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Distance Bounds for Periodically Time-Varying and Tail-Biting LDPC Convolutional Codes

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Abstract

Existence type lower bounds on the free distance of periodically time-varying LDPC convolutional codes and on the minimum distance of tail-biting LDPC convolutional codes are derived. It is demonstrated that the bound on free distance of periodically time-varying LDPC convolutional codes approaches the bound on free distance of general (non-periodic) time-varying LDPC convolutional codes derived in [1] as the period increases. The proof of the bound is based on lower bounding the minimum distance of corresponding tail-biting LDPC convolutional codes, which is of interest in its own right.

Index Terms

Free distance, minimum distance, lower bounds, low-density parity-check (LDPC) codes, LDPC convolutional codes, tail-biting LDPC convolutional codes.

I. INTRODUCTION

LDPC block codes were invented by Gallager [2] in the 1960s. The construction of the corresponding convolutional counterparts, LDPC convolutional codes (LDPCCCs), was first...
presented\textsuperscript{1} in [3]. While minimum distance bounds for block LDPCs were derived in Gallager’s original work [2], the first analytical lower bound on the free distance of LDPCCCs was only derived recently [1]. The proof presented in [1] holds for an ensemble of general (non-periodic) time-varying LDPCCCs and must employ a special expurgation technique to compensate for the non-periodic structure of the ensemble.

In this paper, we derive an existence type lower bound on the free distance of periodically time-varying LDPCCCs. We show that, as the period increases, the new bound approaches the bound on free distance of non-periodic LDPCCCs derived in [1]. The proof presented for the new bound is based on considering the minimum distance of tail-biting LDPCCCs (TB-LDPCCCs) [8]. In particular, we lower bound the minimum distance of TB-LDPCCCs constructed from an ensemble of periodically time-varying LDPCCCs and use this to lower bound the free distance of the original ensemble.

Tail-biting was introduced by Solomon and van Tilborg [9] and independently by Ma and Wolf [10] as a method of terminating a convolutional code without the rate loss caused by standard termination. The resulting tail-biting codes have a dual nature, i.e., they simultaneously have the properties of both block and convolutional codes. As a consequence, their minimum distance depends both on the block length of the tail-biting code and the constraint length of the convolutional code.

The minimum distance of conventional (non-LDPC) tail-biting codes equals the minimum of two related distance measures, \( d_{\text{intra}} \) and \( d_{\text{inter}} \) [11]. The \textit{intra} minimum distance \( d_{\text{intra}} \) reflects the convolutional code properties of the tail-biting code and is lower bounded by the Costello bound [6] on the free distance of convolutional codes. The \textit{inter} minimum distance \( d_{\text{inter}} \) reflects the block code properties of the tail-biting code and is lower bounded by the Varshamov-Gilbert bound [4], [5] on the minimum distance of block codes. Analogous to conventional tail-biting convolutional codes, the minimum distance of TB-LDPCCCs is lower bounded by the minimum of \( d_{\text{intra}} \) and \( d_{\text{inter}} \), where \( d_{\text{intra}} \) is lower bounded by the bound on free distance of LDPCCCs derived in [1] and \( d_{\text{inter}} \) is lower bounded by Gallager’s bound on minimum distance of LDPC block codes [2].

The paper is organized as follows. Section II presents the definition of the LDPCCC code

\textsuperscript{1}The basic idea of LDPCCCs was first described in Tanner’s patent application [7].
ensemble considered. Section III is devoted to lower bounding the minimum distance of TB-LDPCCCs, and a lower bound on the free distance of periodically time-varying LDPCCCs is proved in Section IV. Numerical results are given in Section V, and Section VI concludes the paper.

II. AN LDPC CONVOLUTIONAL CODE ENSEMBLE

In [3], a rate $R = b/c$ binary convolutional code was defined as the set of sequences $v_{[0,\infty)} = (v_0, v_1, \ldots)$, $v_t \in \mathbb{F}_2^c$, satisfying the equality $v_{[0,\infty)} H^T_{[0,\infty]} = 0$, where the semi-infinite syndrome former (transposed parity-check) matrix $H^T_{[0,\infty]}$ is given by

$$H^T_{[0,\infty]} = \begin{pmatrix}
H^T_0(0) & \ldots & H^T_{m_s}(m_s) \\
& \ddots & \ddots \\
& & H^T_0(t) & \ldots & H^T_{m_s}(t + m_s) \\
& & & \ddots & \ddots
\end{pmatrix}$$  \hspace{1cm} (1)

and each entry $H^T_i(t + i)$ is a $c \times (c - b)$ binary matrix. To satisfy an easy encoding property (see [3], [13]), the matrices $H^T_i(t)$ must have full rank for all time instants $t$, and hence we assume that the last $(c - b)$ rows of $H^T_i(t)$ are linearly independent for all $t$. Then the first $b$ symbols of $v_t$ at each time instant $t$ are information symbols and the last $(c-b)$ symbols are parity symbols. The largest $i$ such that $H^T_i(t+i)$ is a non-zero matrix for some $t$ is called the syndrome former memory $m_s$. A $(J, K)$ regular LDPCCC is defined by a syndrome former that contains exactly $J$ ones in each row and $K$ ones in each column (starting from the $(c - b)m_s + 1)$th column).

