Spectral Shape of Doubly-Generalized LDPC Codes: Efficient and Exact Evaluation

Mark F. Flanagan, Senior Member, IEEE, Enrico Paolini, Member, IEEE, Marco Chiani, Fellow, IEEE, and Marc P. C. Fossorier, Fellow, IEEE

Abstract—This paper analyzes the asymptotic exponent of the weight spectrum for irregular doubly-generalized LDPC (D-GLDPC) codes. In the process, an efficient numerical technique for its evaluation is presented, involving the solution of a $4 \times 4$ system of polynomial equations. The expression is consistent with previous results, including the case where the normalized weight or stopping set size tends to zero. The spectral shape is shown to admit a particularly simple form in the special case where all variable nodes are repetition codes of the same degree, a case which includes Tanner codes; for this case it is also shown how certain symmetry properties of the local weight distribution at the CNs induce a symmetry in the overall weight spectral shape function. Finally, using these new results, weight and stopping set size spectral shapes are evaluated for some example generalized and doubly-generalized LDPC code ensembles.

Index Terms—Doubly-generalized LDPC codes, irregular code ensembles, spectral shape, stopping set size distribution, weight distribution.

I. INTRODUCTION

Recently, the design and analysis of coding schemes representing generalizations of Gallager’s low-density parity-check (LDPC) codes [1] has gained increasing attention. This interest is motivated above all by the search for coding schemes which offer a better compromise between waterfall and error floor performance than is currently offered by state-of-the-art LDPC codes.

In the Tanner graph of an LDPC code, any degree-$q$ variable node (VN) may be interpreted as a length-$q$ repetition code, i.e., as a $(q, 1)$ linear block code. Similarly, any degree-$s$ check node (CN) may be interpreted as a length-$s$ single parity-check (SPC) code, i.e., as a $(s, s - 1)$ linear block code. The first proposal of a class of linear block codes generalizing LDPC codes may be found in [2], where it was suggested to replace each CN of a regular LDPC code with a generic linear block code, to enhance the overall minimum distance. The corresponding coding scheme is known as a regular generalized LDPC (GLDPC) code, or Tanner code, and a CN that is not a SPC code as a generalized CN. More recently, irregular GLDPC codes were considered (see for instance [3]). For such codes, the VNs exhibit different degrees and the CN set is composed of a mixture of different linear block codes.

A further generalization step is represented by doubly-generalized LDPC (D-GLDPC) codes [4]. In a D-GLDPC code, not only the CNs but also the VNs may be represented by generic linear block codes. The VNs which are not repetition codes are called generalized VNs. The main motivation for introducing generalized VNs is to overcome some problems connected with the use of generalized CNs, such as an overall code rate loss which makes GLDPC codes interesting mainly for low code rate applications, and a loss in terms of decoding threshold (for a discussion on drawbacks of generalized CNs and on beneficial effects of generalized VNs we refer to [5] and [6], respectively).

A useful tool for analysis and design of LDPC codes and their generalizations is represented by the asymptotic exponent of the weight distribution. As usual in the literature, this exponent will be referred to as the growth rate of the weight distribution or the weight spectral shape of the ensemble, the two expressions being used interchangeably throughout this paper. The growth rate of the weight distribution was introduced in [1] to show that the minimum distance of a randomly generated regular LDPC code with a VN degree of at least three is a linear function of the codeword length with high probability. The same approach was taken in [7] and [8] to obtain related results on the minimum distance of subclasses of Tanner codes.

The growth rate of the weight distribution has been subsequently investigated for unstructured ensembles of irregular LDPC codes. Works in this area are [9]–[12]. In particular, in [12] a technique for evaluation of the growth rate of any (eventually expurgated) irregular LDPC ensemble has been developed, based on Hayman’s formula. The nonbinary weight distribution of nonbinary LDPC codes was analyzed in [13] and [14], while the binary weight distribution of nonbinary LDPC codes was derived in [15] and [16]. Asymptotic weight enumerators of ensembles of irregular LDPC codes based on protographs and on multiple edge types have been derived in [17] and [18], [19], respectively. The approach proposed in [17] has then been extended to protograph GLDPC codes and to protograph D-GLDPC codes in [20] and [21], respectively. In contrast to the present work, the evaluation of the weight
enumerators in [17]–[21] require numerical solution of a high-dimensional optimization problem.

In this paper, an analytical expression for the growth rate of the weight distribution of a general unstructured irregular ensemble of D-GLDPC codes is developed. The present work also extends to the fully-irregular case an expression for the growth rate obtained in [22] assuming a CN set composed of linear block codes all of the same type. In the process of this development, we obtain an efficient evaluation tool for computing the growth rate exactly. This tool always requires the solution of a $4 \times 4$ polynomial system of equations, regardless of the number of VN types and CN types in the D-GLDPC ensemble. As shown through numerical examples, the derived result allows us to evaluate the weight distribution upon replacing the local WEF of each CN with an appropriate polynomial function. This also allows for a reasonably large computational effort. The derived result may be regarded as a generalization to the D-GLDPC case of the rate with a low computational effort. The derived result may be regarded as a generalization to the D-GLDPC case of the rate with a low computational effort. This also allows for a reasonably large computational effort. The derived result may be regarded as a generalization to the D-GLDPC case of the rate with a low computational effort.

The paper is organized as follows. Section II defines the D-GLDPC ensemble of interest, and introduces some definitions and notation pertaining to this ensemble. Section III presents the main result of the paper regarding the evaluation of the weight and stopping set size spectral shapes. Section IV derives the spectral shapes of a family of check-hybrid GLDPC code ensembles as a corollary to the main result, and also identifies a sufficient condition for symmetry of the weight spectral shape for such ensembles. Section V provides a proof of the main result of the paper, and Section VI provides additional proofs for other results in the paper. Section VII provides some examples of spectral shapes of GLDPC and D-GLDPC codes, and Section VIII concludes the paper.

II. PRELIMINARIES AND NOTATION

A D-GLDPC code consists of a set of CNs and a set of VNs. Each of these nodes is associated with a linear ‘local’ code, and with each VN is also associated an encoder (i.e., generator matrix). Each node is equipped with a set of sockets corresponding to the bits of the local codeword. The VN sockets and the CNs sockets are connected together by edges in a one-to-one fashion; the resulting graph is called the Tanner graph of the code. A codeword in such a D-GLDPC code is defined as an assignment of values to the local information bits of each VN such that the corresponding bit values induced on the Tanner graph edges (through local encoding at the VNs) cause each CN to recognize a valid local codeword. An illustration of a simple D-GLDPC code is given in Figure II together with its codeword [1 0 1 0 1 0 0 1]. We next define a sequence $\{M_n\}$ of D-GLDPC code ensembles; many of the following definitions and notations also appear in [23].

In the D-GLDPC code ensemble $M_n$, the number of VNs is denoted by $n$. The different CN types are denoted by the set $I_0 = \{1, 2, \ldots, n_c\}$; the local code $C_t$ of CN type $t \in I_c$ has dimension, length and minimum distance denoted by $n_t$, $s_t$ and $r_t$, respectively. Similarly, the different VN types are denoted by $I_v = \{1, 2, \ldots, n_v\}$; the local code $C_v$ of VN type $t \in I_v$ has dimension, length and minimum distance denoted by $h_t$, $q_t$ and $p_t$ respectively. We assume that the local codes associated with all VNs and CNs have minimum distance at least 2 (i.e., $r_t \geq 2, p_t \geq 2$), and that the dual codes of all these local codes have minimum distance greater than one.

