, we define the following.

Definition 11. Let
p_i, q_i
 denote the probability that a bit input
to the
i-th
bit-level decoder is an error, erasure, respectively. \( p_i \) and
\( q_i \) correspond to a given write
k, but we keep this dependence
implicit. Let \( P_{E_i} \) be the probability of decoding error at bit-
level \( i \), conditioned that all bit-levels \( 1, \ldots, i-1 \) were decoded
correctly.

Since calculating the exact decoding-error probability is dif-
ficult, we instead calculate an upper bound that will offer a
performance guarantee. We calculate an upper bound on
\( P_E \) using the well-known union-bound technique [22]

\[
P_E \leq \sum_{i=1}^{m} P_{E_i}.
\]

Given the input error,erasure probabilities
\( p_i, q_i \) to an error+erasure
bounded-distance decoder of a code with minimum distance
\( d_i \), the bit-level’s decoding-error probability is obtained through the
standard trinomial formula

\[
P_{E_i} = 1 - \sum_{l,j:2l+j \leq d_i-1} \frac{N!}{l!j!(N-l-j)!} p_i^l q_i^j (1-p_i-q_i)^{N-l-j}.
\]

We now need to calculate \( p_i \) and \( q_i \) for each bit-level \( i \), and
they will depend on the chosen WOM code. Moreover, in the
input erasure probability \( q_i \) we need to include erasures due
to errors in lower bit-levels, hence the required conditioning
in Definition 11 on correct decoding of the lower bit-levels.

Calculating \( p_i \) and \( q_i \) for bit-level \( i \) is now explained. For a
binary string \( a \) we use in the sequel \( a_i \) to denote its
\( i \)-th bit,
and \( a_i^j \) to denote its sub-string \( a_i, \ldots, a_j \), for any \( i < j \leq m \).
To calculate
\( p_i \), we marginalize it over the correct bits in the
bit-levels up to \( i \)

\[
p_i \triangleq P_i(\text{error}) = \frac{1}{2^i} \sum_{(b_1, \ldots, b_i) \in \{0,1\}^i} P_i(\text{error}|b_1, \ldots, b_i),
\]

where this expression assumes that written logical states follow
the uniform distribution. A similar calculation is done for
\( q_i \triangleq P_i(\text{erasure}) \), but we skip it to avoid duplicity. Then each
probability in the sum is further marginalized over the correct
physical state of the WOM code feeding the bit-level decoder

\[
P_i(\text{error}|b_1, \ldots, b_i) = \sum_{x \in \{0, \ldots, q-1\}^n} P_i(\text{error}|b_1, \ldots, b_i, x) \cdot P(x|b_1, \ldots, b_i). \quad (11)
\]
The first term in the sum are the probabilities \( P_i(\text{error}|b_1,\ldots,b_i, x) \) that can be calculated from the WOM decoding function and the channel distribution governing the physical-state transitions. Note that \( b_1,\ldots,b_i \) are fully determined given \( x \) from the WOM decoding function and mapping, so we may also write this probability simply as \( P_i(\text{error}|x) \).

The second term in the sum are the probabilities \( P(x|b_1,\ldots,b_i) \) that can be calculated from the WOM update function: there may be multiple physical states in write \( k \) with bit values \( b_1,\ldots,b_i \), and to know the distribution of \( x \) one needs to enumerate all combinations of logical values written in the write sequences \( 1,\ldots,k \) that end with these \( i \) bit values in the \( k \)-th write. This distribution is in general not uniform (even though logical values are uniform), as one \( x \) may be reached by the update function in more write sequences than another with the same logical value. We clarify that \( P(x|b_1,\ldots,b_i) \) does not depend on the channel, only on the WOM code.

To calculate \( P_i(\text{error}|b_1,\ldots,b_i, x) \) we sum the channel’s transition probabilities from \( x \) to all states \( x' \) such that \((\psi_b(x',k))^{-1}_i = (\psi_b(x,k))^{-1}_i \) and \((\psi_b(x',k))^{-1} = 1 - (\psi_b(x,k))_i\). Similarly, to calculate \( P_i(\text{error}|b_1,\ldots,b_i, x) \) we sum the transition probabilities from \( x \) to all states \( x' \) such that \( \psi_i(x',k) = 1 \), or that \((\psi_b(x',k))^{-1}_i \neq (\psi_b(x,k))^{-1}_i \), where two strings are not equal if at least one bit differs between them. These conditions on \( x' \) imply that as we go up in the bit-level hierarchy, \( q_i \) grows while \( p_i \) diminishes.