Now we define a special sub-class of $(J, K)$ regular LDPCCCs, where the component sub-matrices $H^T_i(t)$ are composed of $M \times M$ binary permutation matrices$^2$. Let $a = \text{gcd}(J, K)$ denote the greatest common divisor of $J$ and $K$. Then there exist positive integers $J'$ and $K'$ such that $J = aJ'$ and $K = aK'$ and $\text{gcd}(J', K') = 1$. For $i = 0, 1, \ldots, a - 1$, the $K'M \times J'M$

$^2$This subclass was considered in [1] in order to prove a lower bound on $d_{\text{free}}$ for non-periodic LDPCCCs.
sub-matrices $H_i^T(t + i)$ of the syndrome former are

$$H_i^T(t + i) = \begin{pmatrix}
    P_i^{(0,0)}(t + i) & P_i^{(0,1)}(t + i) & \ldots & P_i^{(0,J'-1)}(t + i) \\
    P_i^{(1,0)}(t + i) & P_i^{(1,1)}(t + i) & \ldots & P_i^{(1,J'-1)}(t + i) \\
    \ldots & \ldots & \ldots & \ldots \\
    P_i^{(K'-1,0)}(t + i) & P_i^{(K'-1,1)}(t + i) & \ldots & P_i^{(K'-1,J'-1)}(t + i)
\end{pmatrix}, \quad (2)$$

where each $P_i^{(k,j)}(t + i)$, $k = 0, 1, \ldots, K' - 1$, $j = 0, 1, \ldots, J' - 1$, is an $M \times M$ permutation matrix. All other entries of the syndrome former are $K'M \times J'M$ zero matrices. We assume that the matrix $H_{[0,\infty]}^T$ is periodically time-varying with period $T$, i.e.,

$$H_i^T(t + i) = H_i^T(t + i + T) \quad \forall i, t. \quad (3)$$

In this case, a code is characterized by a section $H_{[0,T-1]}^T$ of the semi-infinite syndrome former $H_{[0,\infty]}^T$.

A syndrome former $H_{[0,T-1]}^T$ for a periodically time-varying $(3,6)$ regular LDPCCC is shown in Fig. 1. In this case, the code construction parameters are $K = 6$, $J = 3$, $a = 3$, $c = 2M$, and $b = M$. Each matrix $H_i^T(t + i)$ consists of two $M \times M$ permutation matrices, i.e.,

$$H_i^T(t + i) = \begin{pmatrix}
    P_i^{(0)}(t + i) \\
    P_i^{(1)}(t + i)
\end{pmatrix}, \quad (4)$$

where (4) has full rank equal to $M$. Therefore the code rate is $M/2M$. Note that by permuting rows of the syndrome former, an equivalent rate $1/2$ $(3,6)$ regular LDPCCC with syndrome former memory at most $3M - 1$ can be obtained (see [1]).

Now suppose that the $M \times M$ permutation matrices comprising the sub-matrices (2) of the syndrome former $H_{[0,T-1]}^T$ are chosen independently and such that each of the $M!$ possible permutation matrices is equally likely. Then we obtain a random ensemble of $(J,K)$ regular $T$-periodic LDPCCCs, which we designate $C(J,K,M,T)$.

The syndrome formers in the ensemble $C(J,K,M,T)$ have memory $m_s = a - 1$, independent of $M$, while $b$ and $c$ depend on $M$. This ensemble of codes is different from the LDPCCCs considered in [3], [12], and [13], where the codes have varying syndrome former memories $m_s$, while the rate parameters $b$ and $c$ are fixed. For the ensemble $C(J,K,M,T)$, as $M$ increases, i.e., as $b$ and $c$ increase, the syndrome formers become increasingly sparse.

During the encoding process, the information sequences are divided into blocks of $b = (K' - J')M$ symbols, which are input to an LDPC convolutional encoder at each time instant.
Fig. 1. One period of a syndrome former for a code in the ensemble $C(3, 6, M, T)$. 

A discussion of the definition of constraint length for LDPCs is given in [1].
III. A LOWER BOUND ON THE MINIMUM DISTANCE OF TB-LDPCCCS

To avoid cumbersome notation, we focus on the \((3,6)\) regular TB-LDPCCC case, i.e., we consider the ensemble \(\tilde{C}(3, 6, M, T)\), although the same technique can also be applied more generally.

Consider a length \(2TM\) vector \(\tilde{v}_{[0,T-1]} = (\tilde{v}_0^{(0)}, \tilde{v}_0^{(1)}, \ldots, \tilde{v}_{T-1}^{(0)}, \tilde{v}_0^{(1)}), \tilde{v}_t^{(h)} \in \mathbb{F}_2^M\), where \(\tilde{v}_t^{(h)} = (\tilde{v}_t^{1(h)}, \tilde{v}_t^{2(h)}, \ldots, \tilde{v}_t^{M(h)})\), \(t = 0, 1, \ldots, T - 1, h = 0, 1\). A vector \(\tilde{v}_{[0,T-1]}\) is a codeword in a \((3,6)\) regular TB-LDPCCC iff it satisfies the \(TM\) equations (constraints) defined by the syndrome former \(\tilde{H}^T_{[0,T-1]}\), i.e.,
\[
\tilde{v}_{[0,T-1]} \tilde{H}^T_{[0,T-1]} = 0.
\]  
(5)

For the ensemble \(\tilde{C}(3,6, M, T)\), these \(TM\) parity-check equations can be divided into \(T\) sets where the \(t\)th set \(\mathcal{S}^{(t)}\), \(t = 0, 1, \ldots, T - 1\), consists of the \(M\) parity-check equations determined by the six permutation matrices located in the \(t\)th block of columns of the syndrome former \(\tilde{H}^T_{[0,T-1]}\).