For $t \in I_c$, the fraction of edges connected to CNs of type $t$ is denoted by $p_t$. Similarly, for $t \in I_v$, the fraction of edges connected to VNs of type $t$ is denoted by $h_t$. CN and VN type-distribution polynomials are then given by $\rho(x)$ and $\lambda(x)$ respectively, where $\rho(x) := \sum_{t \in I_c} p_t x^{n_t-1}$ and $\lambda(x) := \sum_{t \in I_v} h_t x^{q_t-1}$. If $E$ denotes the number of edges in the Tanner graph, the number of CNs of type $t \in I_c$ is then given by $E \rho_t / s_t$, and the number of VNs of type $t \in I_v$ is then given by $E \lambda_t / q_t$. Denoting as usual $\int_0^1 \rho(x) \, dx$ and $\int_0^1 \lambda(x) \, dx$ by $\rho$ and $\lambda$ respectively, the number of edges in the Tanner graph is given by $E = n / \int \lambda$ and the number of CNs is given by $m = E / \rho$. Therefore, the fraction of CNs

\[ \lambda(x) := \sum_{t \in I_v} h_t x^{q_t-1} \]

An equivalent condition is that the generator matrix of each code $C_t$ has no column consisting entirely of zeros.
of type \( t \in I_c \) and the fraction of VNs of type \( t \in I_v \) are given by
\[
\gamma_t = \frac{\rho_t}{s_t} \quad \text{and} \quad \delta_t = \frac{\lambda_t}{q_t}
\]
respectively. A code in the irregular D-GLDPC ensemble corresponds to a permutation of the \( E \) edges connecting VNs to CNs. The length of any D-GLDPC codeword in the ensemble is given by
\[
N = \sum_{t \in I_c} \left( \frac{E \lambda_t}{q_t} \right) k_t = \frac{n}{\lambda} \sum_{t \in I_v} \lambda_t k_t.
\]
Note that this is a linear function of \( n\). Similarly, the total number of parity-check equations for any D-GLDPC code in the ensemble is given by \( M = \frac{n}{\lambda} \sum_{t \in I_v} \rho_t (s_t - k_t) \).

The ensemble \( \mathcal{M}_n \) is defined according to a uniform probability distribution on all \( \mathcal{E}! \) permutations of the Tanner graph edges. The design rate of the D-GLDPC ensemble is given by
\[
R = 1 - \frac{\sum_{t \in I_c} \rho_t (1 - R_t)}{\sum_{t \in I_v} \lambda_t R_t}
\]
where for \( t \in I_c \) (resp. \( t \in I_v \)), \( R_t \) is the local code rate of a type-\( t \) CN (resp. VN). Each code in the ensemble has a code rate larger than or equal to \( R \).

The WEF for CN type \( t \in I_c \) is given by
\[
A^{(t)}(z) = 1 + \sum_{u=r_t}^{s_t} A^{(t)}_u z^u.
\]
Here, for each \( 0 \leq u \leq s_t \), \( A^{(t)}_u \geq 0 \) denotes the number of weight-\( u \) codewords for CNs of type \( t \). We denote by \( \bar{u}_t \) the maximal weight of a codeword in the local code for CN type \( t \); this is the largest \( u \in \{ r_t, r_t + 1, \ldots, s_t \} \) such that \( A^{(t)}_u > 0 \).

The input-output weight enumerating function (IO-WEF) for VN type \( t \in I_v \) is given by
\[
B^{(t)}(x, y) = 1 + \sum_{u=1}^{k_t} \sum_{v=1}^{q_t} B^{(t)}_{u,v} x^u y^v.
\]
Here \( B^{(t)}_{u,v} \geq 0 \) denotes the number of weight-\( v \) codewords generated by input words of weight \( u \), for VNs of type \( t \). Also, \( B^{(t)}_v \) is the total number of weight-\( v \) codewords for VNs of type \( t \).

Although this paper is focused on the weight spectrum, all of the results developed in Sections III-V can be extended to the stopping set size spectrum. A stopping set of a D-GLDPC code may be defined as any subset \( S \) of the code bits such that, assuming all code bits in \( S \) are erased and all code bits not in \( S \) are not erased, local erasure decoding at the CNs and VNs cannot recover any code bit in \( S \), so no erasure can be recovered by iterative decoding. A local stopping set for a CN is a subset of the local code bits which, if erased, is not recoverable to any extent by the CN. A local stopping set for a VN is a subset of the local code bits together with a subset of the local information bits which, if both subsets are erased, are not recoverable to any extent by the VN. All results derived in this paper for the weight spectrum can be extended to the stopping set size spectrum by simply replacing the WEF for CN type \( t \in I_c \) with its local stopping set enumerating function (SSEF), and replacing the IO-WEF for VN type \( t \in I_v \) with its local input-output stopping set enumerating function (IO-SSEF).

We point out that for both VNs and CNs, the local SSEF depends on the decoding algorithm used to locally recover from erasures; for local maximum \emph{a posteriori} (MAP) decoding, the reader is referred to [23, Appendix B] for details.

Finally, we introduce some mathematical notation. Let \( a(n) \) and \( b(n) \) be two real-valued sequences, where \( b(n) \neq 0 \) for all \( n \); we say that \( a(n) \) is exponentially equivalent to \( b(n) \) as \( n \to \infty \), writing \( a(n) \asymp b(n) \), if and only if
\[
\lim_{n \to \infty} n^{-1} \log \left( a(n)/b(n) \right) = 0.
\]
Throughout this paper, the notation \( e = \exp(1) \) denotes Napier’s number, all the logarithms are assumed to have base \( e \) and for \( 0 < x < 1 \) the notation \( h(x) = -x \log(x) - (1-x) \log(1-x) \) denotes the binary entropy function expressed in nats.

### III. Weight Spectral Shape of Irregular D-GLDPC Code Ensembles

The weight spectral shape of the irregular D-GLDPC ensemble sequence \( \{M_n\} \) is defined by
\[
G(\alpha) := \lim_{n \to \infty} \frac{1}{k_t} \log \mathbb{E}_{M_n}[A_{wn}] \quad (4)
\]
where \( \mathbb{E}_{M_n} \) denotes the expectation operator over the ensemble \( M_n \), and \( A_{wn} \) denotes the number of codewords of weight \( w \) of a randomly chosen D-GLDPC code in the ensemble \( M_n \).

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2The concept of stopping set was first introduced in [24] in the context of LDPC codes. When applied to LDPC codes (i.e., all CNs are SPC codes), the definition of stopping set used in this paper coincides with that in [24].
The limit in (4) assumes the inclusion of only those positive integers \( n \) for which \( \alpha n \in \mathbb{Z} \) and \( E_{M_\alpha}[A_{\alpha n}] \) is positive. Note that the argument of the growth rate function \( G(\alpha) \) is equal to the ratio of D-GLDPC codeword weight to the number of VN; by \( \delta \), this captures the behaviour of codewords whose weight is linear in the block length.

Using standard notation [11], we define the critical exponent codeword weight ratio for \( M_\alpha \) as \( \alpha^* := \inf\{\alpha > 0 \mid G(\alpha) \geq 0\} \). A D-GLDPC ensemble is said to have good growth behavior if \( \alpha^* > 0 \), and is said to have bad growth behavior if \( \alpha^* = 0 \). In [23], it was shown that a D-GLDPC ensemble always has good growth rate behavior if there exist no CNs or VNs with minimum distance 2, while, if there exist both CNs and VNs with minimum distance 2, the ensemble has good growth rate behavior if and only if \( C \cdot V < 1 \), where the (positive) parameters \( C \) and \( V \) are given by

\[
C = 2 \sum_{t : \rho t = 2} \frac{p_t A(t)}{s_t} \quad V = 2 \sum_{t : \rho t = 2} \frac{\lambda_t B(t)}{q_t}.
\]

Note that using (2), we may also define the growth rate with respect to the D-GLDPC code’s block length \( N \) as follows:

\[
H(\omega) := \lim_{N \to \infty} \frac{1}{N} \log E_{M_\omega}[A_{\omega N}] \quad (6)
\]

Note that we must have \( \omega \leq 1 \), while in general we may have \( \alpha > 1 \). It is straightforward to show that

\[
H(\omega) = \frac{G(K_s, \omega)}{K_s} \quad (7)
\]

where \( K_s \) is defined as the ratio of the D-GLDPC code’s block length to the number of VNs, i.e.,

\[
K_s := \frac{N}{n} = \frac{1}{\int \lambda} \sum_{t \in I_c} \lambda_t k_t q_t \quad (8)
\]

Note that the parameter \( K_s \) is independent of \( N \).