**Example 3.** Suppose cells are read with additive zero-mean white Gaussian noise (AWGN) with standard deviation \( \sigma \). Define \( Q(s) = \frac{1}{\sqrt{2\pi}} \int_s^\infty \exp(-u^2/2)\,du \) as the usual Gaussian tail function. For the WOM code in Figure 10, if \( x = (1,1) \) is written in the first write \( (k=1) \), then in bit-level \( i=1 \) we have

\[
P_1(\text{error}|0,(1,1)) = 4 \left[ Q(0.5) - Q(1.5) \right] - \left[ Q\left(\frac{0.5}{\sigma}\right) - Q\left(\frac{1.5}{\sigma}\right) \right] - 3 \left[ Q\left(\frac{0.5}{\sigma}\right) - Q\left(\frac{1.5}{\sigma}\right) \right]^2.
\]

This is the probability that the channel takes us from \( x = (1,1) \) to one of its 4 horizontal/vertical neighbors. Similarly,

\[
P_1(\text{erasure}|0,(1,1)) = 1 - P_1(\text{error}|0,(1,1))
\]

Erasure is the outcome for \( x = (1,1) \) in all \( x' \) states except (negative terms from left to right): states that give an error outcome, when \( x' = x \), and the three diagonal neighbors of \( x = (1,1) \) that are within the region reachable at write number \( k = 1 \). Moreover, in the code of Figure 10 the distribution \( P(x|b_1,\ldots,b_i) \) is uniform for any \( i \) in write \( k = 1 \) because in that write there is a 1-1 mapping between the logical and physical states. However, in write \( k = 4 \) we expect to have

\[
P(x = (2,7)|1,0,0) < P(x = (6,5)|1,0,0),
\]

because \((2,7)\) is reachable from only three states of write \( k = 3 \) while \((6,5)\) is reachable from all states except one.

**D. Decoding-error probability results for 4 WOM codes**

We now use the analysis of Section V-C to calculate and plot the decoding-error performance of the 4 WOM codes in Figures 8,9,10,11. Recall that all the codes have WOM parameters \( (n,q,M,t) = (2,8,8,4) \); in this sub-section we take the block length (number of parallel WOM-code copies) to be \( N = 255 \). The channel noise is AWGN with \( \sigma \) values between 0.22 and 0.26, and the distance profile \((d_1,d_2,d_3)\) of the bit-level codes was optimized empirically to get the best performance for a given total redundancy budget. The results are plotted in Figures 12,13,14,15, each one showing the same upper-bound calculation for a different write number \( k \in \{1,2,3,4\} \) (recall from Section V-C that \( p_i \) and \( q_i \) depend on the WOM code and the write number \( k \)). Each figure shows the decoding-error probability upper bound for the 4 WOM codes, and in comparison to non WOM-coded information stored on the same \( 255 \times 2 \) cells with the same total EC redundancy (the non-WOM redundancy allocation to bit-levels was optimized separately). It is observable that different WOM codes differ considerably in error probability even though they have the same parameters and redundancy. It is also observed that WOM-coded memory enjoys higher reliability than non-WOM, meaning that the same redundancy that is invested in the re-writing feature also serves us in improving reliability. An interesting property is that the ordering of WOM codes in reliability is not absolute but depends on the write number \( k \); for example MLCode3 is the best in \( k = 3 \) but MLCode2 is the best in \( k = 4 \). The tiling construction has the worst reliability in all writes, hence it is not recommended for use. We next show in Figures 16-19 a comparison of the error probability bound of MLCode3 between a decoder that knows the write number and a decoder that does not know the write number. A decoder that does not know the write number cannot use the \( k \)-aware WOM decoding function of Definition 10, and thus misses opportunities to supply the EC-decoder with \( \downarrow \) inputs when errors occur. The plots reveal a significant performance gap in favor of knowing the write number. The smaller gap for \( k = 4 \) (Figure 19) can be explained by looking at MLCode3’s specification in Figure 11, and observing that at the 4-th write there are few Manhattan-1 neighbors belonging to the 3-rd write.

We next look for a simpler characterization of the WOM code’s reliability without the complexity of (11) that requires the enumeration of all pairs of physical states \( x,x' \). For that we define the **propagation index** in the next sub-section.

**E. Propagation index of a WOM code**

In the realistic scenario where noise and inter-cell interference are not too severe, we expect mag-1 errors to be the dominant error type. For that case we propose a simpler characterization of the WOM code’s reliability – one that only enumerates the physical states’ close neighborhoods in the WOM decoding function.

**Definition 12.** Given a valid physical state \( x \) and another physical state \( y \) of a \((n,q,t,M)\) WOM code we define the **error
Figure 12: Error probability bound, $k = 1$.