Let \(\tilde{d}_t^{(h)}\) be the Hamming weight of the vector \(\tilde{v}_t^{(h)}\), \(h = 0, 1, t = 0, 1, \ldots, T - 1\). We then say that \(\tilde{v}_{[0,T-1]}\) has weight composition \(\tilde{d}_{[0,T-1]} = (\tilde{d}_0^{(0)}, \tilde{d}_0^{(1)}, \ldots, \tilde{d}_{T-1}^{(0)}, \tilde{d}_{T-1}^{(1)})\), and the Hamming weight of the vector \(\tilde{v}_{[0,T-1]}\) with weight composition \(\tilde{d}_{[0,T-1]}\) is \(\tilde{d}_{[0,T-1]} = \tilde{d}_0^{(0)} + \tilde{d}_0^{(1)} + \ldots + \tilde{d}_{T-1}^{(0)} + \tilde{d}_{T-1}^{(1)}\). Now note that there exists \(\prod_{t=0}^{T-1} \binom{M}{\tilde{d}_t^{(0)}} \prod_{t=0}^{T-1} \binom{M}{\tilde{d}_t^{(1)}}\) vectors \(\tilde{v}_{[0,T-1]}\) with weight composition \(\tilde{d}_{[0,T-1]}\). Our goal is to calculate the average number of codewords \(\tilde{v}_{[0,T-1]}\) with weight composition \(\tilde{d}_{[0,T-1]}\) for a code in the ensemble \(\tilde{C}(3,6, M, T)\). Finally, in the asymptotic
case, as $M \to \infty$, it is more convenient to operate with the normalized weight composition $\tilde{\rho}_{[0,T-1]} = \left( \tilde{\rho}_0^{(0)}, \tilde{\rho}_1^{(1)}, \ldots, \tilde{\rho}_{T-1}^{(0)}, \tilde{\rho}_{T-1}^{(1)} \right)$, where $\tilde{\rho}_t^{(h)} = \tilde{d}_t^{(h)}/M$.

Letting $\tilde{\rho}_{-2} = \tilde{\rho}_{T-2}$ and $\tilde{\rho}_{-1} = \tilde{\rho}_{T-1}$ for $h = 0, 1$, the probability that a vector $\tilde{v}_{[0,T-1]}$ satisfies the $t$th set of constraints, $S^{(t)}$, $t = 0, 1, \ldots, T - 1$, in the ensemble $\tilde{C}(3, 6, M, T)$ can be upper bounded (see [1], Appendix I) as

$$P(\tilde{\rho}_{[0,T-1]}, S^{(t)}) \leq \exp \left[ \frac{M \tilde{F}(\lambda_1^{(t)}, \lambda_2^{(t)}, \ldots, \lambda_6^{(t)}, \tilde{\rho}_t^{(0)}, \tilde{\rho}_{t-2}, \tilde{\rho}_{t-1}^{(0)}, \tilde{\rho}_{t-1}^{(1)}, \tilde{\rho}_t^{(0)}, \tilde{\rho}_t^{(1)})}{\prod_{i=t-2}^{1} \prod_{h=0}^{1} \left( \frac{M}{\tilde{\rho}_i^{(h)}/M} \right)} \right],$$  

(6)

where $\lambda_1^{(t)}, \lambda_2^{(t)}, \ldots, \lambda_6^{(t)}$ are arbitrary constants,

$$\tilde{F}(\lambda_1, \lambda_2, \ldots, \lambda_6, \rho_1, \rho_2, \ldots, \rho_6) \overset{\text{def}}{=} g(\lambda_1, \lambda_2, \ldots, \lambda_6) - \sum_{k=1}^{6} \lambda_k \rho_k$$  

and

$$g(\lambda_1, \lambda_2, \ldots, \lambda_6) \overset{\text{def}}{=} \ln \frac{\prod_{k=1}^{6} (1 + e^{\lambda_k}) + \prod_{k=1}^{6} (1 - e^{\lambda_k})}{2}.$$

(7)

(8)

We notice that there are $T$ independent constraint sets $S^{(t)}$, $t = 0, 1, \ldots, T - 1$, and the number of vectors $\tilde{v}_{[0,T-1]}$ having normalized weight composition $\tilde{\rho}_{[0,T-1]}$ is

$$N(\tilde{\rho}_{[0,T-1]}) \overset{\text{def}}{=} \prod_{t=0}^{T-1} \left( \frac{M}{\tilde{\rho}_t^{(0)}/M} \right) \prod_{t=0}^{T-1} \left( \frac{M}{\tilde{\rho}_t^{(1)}/M} \right).$$

(9)

Thus the expected number of vectors with normalized weight composition $\tilde{\rho}_{[0,T-1]}$ satisfying all $T$ sets of constraints, i.e., vectors that are codewords in the TB-LDPC code defined by $\tilde{H}_{[0,T-1]}^T$, is given by

$$E(\tilde{\rho}_{[0,T-1]}) \overset{\text{def}}{=} N(\tilde{\rho}_{[0,T-1]}) \prod_{t=0}^{T-1} P(\tilde{\rho}_{[0,T-1]}, S^{(t)}).$$

(10)