The stopping set size spectral shapes of the ensemble sequence \( \{M_\alpha\} \) for the case of bounded distance (BD) and MAP decoding at the CNs, whose definitions are analogous to [4], will be denoted by \( G_\psi(\alpha) \) and \( G_\Phi(\alpha) \), respectively. Similarly the critical exponent stopping set size ratio will be denoted in these cases by \( \alpha^*_\psi \) and \( \alpha^*_\Phi \), respectively.

In this section, we formulate an expression for the growth rate for an irregular D-GLDPC ensemble \( M_\alpha \) over a wider range of \( \alpha \) than was considered in [23] (where the case of small \( \alpha \) was analyzed).

The following theorem constitutes the main result of this paper.

**Theorem 3.1:** The weight spectral shape of the irregular D-GLDPC ensemble sequence \( \{M_\alpha\} \) is given by

\[
G(\alpha) = \sum_{t \in I_c} \delta_t \log B(t)(x_0, y_0) - \alpha \log x_0 + \left( \frac{\int \rho}{\int \lambda} \right) \sum_{s \in I_c} \gamma_s \log A(s)(z_0) + \frac{\log (1 - \beta \int \lambda)}{\int \lambda} \quad (9)
\]

where \( x_0, y_0, z_0 \) and \( \beta \) are the unique positive real solutions to the \( 4 \times 4 \) system of polynomial equations

\[
\begin{align*}
&z_0 \left( \frac{\int \rho}{\int \lambda} \right) \sum_{t \in I_c} \gamma_t A(t)(z_0) = \beta, \\
&x_0 \sum_{t \in I_c} \delta_t B(t)(x_0, y_0) = \alpha, \\
&y_0 \sum_{t \in I_c} \delta_t B(t)(x_0, y_0) = \beta, \\
&\left( \frac{\beta}{\int \lambda} \right) (1 + y_0 z_0) = y_0 z_0.
\end{align*}
\]

This theorem is proved in Section V. It is important to note that the solution always involves a system of 4 equations in 4 unknowns, regardless of the number of different CN and VN types. Also, note that we can solve efficiently for the parameter \( \alpha^* \) without evaluating the entire spectral shape, by simply augmenting the system (10)–(13) with an additional equation which sets the right-hand side of (9) to zero. Note also that substituting for \( \beta \) from (13) will further reduce the system to 3 equations in 3 unknowns. The reader may verify that in the special case of LDPC codes, this result reduces to [12, Corollary 12].

We point out that Theorem 3.1 is consistent with Theorem 4.1 of [23] which provides an expression for \( G(\alpha) \) valid for small \( \alpha \). This can be seen by conducting an analysis of (9)–(13) for the small-\( \alpha \) case. This analysis, detailed in Appendix B, yields the following slightly weaker version of Theorem 4.1 of [23] as a corollary of Theorem 3.1.

**Corollary 3.2:** The weight spectral shape of the irregular D-GLDPC ensemble sequence \( \{M_\alpha\} \) is given by

\[
G(\alpha) = T \psi \log \alpha + \alpha \left[ \frac{\log Q_1^{-1}(1)}{\log Q_2(Q_1^{-1}(1))} + o(\alpha) \right], \quad (14)
\]

where \( r \) denotes the smallest minimum distance over all CN types, \( \psi = r/(r-1) \), and \( T \) denotes the minimum of \( (j-\psi)/i \) over all types \( i \in I_v \) and all pairs \((i, j)\) such that \( B_{ij}(t) > 0 \). The set \( Y_i \) is the set of all types \( t \in I_v \) such that this minimum \( T \) is achieved, and \( P_i \) denotes the corresponding set of pairs \((i, j)\). Finally,

\[
Q_1(x) = \sum_{t \in Y_i} \frac{\lambda_t}{q_t} \sum_{(i,j) \in P_i} j B_{ij}(t) C^{i/j} \left( \frac{f \lambda}{e} \right)^{iT/\psi} x^i \quad (15)
\]

Note that while (10), (13), and (23) are not polynomial as set down here, each may be made polynomial by multiplying across by an appropriate factor.

The result is slightly weaker in the following sense: denoting the left-hand side of (14) by \( F(\alpha) + o(\alpha) \), Corollary 3.2 proves that \( \lim_{\alpha \to 0} [G(\alpha) - F(\alpha)]/\alpha = 0 \). In contrast, Theorem 4.1 of [23] proves that \( |G(\alpha) - F(\alpha)| \leq K \cdot \alpha^2 \), where \( K \) is independent of \( \alpha \) and \( \xi > 0 \) is a known parameter depending on the ensemble. Thus, Theorem 4.1 of [23] provides a stronger statement regarding the rate of convergence of \( G(\alpha) \) to \( F(\alpha) \) as \( \alpha \to 0 \).
and
\[ Q_2(x) = \sum_{t \in E_v} \lambda_t \sum_{(i,j) \in P_t} iB_{ijr}^{(t)} \left( \frac{f(x)}{e} \right)^{T/\psi} x^i, \]
where
\[ C = r \sum_{t : r_t = r} \frac{\rho t A(t)}{s_t} > 0 \]
(this definition of the parameter \( C \) reduces to that given in \[ \text{(5)} \) in the case \( r = 2 \).

When the term \((T/\psi) \alpha \log \alpha \) is nonzero, which happens when either \( r \geq 3 \) or \( p \geq 3 \) where \( p \) is the smallest minimum distance over all VN types, \[ \text{(10)–(13)} \] may be directly used to obtain an approximation to the parameter \( \alpha^* \) without solving the polynomial system in Theorem \[ \text{[3.1]} \) – this approximation is in general very good for ensembles characterized by sufficiently small \( \alpha^* \). To obtain this, we set \( G(\alpha^*) = 0, \alpha^* \neq 0, \) and neglect the \( o(\alpha) \) term in \[ \text{(14)}, yielding \]
\[ \alpha^* \approx [Q_1^{-1}(1)]^{\frac{\lambda}{\lambda}} \cdot Q_2(Q_1^{-1}(1)). \]

For an irregular GLDPC code ensemble, this approximation reduces to
\[ \alpha^* \approx \lambda_p r/(pr-r-p) C^{-p/(pr-r-p)} \frac{e}{p \int \lambda} \]
where \( \lambda_p \) is the fraction of edges connected to the degree-\( p \) VN, i.e., to the VN of lowest degree, and where \( C \) is defined in \[ \text{(17)} \). If the GLDPC code ensemble is variable-regular, \[ \text{(19)} \) reduces to
\[ \alpha^* \approx C^{-p/(pr-r-p)} e. \]

An even simpler expression is obtained for regular LDPC code ensembles of VN degree \( p \geq 3 \) (as \( r = 2 \) for LDPC codes), for which \[ \text{(20)} \) becomes
\[ \alpha^* \approx \frac{e}{(d_c - 1)^{(1 - 2/d_c)}} \]
where \( d_v = p \) is the VN degree and \( d_c \) is the CN degree. Some numerical results on this approximation will be presented in Section \[ \text{VII} \].

**Lemma 3.3**: The derivative of the weight spectral shape of the irregular D-GLDPC ensemble sequence \( \{ \mathcal{M}_n \} \) is given by
\[ G'(\alpha) = -\log x_0, \]
where, for any \( \alpha, x_0 \) is given by the solution to the system of equations \[ \text{(10)–(13)} \). It follows that the stationary points of \( G(\alpha) \) occur at exactly those values of \( \alpha \) for which the solution to \[ \text{(10)–(13)} \) satisfies \( x_0 = 1 \).