Figure 13: Error probability bound, $k = 2$.

Figure 14: Error probability bound, $k = 3$.

Figure 15: Error probability bound, $k = 4$.

Figure 16: Error probability bound for two decoders, ML-Code3, $k = 1$.

Figure 17: Error probability bound for two decoders, ML-Code3, $k = 2$. 
Figure 18: Error probability bound for two decoders, ML-Code3, $k = 3$.

Figure 19: Error probability bound for two decoders, ML-Code3, $k = 4$.

function between $x$ and $y$ in the $k$-th write as

$$
\nu(x, y; k) = \begin{cases} 
1 & \psi(x, k) \neq \psi(y, k) \neq \perp \\
0.5 & \psi(x, k) = \perp \\
0 & \psi(x, k) = \psi(y, k) \neq \perp 
\end{cases}
$$

We define the average error of $x$ in the $k$-th write as

$$
\epsilon(x, k) = \frac{1}{2n} \sum_{l \in \{1, \ldots, n\}, j \in \{-1, 1\}} \nu(x, x + j \cdot e_l; k),
$$

where $e_l$ is the unit vector that is equal to 1 in coordinate $l$ and to 0 in the other $n-1$ coordinates.

The average error measures the average "damage" a mag-1 error causes to a physical state $x$. Note that the above definition of error uses the $M$-ary WOM alphabet and thus is independent of the binary mapping. The WOM code’s propagation index is next defined as an expectation of the average error over the physical states $x$.

Definition 13. Given a $(n, q, t, M)$ WOM code we define the propagation index (PI) in the $k$-th write as

$$
\eta_k = \sum_{x \in \{0, \ldots, q-1\}^n} p_{x,k} \cdot \epsilon(x, k),
$$

where $p_{x,k}$ is the probability to be in physical state $x$ at the $k$-th write. We define the geometric propagation index (GPI) as the propagation index when $p_{x,k}$ is the uniform distribution among the physical states of the $k$-th write. We define the probabilistic propagation index (PPI) as the propagation index when $p_{x,k}$ is the distribution on the physical states induced by the uniform distribution on the logical states in the sequence of $k$ writes.

While PPI better captures the behavior of the WOM code in real write sequences, GPI is easier to calculate and does not depend on the update function of the WOM code.

These definitions are demonstrated in the following example.

Example 4. We are given the WOM code in Figure 20 and wish to calculate its propagation indexes. The GPI and PPI of the WOM code in Figure 20 are

$$
GPI = \begin{cases} 
0.8125 & k = 1 \\
0.8 & k = 2 \\
0.833 & k = 3 
\end{cases},
$$

and

$$
PPI = \begin{cases} 
0.8125 & k = 1 \\
0.8125 & k = 2 \\
0.8633 & k = 3 
\end{cases}.
$$
To calculate the PPI we need a precise specification of the update function without ambiguity when there are multiple update choices, which is included in Figure 20.

We plot in Figure 21 the GPI in (a) and PPI in (b) of the 4 WOM codes analyzed in Section V-D. We see that both GPI and PPI predict for writes $k = 1, 2, 3$ the relative order of the WOM codes in terms of decoding-error probability bound. But only the more refined PPI manages to predict the correct order in write $k = 4$. That means that the advantage of MLCode2 in the 4-th write originates from its update function giving preference to more reliable physical states.

**F. EC-WOM decoding with reliability information**

So far in the paper we only used hard-decision error-and-erasure decoders for the EC-codes in the EC-WOM construction. Toward implementing soft-decision decoders, we next show how to use the inner WOM code to provide symbol reliability (likelihood) information to be used as inputs to soft-decision decoders. This is an attractive source of soft inputs because it comes without need to obtain additional information from the memory channel beyond the quantized levels $\{0, \ldots, q-1\}$. This can be done based on the observation that the reliability of a read physical state $x \in \{0, \ldots, q-1\}^n$ depends on its neighborhood in the WOM code’s decoding function. While in hard decision $x$ is only characterized by whether $\psi(x, k) = \perp$ or not, in soft decision we can characterize the reliability of $x$ based also on $\psi(x', k)$ of its neighboring physical states $x'$. For example, a bit $(\psi_b(x, k))_i = b$ is more reliable if three of its Manhattan-1 neighbors have $(\psi_b(x', k))_i = \perp$, than if all four neighbors have $(\psi_b(x', k))_i = 1 - b$. Our characterization of the reliability of read physical states relies on the following natural definition of reliability.