Substituting (6) into (10), we obtain the upper bound

$$E(\tilde{\rho}_{[0,T-1]}) \leq \exp \left[ \frac{M \sum_{t=0}^{T-1} \tilde{F}(\lambda_1^{(t)}, \lambda_2^{(t)}, \ldots, \lambda_6^{(t)}, \tilde{\rho}_t^{(0)}, \tilde{\rho}_{t-2}, \tilde{\rho}_{t-1}^{(0)}, \tilde{\rho}_{t-1}^{(1)}, \tilde{\rho}_t^{(0)}, \tilde{\rho}_t^{(1)})}{\prod_{t=0}^{T-1} \left( \frac{M}{\tilde{\rho}_t^{(0)}/M} \right) \prod_{t=0}^{T-1} \left( \frac{M}{\tilde{\rho}_t^{(1)}/M} \right)^2} \right],$$

(11)
where we recall that $\tilde{\rho}_{-2}^{(h)} = \tilde{\rho}_{T-2}^{(h)}$ and $\tilde{\rho}_{-1}^{(h)} = \tilde{\rho}_{T-1}^{(h)}, h = 0, 1$. Now using Lemma 1 and Appendix II from [1], we can further upper bound $E(\tilde{\rho}_{[0,T-1]}^{(h)})$ as

$$E(\tilde{\rho}_{[0,T-1]}^{(h)}) \leq \left[ \prod_{t=0}^{T-1} \sigma(\tilde{\rho}_t^{(0)}, M) \prod_{t=0}^{T-1} \sigma(\tilde{\rho}_t^{(1)}, M) \right]^2 \cdot \exp \left[ M \sum_{t=0}^{T-1} F(\lambda_1^{(t)}, \lambda_2^{(t)}, \ldots, \lambda_6^{(t)}, \tilde{\rho}_1^{(t)}, \tilde{\rho}_2^{(t)}, \ldots, \tilde{\rho}_i^{(t)}) \right], \quad (12)$$

where

$$F(\lambda_1, \lambda_2, \ldots, \lambda_6, \rho_1, \rho_2, \ldots, \rho_6) \equiv \tilde{F}(\lambda_1, \lambda_2, \ldots, \lambda_6, \rho_1, \rho_2, \ldots, \rho_6) - \frac{2}{3} \sum_{k=1}^{6} H(\rho_k), \quad (13)$$

$$\sigma(\rho, M) \equiv \begin{cases} 1, & \text{if } \rho = 0, \\ \sqrt{12M} \rho(1-\rho), & \text{otherwise,} \end{cases} \quad (14)$$

and

$$H(\rho) \equiv -\rho \ln \rho - (1-\rho) \ln(1-\rho). \quad (15)$$

Now suppose that for a particular normalized weight composition $\tilde{\rho}_{[0,T-1]}^{(t)}$, there exists a set $\{\lambda_1^{(0)}, \lambda_2^{(0)}, \ldots, \lambda_6^{(0)}, \lambda_1^{(T-1)}, \lambda_2^{(T-1)}, \ldots, \lambda_6^{(T-1)}\}$ of $6T$ coefficients such that the sum of the functions $\sum_{t=0}^{T-1} F(\lambda_1^{(t)}, \lambda_2^{(t)}, \ldots, \lambda_6^{(t)}, \tilde{\rho}_1^{(0)}, \tilde{\rho}_2^{(0)}, \ldots, \tilde{\rho}_i^{(0)})$ is negative. Then the average number of codewords having normalized weight composition $\tilde{\rho}_{[0,T-1]}^{(t)}$ goes to zero exponentially as $M$ tends to infinity. The tightest bound on $E(\tilde{\rho}_{[0,T-1]}^{(t)})$ can be obtained by minimizing each function $F(\lambda_1^{(t)}, \lambda_2^{(t)}, \ldots, \lambda_6^{(t)}, \tilde{\rho}_1^{(t)}, \tilde{\rho}_2^{(t)}, \ldots, \tilde{\rho}_i^{(t)}), t = 0, 1, \ldots, T-1$, with respect to its parameters $\lambda_1^{(t)}, \lambda_2^{(t)}, \ldots, \lambda_6^{(t)}$.

We begin by defining

$$G(\rho_1, \rho_2, \ldots, \rho_6) \equiv \min_{\lambda_1, \lambda_2, \ldots, \lambda_6} F(\lambda_1, \lambda_2, \ldots, \lambda_6, \rho_1, \rho_2, \ldots, \rho_6). \quad (16)$$

Then from (12) we obtain

$$E(\tilde{\rho}_{[0,T-1]}^{(t)}) \leq \left[ \prod_{t=0}^{T-1} \sigma(\tilde{\rho}_t^{(0)}, M) \prod_{t=0}^{T-1} \sigma(\tilde{\rho}_t^{(1)}, M) \right]^2 \exp \left[ M \sum_{t=0}^{T-1} G(\tilde{\rho}_1^{(0)}, \tilde{\rho}_2^{(0)}, \ldots, \tilde{\rho}_t^{(0)}) \right],$$

or alternatively

$$E(\tilde{\rho}_{[0,T-1]}^{(t)}) \leq \left[ \prod_{t=0}^{T-1} \sigma(\rho_t^{(0)}, M) \prod_{t=0}^{T-1} \sigma(\rho_t^{(1)}, M) \right]^2 \exp \left[ MG(\tilde{\rho}_{[0,T-1]}^{(t)}) \right], \quad (17)$$
where

\[ G(\tilde{\rho}_{[0,T-1]}) = \sum_{t=0}^{T-1} G(\tilde{\rho}_{t-2}, \tilde{\rho}_{t-2}, \ldots, \tilde{\rho}_{t}) \]  

(18)

is a function of the 2T-dimensional normalized weight vector \( \tilde{\rho}_{[0,T-1]} = (\tilde{\rho}_0, \tilde{\rho}_1, \ldots, \tilde{\rho}_{T-1}) \).