**Proof**: Note that in the solution for the weight spectral shape function given by Theorem \[ \text{[3.1]} \) each of the parameters \( x_0, y_0, z_0 \) and \( \beta \) can be regarded as an implicit function of \( \alpha \). Hence, differentiating \[ \text{(9)} \) directly and using \[ \text{(10), (11)} \) and \[ \text{(12)} \) yields
\[ G'(\alpha) = -\log x_0 + \frac{\beta z_0}{z_0 \frac{d \log (y_0 z_0)}{d \alpha}} - \frac{1}{1 - \beta f \lambda \frac{d \beta}{d \alpha}} - \frac{y_0 z_0}{\beta f \lambda \frac{d \beta}{d \alpha}}, \]
where we have used \[ \text{(13)} \) in the final line. It remains to prove that the final pair of terms in this expression sum to zero. To show this, we write \[ \text{(13)} \) as
\[ \log (\beta f \lambda) + \log (1 + y_0 z_0) = \log(y_0 z_0) \]
Differentiating this expression yields
\[ \frac{1}{\beta f \lambda} \frac{d \beta}{d \alpha} = \frac{1}{y_0 z_0 (1 + y_0 z_0)} \frac{d [y_0 z_0]}{d \alpha} \]
and so
\[ \frac{y_0 z_0}{\beta f \lambda} \frac{d \beta}{d \alpha} = \frac{1}{(1 + y_0 z_0) f \lambda} \frac{d [y_0 z_0]}{d \alpha} = \frac{\beta}{y_0 z_0} \frac{d [\log(y_0 z_0)]}{d \alpha} \]
where in the final line we have again used \[ \text{(13)} \). This establishes the result of the Lemma.

**Theorem 3.4**: For any D-GLDPC code ensemble, the weight spectral shape \( H(\omega) \) as defined in \[ \text{(7)} \) has a stationary point at \( \omega = 1/2 \), and the function value at this point is given by \( H(1/2) = R log 2 \).

**Proof**: Note that (using \[ \text{(7)} \)) it is equivalent to show that the weight spectral shape \( G(\alpha) \) of Theorem \[ \text{[3.1]} \) has a stationary point at \( \alpha = K_s/2 \) at which point \( G(\alpha) = K_s R log 2 \). We will show that
\[ \alpha = \frac{K_s}{2}; \ x_0 = y_0 = z_0 = 1; \ \beta = \frac{1}{2 f \lambda}, \]
where \[ \text{(19)} \) makes \[ \text{(10)–(13)} \) a consistent system of equations. This solution has the property that \( x_0 = 1 \), and therefore by Lemma \[ \text{[3.3]} \) a stationary point of the weight spectral shape exists at \( G(K_s/2) = K_s R log 2 \) (or equivalently at \( H(1/2) = R log 2 \)).

First, it is easy to check that \[ \text{(13)} \) holds under \[ \text{(22)} \). Making the substitutions \[ \text{(22)} \) in \[ \text{(10)} \) yields
\[ \frac{(\beta \rho)}{\sum_{t \in I'} \gamma_t} \frac{\partial A_t^{(1)}}{(A_t^{(1)})^2} = \frac{1}{2}. \]
Note that \( A_t^{(1)} = \sum_{u=0}^{s_t-1} A_t^{(u)} = 2^{d_t} \) and \( \frac{\partial A_t^{(1)}}{\partial z} (1) = \sum_{u=0}^{s_t-1} u A_t^{(u)} = \sum_{c \in C_t} w(H(c)) \) where \( w(H(c)) \) denotes the Hamming weight of the codeword \( c \). Therefore, \[ \text{(23)} \) reduces to
\[ \frac{(\beta \rho)}{\sum_{t \in I'} \gamma_t} \left[ \sum_{c \in C_t} w(H(c)) \right] = \frac{1}{2}. \]
Note that the quantity in square brackets is equal to the average weight of a codeword in the local code \( C_t \). Since we assume that the dual code of \( C_t \) has minimum distance greater than one, it follows from the MacWilliams identities (see, e.g., [25]) that the average codeword weight is equal to \( s_t/2 \). Using this fact, \[ \text{(24)} \) reduces to
\[ \frac{(\beta \rho)}{\sum_{t \in I'} \gamma_t s_t} = \frac{1}{2}, \]
which may be seen using the definition \[ \text{(1)} \) of efficient point and the fact that \( \sum_{t \in I'} \rho_t = 1 \). Similarly, making the substitutions \[ \text{(22)} \) in \[ \text{(12)} \) yields
\[ \sum_{t \in I'} \delta_t \left[ \sum_{c \in C_t} w(H(c)) \right] = \frac{1}{2 f \lambda}. \]
Since we assume that all VNs have dual codes with minimum distance greater than one, this reduces to

\[ \sum_{t \in I_c} \frac{\delta_t q_t}{2} = \frac{1}{2} \frac{1}{\lambda}, \]

which may be seen using the definition (1) of \( \delta_t \) and the fact that \( \sum_{t \in I_c} \lambda_t = 1 \). Making the substitutions (22) in (1) yields

\[ \sum_{t \in I_c} \delta_t \left[ \sum_{u=0}^{k_t} \sum_{v=0}^{q_t} u B^{(t)}_{u,v} \right] = \frac{K_s}{2}. \]

For \( t \in I_c \), the quantity in square brackets is equal to the average weight of an information word for the local code \( C_t \); this is always equal to \( k_t/2 \). Thus we obtain

\[ \sum_{t \in I_c} \frac{\delta_t k_t}{2} = \frac{K_s}{2}, \]

which follows from the definition (1) of \( \delta_t \) and the definition (3) of \( K_s \). Finally, making the substitutions (22) in (9) yields

\[ G(K_s/2) = \sum_{t \in I_c} \delta_t \log 2^{k_t} + \left( \int \frac{\rho}{\lambda} \right) \sum_{t \in I_c} \frac{\gamma_t \log 2^{h_t} - \log 2}{\lambda} \]

\[ = \log 2 \left[ \sum_{t \in I_c} \frac{\lambda t k_t}{q_t} + \sum_{t \in I_c} \frac{\rho t h_t}{s_t} - 1 \right] \]

\[ = \log 2 \left[ \frac{\sum_{t \in I_c} \frac{\lambda t k_t}{q_t}}{\lambda} \right] \left[ 1 - \frac{1 - \sum_{t \in I_c} \frac{\rho t h_t}{s_t}}{\sum_{t \in I_c} \frac{\lambda t k_t}{q_t}} \right] \]

\[ = K_s R \log 2, \]

where we have used the expression (3) for the (design) rate \( R \) of the D-GLDPC code. □

A. Evaluation of the Spectral Shape

In the following, we show that a compact expression for the spectral shape follows in this case as a natural corollary to Theorem 5.1. First we introduce the following definition (recall that for \( t \in I_c, \bar{u}_t \) is the maximal weight of a codeword in the local code for CN type \( t \in I_c \)).

**Definition 4.1:** Let

\[ M := \left( \int \rho \right) \sum_{t \in I_c} \gamma_t \bar{u}_t \leq 1 \]  \hspace{1cm} (25)

and define the function \( f : \mathbb{R}^+ \rightarrow [0, M] \) as

\[ f(z) = \left( \int \rho \right) \sum_{t \in I_c} z^{\gamma_t} \frac{d A^{(t)}(z)}{dz}. \]  \hspace{1cm} (26)

Note that we have \( M = 1 \) if and only if \( \bar{u}_t = s_t \) for all \( t \in I_c \).

**Lemma 4.1:** The function \( f \) fulfills the following properties:

1) \( f(0) = f'(0) = 0; \)
2) \( f \) is monotonically increasing for all \( z > 0; \)
3) \( \lim_{z \rightarrow +\infty} f(z) = M. \)

Note that, due to Lemma 4.1, the inverse of \( f \), denoted by \( f^{-1} : [0, M] \rightarrow \mathbb{R}^+ \), is well-defined. We are now in a position to state the main result of this section.