**Definition 14.** Let $x$ be some (not necessarily valid) WOM physical state. We define the reliability of the $i$-th bit of $x$ in the $k$-th write as

$$r_i(x, k) = \frac{\sum_y \lambda(y, k) P(y \rightarrow x)}{\sum_y |\lambda(y, k)| P(y \rightarrow x)},$$

where

$$\lambda(y, k) = \begin{cases} 1 & (\psi_b(y, k))_i = 0 \\ -1 & (\psi_b(y, k))_i = 1 \\ 0 & (\psi_b(y, k))_i = \perp \end{cases}$$

and $P(y \rightarrow x)$ is the probability that the channel takes the input physical state $y$ to the output physical state $x$.

Note that the sums in (14) are over all physical states $y$, including $y = x$. In practice we will only sum over $y \in Ne(x)$, that is, physical states in the neighborhood of $x$ which is defined as the physical states with non-negligible $P(y \rightarrow x)$. In that case $x$ is called the center of the neighborhood $Ne(x)$.

Definition 14 for reliability is natural because the numerator adds transition probabilities with their sign reflecting the logical bit values, while the denominator normalizes with the sum of absolute values of the same probabilities. In general $0 \leq r_i(x, k) \leq 1$; in particular $r_i(x, k) = 1$ if and only if all valid physical states in $Ne(x)$ have the same logical bit.

**Example 5.** Consider the WOM code shown in Figure 10, and let $x = (6, 1)$. $x$ is not valid, in particular in the third write, i.e., $\psi(x, 3) = \perp$. Suppose the channel is defined to cause a mag-1 error with probability $p$ in a uniformly selected direction, and no error with probability $1 - p$. Then because $x$ has only one valid physical state $y = (6, 2)$ in $Ne(x)$, we have that $r_i(x, 3) = 1, \forall i \in \{1, 2, 3\}$. This means that $x$ gives certain bit values, even though it is itself an invalid state.

To obtain simple and robust soft-decision decoders, we focus next on decoders that only require the ordering of the symbol reliabilities at the decoder input, without assigning numerical values to individual-symbol reliabilities. Order-based soft-decision decoders based on GMD [9], such as Ordered...
Statistics [10] are known to offer very good decoding performance. To capture the more significant behaviors of EC-WOM codes with respect to reliability ordering, we next assume the channel is AWGN with a small $\sigma$. With small $\sigma$ we can take as a good approximation the neighborhood of $x$ to be $x$ and its 4 Manhattan-1 neighbors (2 horizontal and 2 vertical). Formally, from now on we assume $Ne(x) = \{ y : |x_1 - y_1| + |x_2 - y_2| \leq 1 \}$. Now we can define the following.

**Definition 15.** Let $x$ be some (not necessarily valid) WOM physical state. We define the neighborhood profile of the $i$-th bit of $x$ in the $k$-th write as the tuple $g_i(x, k) = [l_0, l_1]$, where $l_0$ is the number of zeros, $l_1$ is the number of ones, and $4 - l_0 - l_1$ is the number of $\perp$s in the multiset $\{(\psi_b(y, k))_i : y \in Ne(x) \setminus x\}$.

Clearly if we know both $(\psi_b(x, k))_i$ and $g_i(x, k)$ we can calculate the reliability $r_i(x, k)$. In Figure 22 we plot the reliability values as a function of the neighborhood profile $[l_0, l_1]$ for $x$ with $(\psi_b(x, k))_i = 0$ in (a), and for $x$ with $(\psi_b(x, k))_i = \perp$ in (b). The values for $x$ with $(\psi_b(x, k))_i = 1$ are symmetric to (a), only exchanging the roles of $l_0$ and $l_1$. With a neighborhood center of $\perp$ the roles of $l_0, l_1$ are symmetric, so the profiles missing from (b) can be completed by exchanging $l_0, l_1$. In the plots the reliabilities are sorted from highest to lowest.

Looking at the $x$-axis of Figure 22 we have a way to order decoder inputs by reliability for soft-decision decoding: the most reliable neighborhood profiles are those with reliability 1 in Figure 22 (a),(b), then the other profiles in decreasing order in (a), and lastly those with lowest reliability in (b). While Figure 22 is specific for $\sigma = 0.3$, the next results show that the ordering of neighborhood profiles is the same for all values of $\sigma$ below some threshold. Let $[l_0, l_1]_0$ be the reliability of neighborhood profile $[l_0, l_1]$ when the neighborhood center’s logical bit is 0, and $[l_0, l_1]_\perp$ be the reliability when the center is $\perp$. Then we have the following two propositions.