We now let \( \tilde{\rho}_{[0,T-1]} = \frac{1}{2T}(\sum_{t=0}^{T-1} \tilde{\rho}_t^{(0)} + \sum_{t=0}^{T-1} \tilde{\rho}_t^{(1)}) \) be the normalized Hamming weight of the vector \( \tilde{\rho}_{[0,T-1]} \). If the function \( G(\tilde{\rho}_{[0,T-1]}) \) is negative for all \( \tilde{\rho}_{[0,T-1]} \) with normalized weight \( \tilde{\rho}_{[0,T-1]} = \rho_0 \), then the corresponding mathematical expectation \( E(\tilde{\rho}_{[0,T-1]}) \) goes to zero exponentially with \( M \) as \( M \) tends to infinity\(^4\).

The total number of weight compositions \( \tilde{d}_{[0,T-1]} = M\tilde{\rho}_{[0,T-1]} \) is upper bounded by \((M+1)^{2T}\). Hence, if \( G(\tilde{\rho}_{[0,T-1]} < 0 \) for a \( \tilde{\rho}_{[0,T-1]} \) having normalized weight \( \tilde{\rho}_{[0,T-1]} = \rho^* \), and \( T \) is finite, the average number of nonzero codewords having weight \( \tilde{d}_{[0,T-1]} \leq 2\rho^*MT \) tends to zero as \( M \) tends to infinity\(^4\).

Note that any code in \( \tilde{C}(3, 6, M, T) \) always has codewords \( \tilde{v}_{[0,T-1]} = (\tilde{v}_0^{(0)}, \tilde{v}_0^{(1)}, \ldots, \tilde{v}_{T-1}^{(0)}, \tilde{v}_{T-1}^{(1)}) \) with \( \tilde{v}_t^{(0)} = \tilde{v}_t^{(1)} = 1 \) for some \( t \), where \( 1 \) is the \( M \)-dimensional all-ones vector, and the remaining \( 2T - 2 \) components of \( \tilde{v}_{[0,T-1]} \) equal to the \( M \)-dimensional all-zero vector. Such a codeword \( \tilde{v}_{[0,T-1]} \) has weight \( 2M \), and this is an upper bound on the minimum distance of codes in \( \tilde{C}(3, 6, M, T) \). Hence, it is sufficient to look at weight compositions with Hamming weight \( \tilde{d}_{[0,T-1]} < 2M \), i.e., \( \rho_{[0,T-1]} < \frac{1}{T} \).

We now summarize the arguments above in the following theorem:

**Theorem 1**: Suppose that the function \( G(\tilde{\rho}_{[0,T-1]}) \) is negative for all \( \tilde{\rho}_{[0,T-1]} \) of normalized weight \( \tilde{\rho}_{[0,T-1]} \leq \rho^* \). Then, in the ensemble \( \tilde{C}(J, K, M, T) \), there exists a TB-LDPCCC with minimum distance \( \tilde{d}_{[0,T-1]} \) lower bounded by

\[ \tilde{d}_{[0,T-1]} \geq \min\{2MT\rho^*, 2M\}. \]  

(19)

The parameter \( \rho^* \) can be calculated numerically. The results of this calculation are presented in Section V. There we will see that, analogous to conventional tail-biting convolutional codes, for relatively small \( T \) the inter minimum distance \( d_{\text{inter}} \) determines the minimum distance of the TB-LDPCCC, while for larger \( T \) the intra minimum distance \( d_{\text{intra}} \), i.e., the free distance of the \( T \)-periodic LDPCCC, determines the minimum distance of the TB-LDPCCC.

\(^4\)See [1] for a mathematically precise formulation of this result.
In the following section, we show how this existence bound on the minimum distance of TB-LDPCCs leads to an existence bound on the free distance of the $T$-periodic LDPCCs.

IV. A LOWER BOUND ON THE FREE DISTANCE OF $T$-PERIODIC LDPCCCs

We begin by considering TB-LDPCCs of length $2\kappa T M$, where $\kappa$ is an integer satisfying $\kappa \geq 1$. Any codeword $\tilde{v}_{[0,\kappa T-1]}$ in this code satisfies

$$\tilde{v}_{[0,\kappa T-1]} \tilde{H}^T_{[0,\kappa T-1]} = 0,$$

where the transposed parity-check matrix $\tilde{H}^T_{[0,\kappa T-1]}$ of the TB-LDCCC is constructed from a syndrome former $H^T_{[0,\kappa T-1]}$ of a $T$-periodic LDCCC by wrapping back the last two blocks of columns (see Section II). The product of a codeword $\tilde{v}_{[0,\kappa T-1]}$ in the length $2\kappa T M$ TB-LDCCC and the syndrome former $H^T_{[0,\kappa T-1]}$ of the $T$-periodic LDCCC defines an $M(\kappa T+2)$-dimensional syndrome vector

$$s_{[0,\kappa T+1]} = \tilde{v}_{[0,\kappa T-1]} H^T_{[0,\kappa T-1]},$$

where the syndrome vector

$$s_{[0,\kappa T+1]} = (s_0, s_1, \ldots, s_{\kappa T+1})$$

is a concatenation of $M$-dimensional subvectors $s_t = (s_{t1}, s_{t2}, \ldots, s_{tM})$, $t = 0, 1, \ldots, \kappa T+1$.