**Theorem 4.2:** Consider a GLDPC code ensemble with a regular VN set, composed of repetition codes all of length \( q \), and a hybrid CN set, composed of a mixture of \( n_c \) different linear block code types. Then, the weight spectral shape of the ensemble is given by

\[ G(\alpha) = (1 - q) h(\alpha) - q \alpha \log f^{-1}(\alpha) \]

\[ + q \left( \int \rho \right) \sum_{t \in I_c} \gamma_t \log A^{(t)}(f^{-1}(\alpha)) \]  \hspace{1cm} (27)

**Proof:** The ensemble constitutes a special case of a D-GLDPC code ensemble where all VNs are repetition codes of the same length, i.e., a check-hybrid GLDPC code ensemble with regular CNs. The proofs of all lemmas in this section are deferred to Section VI.

**IV. SPECTRAL SHAPE OF CHECK-HYBRID GLDPC CODES**

In this section we consider the special case of a D-GLDPC code ensemble where all VNs are repetition codes of the same length, i.e., a check-hybrid GLDPC code ensemble with regular CNs. The proofs of all lemmas in this section are deferred to Section VI.

Although Theorem 4.4 proves only that a stationary point of the weight spectral shape \( H(\omega) \) exists at \( \omega = 1/2 \), we conjecture that this point represents a global maximum of the weight spectral shape for any D-GLDPC code ensemble. Indeed, this is empirically observed for all ensembles we have investigated, even though the number of stationary points of the weight spectral shape is found to vary from 1 to 3 (c.f. Figures 3 4 and 5 later in this paper). Previous work for the special case of LDPC codes also assumed this point to be a global maximum (c.f. the statement of Lemma 7 in [26]).

Note that Theorems 4.4 only holds in general for the weight spectral shape, and not for the stopping set size spectral shape. It is interesting to note that this phenomenon, where the maximum weight spectral shape value of \( R \log 2 \) occurs at half the block length, appears to be quite a general one, occurring widely across many ensembles: for example, all of the spectral shape plots for protograph-based LDPC codes contained in [20] have this property (these were obtained by non-analytical optimization methods). Also, Gallager’s ensemble described in [13], Section 2.1], where the parity-check matrix contains statistically independent equiprobable binary entries, shares the same property (note that this is not a low-density code ensemble).
where the values of $x_0$, $y_0$, $z_0$, $\beta$ in (29) are found by solving the $4 \times 4$ polynomial system

$$
(f \rho) \sum_{t \in I_c} \gamma t \frac{d A(t) (z_0)}{dz} A(t) (z_0) = \frac{\beta}{q},
$$

and

$$
\frac{x_0 y_0}{1 + x_0 y_0} = \alpha,
$$

$$
\frac{x_0 y_0}{1 + x_0 y_0} = \frac{\beta}{q},
$$

and

$$
\frac{z_0 y_0}{1 + z_0 y_0} = \frac{\beta}{q}.
$$

Note that we are certain of the existence of a unique real solution to the polynomial system such that $x_0 > 0$, $y_0 > 0$, $z_0 > 0$, $\beta > 0$, due to Hayman’s formula. We solve this system of equations sequentially for the variables $\beta$, $z_0$, $y_0$ and $x_0$ (respectively). First, combining (31) and (32) yields

$$
\beta = q \alpha.
$$

Substituting (34) into (30) yields $f(z_0) = \alpha$ which may be written as

$$
z_0 = f^{-1}(\alpha).
$$

Using (34) and (35) in (33) yields

$$
y_0 = \frac{\alpha}{(1 - \alpha) f^{-1}(\alpha)}.
$$

Finally, substituting (34) and (36) into (32) yields

$$
x_0 = \frac{1}{\alpha} (1 - q) \left( f^{-1}(\alpha) \right)^q.
$$

Substituting (34), (35), (36) and (37) into (29), and simplifying, leads to (27).

The expression (27) holds regardless of whether the ensemble has good or bad growth rate behavior. Note that, according to (27), the growth rate $G(\alpha)$ is well-defined only for $\alpha \in [0, M]$. This is as expected due to the following reasoning. A codeword of weight $\alpha n$ naturally induces a distribution of bits on the Tanner graph edges, $\alpha n q$ of which are equal to 1. Also note that the maximum number of ones in this distribution occurs when a maximum weight local codeword is activated for each of the $\gamma t m$ CNs of type $t \in I_c$, and is thus given by $n \sum_{t \in I_c} \gamma t \bar{u}_t$. Hence, we have $\alpha n q \leq n \sum_{t \in I_c} \gamma t \bar{u}_t$, i.e., $\alpha \leq M$.

By considering Theorem 4.2 in the special case of Tanner codes, we obtain the following corollary.

**Corollary 4.3:** Consider a Tanner code ensemble where all variable component codes are length-$q$ repetition codes and where all check component codes are length-$s$ codes with weight enumerating function $A(z) = 1 + \sum_{u=r}^{\infty} A u z^u$. The weight spectral shape of this ensemble is given by

$$
G(\alpha) = (1 - q) h(\alpha) - q \alpha \log(f^{-1}(\alpha)) + \frac{q}{s} \log A(f^{-1}(\alpha)),
$$

where the function $f$ is given by (special case of (26))

$$
f(z) = \frac{z A'(z)}{s A(z)},
$$

and $f^{-1} : [0, M] \rightarrow \mathbb{R}^+$ is well-defined, where $M = \frac{n}{r}$ and $\bar{u}$ denotes the largest $u \in \{r, r + 1, \ldots, s\}$ such that $A_u > 0$.

Note that, in the special case where all CNs are SPC codes, (38) becomes equal to the spectral shape expression for regular LDPC codes developed in [11] Theorem 2] for the case of stopping sets. Also note that, in some cases, (38) can be expressed analytically as $f^{-1}(\alpha)$ admits an analytical form. As shown in Appendix C this is the case, for instance, of (3, 6) and (4, 8) regular LDPC code ensembles. This shows that some of Gallager’s $B_j(k, \lambda)$ functions [11] can be expressed in closed form.

**B. Symmetry of the Weight Spectral Shape**

As in the previous subsection, consider a GLDPC code ensemble with a regular VN set and a hybrid CN set. Next, we show how a symmetry in the overall weight spectral shape of the ensemble is induced by local symmetry properties in the WEFs of the CNs.

**Definition 4.2:** The WEF of CN type $t \in I_c$ is said to be symmetric if and only if $A_{\bar{u}_t}^{(t)} = A_{\bar{u}_t}^{(t)}$ for all $u \in \{0, 1, \ldots, \bar{u}_t\}$.

**Lemma 4.4:** The WEF of CN type $t \in I_c$ is symmetric if and only if the all-1 codeword belongs to the code.

**Lemma 4.5:** The WEF of CN type $t \in I_c$ fulfills

$$
A^{(t)}(z) = z^{\bar{u}_t} A^{(t)}(z^{-1}),
$$

for all $z \in \mathbb{R}^+$ if and only if it is symmetric (equivalently, if and only if $\bar{u}_t = s_t$).

**Lemma 4.6:** If $A^{(t)}(z)$ is symmetric for every $t \in I_c$ (i.e., if $M = 1$), then the inverse function $f^{-1}$ fulfills

$$
f^{-1}(M - \alpha) = \frac{1}{f^{-1}(\alpha)}
$$

for all $\alpha \in (0, M)$.