**Proposition 6.** For any $\sigma \leq 0.519$ we have the following

\[
\begin{align*}
[4, 0]_0 &= [3, 0]_0 = [2, 0]_0 = [1, 0]_0 = [0, 0]_0 = 0, \\
[4, 0]_\perp &= [3, 0]_\perp = [2, 0]_\perp = [1, 0]_\perp = 1, \\
[3, 1]_0 &\geq [2, 1]_0 \geq [1, 1]_0 \geq [0, 1]_0, \\
[2, 2]_0 &\geq [1, 2]_0 \geq [0, 2]_0, \\
[1, 3]_0 &\geq [0, 3]_0 \geq [0, 4]_0,
\end{align*}
\]

and

\[
\begin{align*}
[3, 1]_\perp &\geq [2, 1]_\perp \geq [2, 2]_\perp = [1, 1]_\perp = [0, 0]_\perp = 0.
\end{align*}
\]

The proof of Proposition 6 is straightforward: the equalities are implied directly from (14) and the inequalities follow from the monotonicity in one of $l_0$ or $l_1$ while the other is fixed. The condition on $\sigma$ ensures that in the entire neighborhood the most reliable binary value equals to the value in the center of the neighborhood (and thus in (17)-(19) a higher $l_0$ or lower $l_1$, while the other in fixed, implies higher reliability). The cutoff value of $\sigma = 0.519$ is when $[0, 4]_0$ (the lowest reliability of a neighborhood profile) becomes negative.

Let $z_1$ denote the probability that a particular cell is read with a mag-1 error (recall Definition 7), and let $z_0$ denote the probability that the cell is read with no error. With the AWGN noise model we have $z_1 = Q(0.5/\sigma) - Q(1.5/\sigma)$ and $z_0 = Q(-0.5/\sigma) - Q(0.5/\sigma)$. For $y$ that is a particular Manhattan-1 neighbor of $x$ we have $P(y \rightarrow x) = z_1 z_0$. We also have $P(x \rightarrow x) = z_0^2$. We can now prove a less obvious ordering relation among the neighborhood reliabilities.

**Proposition 7.** For any $\sigma \leq 0.795$ we have

\[
\begin{align*}
[0, 1]_0 &\geq [2, 2]_0, \\
[0, 2]_0 &\geq [1, 3]_0.
\end{align*}
\]
Moreover, for any $\sigma \leq 0.341$ we have,

$$[0, 4]_0 \geq [3, 1]_L.$$  \hfill (23)

**Proof:** Recall that $P(y \to x) = z_1 z_0$, $P(x \to x) = z_0^2$. Substituting these probabilities in (14) we get the reliabilities $\sigma$ for any $\sigma$.

**Corollary 8.** For any $\sigma \leq 0.341$ the order of neighborhood profiles in non-increasing reliability is

$$[4, 0]_0, [3, 0]_0, [2, 0]_0, [1, 0]_0, [0, 0]_0,$$
$$[4, 0]_L, [3, 0]_L, [2, 0]_L, [1, 0]_L,$$
$$[3, 1]_0, [2, 1]_0, [1, 1]_0, [0, 1]_0, [2, 2]_0,$$
$$[1, 2]_0, [0, 2]_0, [1, 3]_0, [0, 3]_0, [0, 4]_0,$$
$$[3, 1]_L, [2, 1]_L, [2, 2]_L, [1, 1]_L, [0, 0]_L.$$  

and recall that $[l_0, l_1]_1$ can be entered in that order by the symmetry $[l_0, l_1]_1 = [l_1, l_0]_0$.

**VI. CONCLUSION**

The key message to take from this study of EC-WOM codes is the interesting inter-relations between the WOM and error-correction features of the memory. When WOM and error correction are considered jointly, some of the WOM redundancy can be directed for better error correction, reducing the overall cost of obtaining both features. Concrete examples of this are given in the constructions and analyses of this paper. Another benefit of concatenating WOM and EC codes is studied in the second part of the paper: obtaining soft information for EC decoding from the WOM decoder, without need to increase the raw measurement precision of the memory cells. There are many interesting problems left open by this paper: 1) improving the constructions of this paper for the same WOM parameters, 2) extending the constructions to additional WOM parameters (also beyond $n = 2$, which looks non-trivial) and other error models, 3) concatenating WOM codes with other types of codes like LDPC or polar codes, 4) developing soft-decision decoders using the WOM reliability information, and 5) studying the fundamental limits of WOM+EC combination with low complexity. Even further, an interesting direction beyond the scope of the current scheme is how to design the EC+WOM code to "recover" from a write error detected in an intermediate write, so that subsequent writes can still be served.

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**REFERENCES**

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