Since $\tilde{v}_{[0,\kappa T-1]}$ satisfies (20) and $\tilde{H}^T_{[0,\kappa T-1]}$ is constructed from $H^T_{[0,\kappa T-1]}$ using the wrapping back procedure of Fig. 2, the subvectors $s_t$ satisfy the conditions

$$s_0 = s_{\kappa T},$$

$$s_1 = s_{\kappa T+1},$$

and

$$s_t = 0, \quad t = 2, 3, \ldots, \kappa T - 1.$$

Therefore (20), which defines the codewords of the TB-LDCCC, is equivalent to (21), where the syndrome vector $s_{[0,\kappa T+1]}$ satisfies conditions (23)–(25).

Lemma 1: Let $\tilde{d}_{[0,\kappa T-1]}$, for any integer $\kappa \geq 1$, be the minimum distance of the TB-LDCCC defined by (20). Then there exists $\kappa_0 > 0$ such that for any $\kappa \geq \kappa_0$ the free distance $d_{\text{free}}$ of the $T$-periodic LDCCC with syndrome former $H^T_{[0,\kappa T-1]}$ is lower bounded by

$$d_{\text{free}} \geq \tilde{d}_{[0,\kappa T-1]}.$$
Proof: The proof follows from the definitions of free distance and row distance. The $L$th order row distance $d_r^L$ of a periodically time-varying convolutional code is defined [11] as the minimum weight of all code sequences having a nonzero segment of length at most $L + m + 1$ (in this case, the code sequences are composed of blocks of length $2M$), where $m$ is the encoder memory. In turn, the free distance $d_{\text{free}}$ is defined as

$$d_{\text{free}} = \min_L d_r^L,$$

where $d_r^L$ is monotonically decreasing with $L$ and there exists an integer $L_0$ such that

$$d_r^{L_0} = d_{\text{free}},$$

for any $L > L_0$.

Thus, we can find a sufficiently large $\kappa_0$ such that, for any $\kappa \geq \kappa_0$, the code sequences of the length $2\kappa TM$ tail-biting code include all possible nonzero segments of length $L_0 + m + 1$ blocks of the $T$-periodic convolutional code. This implies that

$$d_{\text{free},L_0} \geq d_{[0,\kappa T-1]},$$

which, along with (28), leads to (26).

Now let $\tilde{\boldsymbol{v}}_{[0,\kappa T-1]} = (\tilde{\boldsymbol{v}}_{[0,T-1]}, \tilde{\boldsymbol{v}}_{[T,2T-1]}, \ldots, \tilde{\boldsymbol{v}}_{[(\kappa-1)T,\kappa T-1]})$ be a codeword in the length $2\kappa TM$ TB-LDPCCC, i.e., it satisfies (20). Note that this codeword can be represented as

$$\tilde{\boldsymbol{v}}_{[0,\kappa T-1]} = (\tilde{\boldsymbol{v}}_{[0,T-1]}, \tilde{\boldsymbol{v}}_{[T,2T-1]}, \ldots, \tilde{\boldsymbol{v}}_{[(\kappa-1)T,\kappa T-1]}),$$

where

$$\tilde{\boldsymbol{v}}_{[(i-1)T,i T-1]} = (\tilde{\boldsymbol{v}}_{[(i-1)T,0]}, \tilde{\boldsymbol{v}}_{[(i-1)T,1]}, \ldots, \tilde{\boldsymbol{v}}_{[(i-1)T,T-1]}, \tilde{\boldsymbol{v}}_{[iT-1]}), \quad i = 1, \ldots, \kappa.$$

Then consider the sequence

$$\tilde{\boldsymbol{v}}_{[0,T-1]} = \tilde{\boldsymbol{v}}_{[0,T-1]} + \tilde{\boldsymbol{v}}_{[T,2T-1]} + \ldots + \tilde{\boldsymbol{v}}_{[(\kappa-1)T,\kappa T-1]},$$

i.e., the modulo-2 sum of the components of the codeword $\tilde{\boldsymbol{v}}_{[0,\kappa T-1]}$ given in (30). The following lemma proves that $\tilde{\boldsymbol{v}}_{[0,T-1]}$ is a codeword in the TB-LDPCCC consisting of only one period of the $T$-periodic LDPC.

**Lemma 2:** The sequence $\tilde{\boldsymbol{v}}_{[0,T-1]}$ satisfies

$$\tilde{\boldsymbol{v}}_{[0,T-1]} \tilde{\boldsymbol{H}}^{T}_{[0,T-1]} = \mathbf{0},$$

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i.e., it is a code sequence of the TB-LDPCCC with length equal to one period of the $T$-periodic LDPCCC.