**Theorem 4.7:** Consider a GLDPC code ensemble with a regular VN set, composed of repetition codes all of length $q$, and a hybrid CN set, composed of a mixture of $n_c$ different linear block code types. If $A^{(t)}(z)$ is symmetric for each $t \in I_c$ (equivalently, if $M = 1$), then the spectral shape of the ensemble fulfills

$$
G(M - \alpha) = G(\alpha)
$$

for all $\alpha \in (0, M)$.
Proof: Assume that $A^{(t)}(z)$ is symmetric for each $t \in I_c$ (i.e., $M = 1$). From (27) we have:

$$G(M - \alpha) = (1 - q) h(M - \alpha) - q(M - \alpha) \log f^{-1}(M - \alpha) + q(f) \sum_{t \in I_c} \gamma_t \log A^{(t)}(f^{-1}(M - \alpha))$$

$\equiv (1 - q) h(M - \alpha) - q(M - \alpha) \frac{1}{f^{-1}(\alpha)}$

$+ q(f) \sum_{t \in I_c} \gamma_t \log A^{(t)}(f^{-1}(\alpha))$

$\equiv G(\alpha)$  

where (a) follows from Lemma 4.6 (b) from Lemma 4.5 and (c) from $M = 1$.

We remark that the converse of this result, i.e., that if $G(\alpha) = G(M - \alpha)$ for all $\alpha \in (0, M)$, then $A^{(t)}(z)$ is symmetric for every $t \in I_c$ (and therefore $M = 1$), appears to hold for almost all code ensembles; however this converse appears to be difficult to prove in the general case.

V. PROOF OF THEOREM 3.1

In this section we prove Theorem 3.1. The proof uses the concepts of assignment and split assignment, defined next. These concepts were introduced in [12] and [23], respectively.

Definition 5.1: An assignment is a subset of the edges of the Tanner graph. An assignment is said to have weight $k$ if it has $k$ elements. An assignment is said to be check-valid if the following condition holds: supposing that each edge of the assignment carries a 1 and each of the other edges carries a 0, each CN recognizes a valid local codeword.

Definition 5.2: A split assignment is an assignment, together with a subset of the D-GLDPC code bits (called a codeword assignment). A split assignment is said to have split weight $(u, v)$ if its assignment has weight $v$ and its codeword assignment has $u$ elements. A split assignment is said to be check-valid if its assignment is check-valid. A split assignment is said to be variable-valid if the following condition holds: supposing that each edge of its assignment carries a 1 and each of the other edges carries a 0, and supposing that each D-GLDPC code bit in the codeword assignment is set to 1 and each of the other code bits is set to 0, each VN recognizes a local input word and the corresponding valid local codeword.

For ease of presentation, the proof is broken into two parts.

A. Number of Check-Valid Assignments of Weight $\delta m$

First we derive an expression, valid asymptotically, for the number of check-valid assignments of weight $\delta m$. For each $t \in I_c$, let $\epsilon_t m$ denote the portion of the total weight $\delta m$ apportioned to CNs of type $t$. Then $\epsilon_t \geq 0$ for each $t \in I_c$, and $\sum_{t \in I_c} \epsilon_t = \delta$. Also denote $\epsilon = (\epsilon_1 \epsilon_2 \cdots \epsilon_n)$.

Consider the set of $\gamma_t m$ CNs of a particular type $t \in I_c$, where $\gamma_t$ is given by (1). Using generating functions, the number of check-valid assignments of weight $\epsilon_t m$ is given by

$$N_{c,t}(\gamma_t m)(\epsilon_t m) = \text{Coeff} \left[ (A^{(t)}(x))^{\gamma_t m}, x^{\epsilon_t m} \right]$$

where $\text{Coeff} \left[ p(x), x^c \right]$ denotes the coefficient of $x^c$ in the polynomial $p(x)$. We next apply Lemma A.1, substituting $A(x) = A^{(t)}(x)$, $\ell = \gamma_t m$ and $\xi = \epsilon_t / \gamma_t$; we obtain that as $m \to \infty$

$$N_{c,t}(\gamma_t m)(\epsilon_t m) = \text{Coeff} \left[ (A^{(t)}(x))^{\gamma_t m}, x^{\epsilon_t m} \right]$$

$$\equiv \exp \left\{ m \left( \gamma_t \log A^{(t)}(z_{0,t}) - \epsilon_t \log z_{0,t} \right) \right\}$$

where, for each $t \in I_c$, $z_{0,t}$ is the unique positive real solution to

$$\gamma_t \frac{\partial A^{(t)}(z_{0,t})}{\partial z_{0,t}} \cdot z_{0,t} = \epsilon_t .$$

The number of check-valid assignments of weight $\delta m$ satisfying the constraint $\epsilon$ is obtained by multiplying the numbers of check-valid assignments of weight $\epsilon_t m$ over $\gamma_t m$ CNs of type $t$, for each $t \in I_c$.

$$N_{c}(\delta m) = \prod_{t \in I_c} N_{c,t}(\gamma_t m)(\epsilon_t m).$$

The number $N_c(\delta m)$ of check-valid assignments of weight $\delta m$ is then equal to the sum of $N_c(\delta m)$ over all admissible vectors $\epsilon$; therefore by (44), as $m \to \infty$

$$N_c(\delta m) \equiv \sum_{\epsilon : \sum_{t \in I_c} \epsilon_t m = \delta} \exp \left\{ mW(\epsilon) \right\}$$

where

$$W(\epsilon) = \sum_{t \in I_c} \left( \gamma_t \log A^{(t)}(z_{0,t}) - \epsilon_t \log z_{0,t} \right).$$

As $m \to \infty$, the asymptotic expression is dominated by the distribution $\epsilon$ which maximizes the argument of the exponential function.

Therefore as $m \to \infty$

$$N_c(\delta m) \equiv \exp \left\{ mX \right\}$$

and the maximization is subject to the constraint

$$V(\epsilon) = \sum_{t \in I_c} \epsilon_t = \delta$$

together with $\epsilon_t \geq 0$ for each $t \in I_c$, and for every $t \in I_c$, $z_{0,t}$ is the unique positive real solution to (45). Note that for each $t \in I_c$, (45) provides an implicit definition of $z_{0,t}$ as a function of $\epsilon_t$.

We solve this optimization problem using Lagrange multipliers; at the maximum, we must have

$$\frac{\partial W(\epsilon)}{\partial \epsilon_t} = \lambda \frac{\partial V(\epsilon)}{\partial \epsilon_t}$$

Observe that as $m \to \infty$, $\sum_{t} \exp(mZ_t) \equiv \exp(m \max_t(Z_t))$
for all $t \in I_v$, where $\lambda$ is the Lagrange multiplier. This yields
\[
\frac{\partial z_{0,t}}{\partial t} \left[ \gamma_t \frac{dA^t(z_{0,t})}{A^t(z_{0,t})} - \epsilon_t \right] - \log z_{0,t} = \lambda.
\] (53)
The term in square brackets is equal to zero due to (45); therefore this simplifies to $\log z_{0,t} = -\lambda$ for all $t \in I_v$. We conclude that all of the $\{z_{0,t}\}$ are equal, and we may write
\[
z_{0,t} = z_0 \quad \forall t \in I_v.
\] (54)
Making this substitution in (49) and using (51) we obtain
\[
N_c(\delta m) = \exp \left\{ m \left( \sum_{t \in I_v} \gamma_t \log A^t(z_0) - \delta \log z_0 \right) \right\}.
\] (55)
Summing (45) over $t \in I_v$ and using (51) and (54) implies that the value of $z_0$ in (55) is the unique positive real solution to (10) (here we have defined $\beta$ through the relationship $\beta n = \delta m$, and we have also used the fact that $n \int \rho = m \int \lambda$).

B. Polynomial-System Solution for the Growth Rate
Consider the set of $\delta n$ VN's of a particular type $t \in I_v$, where $\delta_t$ is given by (11). Using generating functions, the number of variable-valid split assignments (over these VNs) of split weight $(\alpha_t n, \beta_t n)$ is given by
\[
N_{v,t}^{(\delta n)}(\alpha_t n, \beta_t n) = \text{Coeff}[\left( B^t(x, y) \right)^{\delta_t n}, x^{\alpha_t n} y^{\beta_t n}]
\] where Coeff $[p(x, y), x^{\gamma} y^{\delta}]$ denotes the coefficient of $x^{\gamma} y^{\delta}$ in the bivariate polynomial $p(x, y)$. Next we apply Lemma A.2, substituting $B(x, y) = B^t(x, y)$, $\ell = \delta_t n$, $\xi = \alpha_t / \delta_t$ and $\theta = \beta_t / \delta_t$; we obtain that as $n \to \infty$
\[
N_{v,t}^{(\delta n)}(\alpha_t n, \beta_t n) = \text{Coeff}[\left( B^t(x, y) \right)^{\delta_t n}, x^{\alpha_t n} y^{\beta_t n}]
\] \[\approx \exp\left\{ nX_t^{(\delta_t)}(\alpha_t, \beta_t) \right\}
\] (56)
where
\[
X_t^{(\delta_t)}(\alpha_t, \beta_t) = \delta_t \log B^t(x_{0,t}, y_{0,t}) - \alpha_t \log x_{0,t} - \beta_t \log y_{0,t}
\] (57)
and where $x_{0,t}$ and $y_{0,t}$ are the unique positive real solutions to the pair of simultaneous equations
\[
\delta_t \frac{\partial B^t(x_{0,t}, y_{0,t})}{\partial x} \cdot x_{0,t} = \alpha_t
\] (58)
and
\[
\delta_t \frac{\partial B^t(x_{0,t}, y_{0,t})}{\partial y} \cdot y_{0,t} = \beta_t.
\] (59)

Next, note that the expected number of D-GLDPC codewords of weight $\alpha n$ in the ensemble $\mathcal{M}_n$ is equal to the sum over $\beta$ of the expected numbers of split assignments of split weight $(\alpha_t n, \beta_t n)$ which are both check-valid and variable-valid, denoted $N_{v,c}^{(\alpha_t n, \beta_t n)}$:
\[
\mathbb{E}_{\mathcal{M}_n}[A_{\alpha n}] = \sum_{\beta} \mathbb{E}_{\mathcal{M}_n}[N_{v,c}^{(\alpha_t n, \beta_t n)}].
\]
This may then be expressed as
\[
\mathbb{E}_{\mathcal{M}_n}[A_{\alpha n}] = \sum_{\alpha_t \geq 0, \beta_t \geq 0, t \in I_v} \sum_{\alpha_t = \alpha} P_{\text{valid}}(\beta n) \times \prod_{t \in I_v} N_{c,t}^{(\delta n)}(\alpha_t n, \beta_t n)
\] (60)
where
\[
\beta := \sum_{t \in I_v} \beta_t.
\] (61)
Here $P_{\text{valid}}(\beta n)$ denotes the probability that a randomly chosen assignment of weight $\beta n$ is check-valid, and is given by
\[
P_{\text{valid}}(\beta n) = N_c(\beta n) / \left( E_{\beta n} \right).
\]
Applying (25) eqn. (25), we find that as $n \to \infty$
\[
\left( E_{\beta n} \right) = \left( \frac{n \int \lambda}{\beta n} \right) \approx \exp\left\{ n \frac{1}{\lambda} h(\beta / \lambda) \right\}
\]
Combining this result with (65), we obtain that as $n \to \infty$
\[
\mathbb{E}_{\mathcal{M}_n}[A_{\alpha n}] \approx \sum_{\alpha_t \geq 0, \beta_t \geq 0, t \in I_v} \sum_{\alpha_t = \alpha} \exp\left\{ n \left( \sum_{t \in I_v} X_t^{(\delta_t)}(\alpha_t, \beta_t) + Y(\beta) \right) \right\}
\] (62)
Note that the sum in (62) is dominated asymptotically by the term which maximizes the argument of the exponential function. Thus, denoting the two vectors of independent variables by $\alpha = (\alpha_t)_{t \in I_v}$ and $\beta = (\beta_t)_{t \in I_v}$, we have
\[
G(\alpha) = \max_{\alpha, \beta} S(\alpha, \beta)
\] (63)
where
\[
S(\alpha, \beta) = \sum_{t \in I_v} X_t^{(\delta_t)}(\alpha_t, \beta_t) + Y(\beta)
\] (64)
where $\beta$ is given by (61), and the maximization is subject to the constraint
\[
R(\alpha, \beta) = \sum_{t \in I_v} \alpha_t = \alpha
\] (65)
together with $\alpha_t \geq 0$ and appropriate inequality constraints on $\beta_t$ for each $t \in I_v$.

Note that (10) provides an implicit definition of $z_0$ as a function of $\beta$. Similarly, for any $t \in I_v$, (58) and (59) provide implicit definitions of $x_{0,t}$ and $y_{0,t}$ as functions of the two variables $\alpha_t$ and $\beta_t$.

We solve the constrained optimization problem using Lagrange multipliers; at the maximum, we must have
\[
\frac{\partial S(\alpha, \beta)}{\partial \alpha_t} = \mu \frac{\partial R(\alpha, \beta)}{\partial \alpha_t}
\]
for all \( t \in I_v \), where \( \mu \) is the Lagrange multiplier. This yields
\[
\frac{\partial x_{0,t}}{\partial \alpha_t} \left[ \frac{\partial \mu(t)}{\partial x_0} (x_{0,t}, y_{0,t}) \right] - \frac{\alpha_t}{x_{0,t}} - \log x_{0,t} \\
+ \frac{\partial y_{0,t}}{\partial \alpha_t} \left[ \frac{\partial \mu(t)}{\partial y_0} (x_{0,t}, y_{0,t}) \right] - \frac{\beta t}{y_{0,t}} = \mu.
\]

The terms in square brackets are zero due to (58) and (59) respectively; therefore this simplifies to \( \log x_{0,t} = -\mu \) for all \( t \in I_v \). We conclude that all of the \( \{x_{0,t}\} \) are equal, and we may write
\[
x_{0,t} = x_0 \quad \forall t \in I_v.
\]

At the maximum, we must also have
\[
\frac{\partial S(\alpha, \beta)}{\partial \beta_t} = \mu \frac{\partial R(\alpha, \beta)}{\partial \beta_t}
\]
for all \( t \in I_v \). This yields
\[
\frac{\partial x_{0,t}}{\partial \alpha_t} \left[ \frac{\partial \mu(t)}{\partial x_0} (x_{0,t}, y_{0,t}) \right] - \frac{\alpha_t}{x_{0,t}} - \log y_{0,t} - \log z_0 \\
+ \frac{\partial y_{0,t}}{\partial \alpha_t} \left[ \frac{\partial \mu(t)}{\partial y_0} (x_{0,t}, y_{0,t}) \right] - \frac{\beta t}{y_{0,t}} - \log (1 - \beta \int \lambda) \\
+ \frac{\partial x_{0,t}}{\partial \beta_t} \left[ \frac{\partial x_{0,t}}{\partial \beta_t} \right] - \frac{\partial y_{0,t}}{\partial \beta_t} \left[ \frac{\partial x_{0,t}}{\partial \beta_t} \right] - \frac{\alpha_t}{x_{0,t}} - \log y_{0,t} - \log z_0 \\
+ \frac{\partial y_{0,t}}{\partial \beta_t} \left[ \frac{\partial y_{0,t}}{\partial \beta_t} \right] - \frac{\beta t}{y_{0,t}} - \log (1 - \beta \int \lambda)
\]
\[
+ \frac{\partial x_{0,t}}{\partial \beta_t} \left[ \frac{\partial x_{0,t}}{\partial \beta_t} \right] - \frac{\partial y_{0,t}}{\partial \beta_t} \left[ \frac{\partial x_{0,t}}{\partial \beta_t} \right] - \frac{\alpha_t}{x_{0,t}} - \log y_{0,t} - \log z_0 \\
+ \frac{\partial y_{0,t}}{\partial \beta_t} \left[ \frac{\partial y_{0,t}}{\partial \beta_t} \right] - \frac{\beta t}{y_{0,t}} - \log (1 - \beta \int \lambda)
\]
\[
= 0. \tag{67}
\]