**Proof:** Equality (33) is equivalent (see (21)) to the equality

$$ \tilde{v}_{[0,T-1]} H_T^{[0,T-1]} = s_{[0,T+1]} = (s_0, s_1, \ldots, s_{T+1}), \quad (34) $$

where

$$ s_0 = s_T, \quad (35) $$

$$ s_1 = s_{T+1}, \quad (36) $$

and

$$ s_t = 0, \quad t = 2, 3, \ldots, T - 1. \quad (37) $$

Therefore, instead of proving (33), we can alternatively prove that the syndrome vectors in (34) satisfy conditions (35)–(37). We begin by defining

$$ \tilde{v}_{[(i-1)T,iT-1]} H_T^{[0,T-1]} \overset{\text{def}}{=} s_{[0,T+1]}^{(i)} = (s_0^{(i)}, s_1^{(i)}, \ldots, s_{T+1}^{(i)}), \quad i = 1, 2, \ldots, \kappa. \quad (38) $$

Then it follows from (21) and (38) that the syndrome vector $s_{[0,\kappa T+1]}$ of the length $2\kappa TM$ TB-LDPCCC satisfies

$$ (s_0, s_1, \ldots, s_{\kappa T+1}) = (s_0^{(1)}, s_1^{(1)}, \ldots, s_{T-1}^{(1)}, s_T^{(1)}, s_{T+1}^{(1)}, 0, 0, \ldots, 0, 0, \ldots) $$

$$ + (0, 0, \ldots, 0, s_0^{(2)}, s_1^{(2)}, s_2^{(2)}, \ldots, s_{T-1}^{(2)}, s_T^{(2)}, s_{T+1}^{(2)}, 0, \ldots, ) $$

$$ + (0, 0, \ldots, 0, 0, \ldots, 0, 0, 0, 0, s_0^{(3)}, s_1^{(3)}, s_2^{(3)}, \ldots, s_{T+1}^{(3)}, 0, \ldots) $$

$$ + \ldots, \quad (39) $$

where the addition is modulo 2.

Since $\tilde{v}_{[0,\kappa T-1]}$ ia a codeword in the length $2\kappa TM$ TB-LDPCCC, conditions (23)–(25) are satisfied for the syndrome vector $s_{[0,\kappa T+1]}$. This, together with (39), implies that the vectors $s_t^{(i)}$ should satisfy

$$ s_0^{(1)} = s_T^{(\kappa)}, \quad (40) $$

$$ s_1^{(1)} = s_{T+1}^{(\kappa)}, \quad (41) $$

$$ s_0^{(i)} = s_T^{(i-1)}, \quad i = 2, 3, \ldots, \kappa, \quad (42) $$

$$ s_1^{(i)} = s_{T+1}^{(i-1)}, \quad i = 2, 3, \ldots, \kappa, \quad (43) $$

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From the definition of \( \vec{v}_{[0,T-1]} \) in (32), the definition of its syndrome vector \( s_{[0,T+1]}^{(i)} \) in (34), and the definition of \( s_{[0,T+1]}^{(i)} \) in (38), it follows that

\[
\vec{s}_{[0,T+1]} = s_{[0,T+1]}^{(1)} + s_{[0,T+1]}^{(2)} + \cdots + s_{[0,T+1]}^{(\kappa)},
\]

where the addition is modulo 2. It can be checked that (40)-(45) imply (35)-(37), and hence the lemma is proved.

The next lemma relates the minimum distances of the length \( 2\kappa TM \) TB-LDPCCC and the length \( 2TM \) TB-LDPCCC constructed from the same \( T \)-periodic LDPCCC.

**Lemma 3:** For any positive integer \( \kappa \), the minimum distance \( \tilde{d}_{[0,\kappa T-1]} \) of the length \( 2\kappa TM \) TB-LDPCCC defined by (20) is lower bounded by the minimum distance \( \tilde{d}_{[0,T-1]} \) of the length \( 2TM \) TB-LDPCCC, i.e., \( \tilde{d}_{[0,\kappa T-1]} \geq \tilde{d}_{[0,T-1]} \).

**Proof:** For any codeword \( \tilde{\vec{v}}_{[0,\kappa T-1]} \) in the length \( 2\kappa TM \) TB-LDPCCC represented by (30),

\[
w_H (\tilde{\vec{v}}_{[0,\kappa T-1]}) = w_H (\tilde{\vec{v}}_{[0,T-1]}) + w_H (\tilde{\vec{v}}_{[T,2T-1]}) + \cdots + w_H (\tilde{\vec{v}}_{[(\kappa-1)T,T-1]}) ,
\]

where \( w_H(\cdot) \) is the Hamming weight operator. From (32) and the triangle inequality it then follows that

\[
w_H (\tilde{\vec{v}}_{[0,T-1]}) \leq w_H (\tilde{\vec{v}}_{[0,T-1]}) + \cdots + w_H (\tilde{\vec{v}}_{[(\kappa-1)T,T-1]}) .
\]

Since from Lemma 2 we know that \( \tilde{\vec{v}}_{[0,T-1]} \) is a codeword in the length \( 2TM \) TB-LDPCCC, it follows that

\[
\tilde{d}_{[0,T-1]} \leq w_H (\tilde{\vec{v}}_{[0,T-1]}) .
\]

Combining (46)–(48) we obtain for the codeword \( \tilde{\vec{v}}_{[0,\kappa T-1]} \) of the length \( 2\kappa TM \) TB-LDPCCC that

\[
\tilde{d}_{[0,T-1]} \leq w_H (\tilde{\vec{v}}_{[0,\kappa T-1]}) ,
\]

which directly implies

\[
\tilde{d}_{[0,T-1]} \leq \tilde{d}_{[0,\kappa T-1]} ,
\]

and the lemma is proved.

Theorem 2 now follows directly from Lemmas 1 and 3.
Theorem 2: The free distance $d_{\text{free}}$ of any $T$-periodic LDPCCC from the ensemble $C(3, 6, M, T)$ is lower bounded by the minimum distance of the corresponding TB-LDPCCC of block length $2TM$, i.e.,

$$d_{\text{free}} \geq \tilde{d}_{[0, T-1]}.$$  \hspace{1cm} (51)

The distance properties of convolutional codes are characterized by the ratio $d_{\text{free}}/\nu$. In Section V, we use Theorems 1 and 2 to numerically calculate the lower bound $\alpha_{\text{LDPCCC}} = \tilde{d}_{[0, T-1]}/\nu$ on $d_{\text{free}}/\nu$.

V. Numerical Results

Consider the function $G(\tilde{\rho}_{[0, T-1]})$ defined in Section III. According to the condition of Theorem 1, we must find the maximum $\rho^*_{[0, T-1]}$ such that $G(\tilde{\rho}_{[0, T-1]}) < 0$ for all $\tilde{\rho}_{[0, T-1]}$ with normalized weight $\tilde{\rho}_{[0, T-1]} \leq \rho^*_{[0, T-1]}$.

The numerical procedure used to find $\rho^*_{[0, T-1]}$ can be outlined as follows. For a given $\tilde{\rho}_{[0, T-1]}$, we find the vector $\tilde{\rho}^{*}_{[0, T-1]}$ with normalized weight $\tilde{\rho}^{*}_{[0, T-1]}$ that maximizes the function $G(\tilde{\rho}_{[0, T-1]})$ using numerical optimization. If $G(\tilde{\rho}_{[0, T-1]}) > 0$, we decrease $\tilde{\rho}_{[0, T-1]}$; otherwise (i.e., if $G(\tilde{\rho}_{[0, T-1]}) < 0$) we increase $\tilde{\rho}_{[0, T-1]}$ and repeat the procedure until the function $G(\tilde{\rho}_{[0, T-1]})$ becomes less than, for example, $-10^{-5}$. The end result of the procedure is a vector $\tilde{\rho}^{*}_{[0, T-1]} = (\tilde{\rho}^{(0)}, \tilde{\rho}^{(1)}, \ldots, \tilde{\rho}^{(0)}_{T-1}, \tilde{\rho}^{(1)}_{T-1})$ and its normalized weight $\rho^*_{[0, T-1]} = \frac{1}{2T}(\sum_{t=0}^{T-1} \tilde{\rho}^{(0)}_{t} + \sum_{t=0}^{T-1} \tilde{\rho}^{(1)}_{t})$.

In Fig. 3, the resulting maximizing vectors $\tilde{\rho}^{*}_{[0, T-1]}$ are shown for $T = 5, 12, 13$. In particular, for $T = 5$, we have

$$\tilde{\rho}^{*}_{[0, T-1]} = (0.023, 0.023, \ldots, 0.023).$$  \hspace{1cm} (52)

The normalized weight $\tilde{\rho}_{[0, T-1]} = 0.023$ of this vector is close to Gallager’s lower bound on the minimum distance to block length ratio of LDPC block codes derived in [2]. This confirms the general rule that, for relatively small block lengths, the inter minimum distance $d_{\text{inter}}$, which reflects the block code properties of tail-biting convolutional codes, determines their minimum distance. Indeed, the same tendency is observed for $T = 3, 4, \ldots, 11$, i.e., each component of the maximizing vector is approximately equal to 0.023, resulting in a normalized weight of 0.023.

For $T = 11$, we see from (52) that the sum of the components of the maximizing vector $\tilde{\rho}^{*}_{[0, T-1]}$ is approximately 0.5. The same is also true for $T = 12$ and $T = 13$ (even though as shown in...
Figs. 3 (b) and 3 (c), the components of the maximizing vector are no longer approximately equal in these cases), and further increases in the period do not lead to higher values of this sum. It follows that the normalized weight drops as $T$ increases beyond 11. This effect is observed due to the intra minimum distance $d_{\text{intra}}$, which is lower bounded by the bound on free distance of LDPCCCs derived in [1]. This bound scales as $\alpha_{\text{LDPCCC}(3, 6)\nu} = 6\alpha_{\text{LDPCCC}(3, 6)M} \approx 0.5M$ and represents an upper bound on the minimum distance of TB-LDPCCCs. In other words, the minimum distance to block length ratio of TB-LDPCCCs decreases as $T$ increases beyond 11, since the block length continues to increase while the minimum distance cannot grow beyond the constant $d_{\text{intra}}$. 
The distance ratios for the various LDPCCC ensembles are presented in Fig 4. We see that the minimum distance to block length ratio for TB-LDPCCCs is equal to Gallager’s ratio for LDPC block codes when the period $T$ is small. For larger periods, however, the ratio drops and tends to zero, due to the effect of $d_{\text{intra}}$, as noted above. On the other hand, the free distance to constraint length ratio for $T$-periodic LDPCCCs grows with increasing $T$ and approaches the ratio derived in [1] for general (non-periodic) time-varying LDPCCCs as $T$ increases beyond 11.

![Fig. 4. Distance ratios for $T$-periodic LDPCCCs and TB-LDPCCCs as a function of the period $T$.]

**VI. CONCLUSIONS**

In this paper, we derived a lower bound on the free distance of periodically time-varying $(J, K)$ regular LDPCCCs and a lower bound on the minimum distance of the associated TB-LDPCCCs. Theorems 1 and 2 give analytical expressions for these bounds in the general case. Using these expressions, we calculated numerically the bounds on free distance and minimum distance for the practically interesting $(3, 6)$ regular LDPCCC case. In the limiting cases, for $T > 11$ the free distance bound corresponds to the bound for general (non-periodic) time-varying...
LDPCCCs derived in [1], and for $T = 3$ the minimum distance bound corresponds to Gallager’s bound for LDPC block codes.

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