The terms in square brackets are zero due to (58), (59) and (10) respectively; therefore this simplifies to
\[
z_0 y_{0,t} \left( 1 - \frac{\beta \int \lambda}{\beta \int \lambda} \right) = 1 \quad \forall t \in I_v. \tag{68}
\]

We conclude that all of the \( \{y_{0,t}\} \) are equal, and we may write
\[
y_{0,t} = y_0 \quad \forall t \in I_v. \tag{69}
\]

Rearranging (68) we obtain (13). Also, summing (58) over \( t \in I_v \) and using (65) and (66) yields (11). Similarly, summing (59) over \( t \in I_v \) and using (61) and (69) yields (12). Substituting back into (64) and using (66), (69), (55) and (61) yields
\[
G(\alpha) = \sum_{t \in I_v} \delta_t \log B(t)(x_0, y_0) - \alpha \log x_0 - \log y_0 \\
+ \left( \frac{\rho}{\int \lambda} \right) \sum_{s \in I_v} \gamma_s \log A(s)(z_0) - \beta \log z_0 - \frac{h(\beta \int \lambda)}{\int \lambda} \tag{70}
\]

where \( x_0, y_0, z_0 \) and \( \beta \) are the unique positive real solutions to the \( 4 \times 4 \) system of equations (10), (11), (12) and (13).

Finally, (13) leads to the observation that
\[
-\beta \log z_0 - \beta \log y_0 = \frac{h(\beta \int \lambda)}{\int \lambda} = \frac{\log (1 - \beta \int \lambda)}{\int \lambda}
\]
which, when substituted in (70), leads to (9).
initial phase the $2^{h_t}$ balls are placed into the bins according to the WEF $A^{(t,1)}(u)$ corresponding to $G^{(1)}_t$. Note that no balls are placed into bins with label in $\{u_t + 1, \ldots, s_t\}$ and that, due to the symmetry of $A^{(t,1)}(u)$, the number of balls in bins $u$ and $u_t - u$ is the same for all $u \in \{0,\ldots,u_t\}$. Then, in the adjustment phase the balls are moved in order to match the WEF $A^{(t)}(u)$, according to $G^{(2)}_t$. The key observation here is that, in the adjustment phase, every ball must either stay in its current bin, or else move to the right. This is because, for any information word $v$, of length $h_t$, the Hamming weight of $vG^{(t)}_1$ cannot be larger than the Hamming weight of $vG^{(t)}_1$.

Denoting by $u_{\min}$ the minimum weight such that $A^{(t)}_{u} \neq A^{(t)}_{u}$, we must have $A^{(t,1)}_{u_{\min}} \geq A^{(t)}_{u_{\min}}$ (since no ball can move into bin $u_{\min}$, while at least one ball has moved out). Note that the number of balls in all bins corresponding to weights smaller than $u_{\min}$ remains unchanged. Since $A^{(t,1)}(z)$ is symmetric, and since so is $A^{(t)}(z)$ by hypothesis, we must also have $A^{(t,1)}_{u_{\min}} > A^{(t)}_{u_{\min}}$. It follows that the total number of balls in bins with labels in $\{u_t - u_{\min}, \ldots, u_t\}$ has decreased during the adjustment phase. But this is a contradiction, and so the result is established.

**Proof of Lemma 4.3** Assume $A^{(t)}(z)$ is symmetric (and therefore $u_t = s_t$). We have

$$z^{s_t}A^{(t)}(z^{-1}) = \sum_{u=0}^{s_t} A^{(t)}(z^{s_t-u}) = \sum_{v=0}^{s_t} A^{(t)}(z^{v}) = A^{(t)}(z)$$

where the final equality is due to $A^{(t)}_{z^{u}} = A^{(t)}_{z}$ for all $z \in \{0,1,\ldots,s_t\}$. Conversely, assume that (40) is satisfied for all $z \in \mathbb{R}^+$. It can be easily recast as

$$\sum_{u=0}^{s_t} A^{(t)}(z^{u}) = \sum_{u=0}^{u_t} A^{(t)}_{u_{\min}}(z^{u}).$$

In order for this equality to be satisfied for all $z \in \mathbb{R}^+$, we must have $A^{(t)}_{u_{\min}} = A^{(t)}_{u}$ for all $u \in \{0,1,\ldots,s_t\}$. Hence, $A^{(t)}(z)$ must be symmetric.

**Proof of Lemma 4.4** We prove that the function $f(z)$ fulfills

$$f(z) = M - f\left(z^{-1}\right) \tag{71}$$

$\forall z \in \mathbb{R}^+$ if $A^{(t)}(z)$ is symmetric for every $t \in I_c$. The result is then obtained by applying the inverse function $f^{-1}$ to both sides of (71) and by letting $f(z^{-1}) = \alpha$ for all $z \in \mathbb{R}^+ \setminus \{0\}$.

Assuming that the WEF of CN type $t \in I_c$ is symmetric, we have, differentiating (40),

$$dA^{(t)}(z)/dz = -z^t \cdot A^{(t)}(z^{-1}) dA^{(t)}(z^{-1})/dz^{-1} + z^t \cdot A^{(t)}(z^{-1}).$$

Multiplying by $z$ and using (40) yields

$$\frac{dA^{(t)}(z)}{dz} = -z^t \cdot A^{(t)}(z^{-1}) dA^{(t)}(z^{-1})/dz^{-1} + \bar{u}_t A^{(t)}(z). \tag{72}$$

**Table I**

| Ensemble 1 | Variable nodes | Check nodes |
|------------|----------------|-------------|
| 1:repetition−2 | $\lambda_1 = 0.055646$ | $\rho_1 = 0.965221$ |
| 2:SPC−7 (C) | $\lambda_2 = 0.944354$ | $\rho_2 = 0.034779$ |

| Ensemble 2 | Variable nodes | Check nodes |
|------------|----------------|-------------|
| 1:repetition−2 | $\lambda_1 = 0.022647$ | $\rho_1 = 0.965221$ |
| 2:SPC−7 (C) | $\lambda_2 = 0.100000$ | $\rho_2 = 0.034779$ |
| 3:SPC−7 (A) | $\lambda_3 = 0.539920$ | $\rho_2 = 0.034779$ |
| 4:SPC−7 (S) | $\lambda_4 = 0.337432$ | $\rho_2 = 0.034779$ |

Then,

$$M - f\left(z^{-1}\right) = M - f\left(\rho \sum_{t \in I_c} \gamma_t \left(\frac{-z^{-1}dA^{(t)}(z^{-1})}{dA^{(t)}(z^{-1})} + A^{(t)}(z^{-1})\right)\right) \tag{73}$$

$\equiv \left(\rho \sum_{t \in I_c} \gamma_t \left(\frac{-z^{-1}dA^{(t)}(z^{-1})}{dA^{(t)}(z^{-1})} + A^{(t)}(z^{-1})\right)\right) \tag{74}$

$\equiv \left(\rho \sum_{t \in I_c} \gamma_t \left(\frac{-z^{-1}dA^{(t)}(z^{-1})}{dA^{(t)}(z^{-1})} + A^{(t)}(z^{-1})\right)\right) \tag{75}$

where we have used (25) and (40) in (a), and (72) in (b).

**VII. Examples**

In this section, the spectral shapes of some example GLDPC and D-GLDPC ensembles are evaluated using the polynomial solution of Theorem 3.1. We will consider both BD and MAP CN decoding. Considering the former case, note that under bounded distance decoding, the non-empty stopping sets are precisely those sets which have $r_t$ or more erased code bits; therefore the local SSEF (BD-SSEF) is given by

$$\Psi^{(t)}(z) = 1 + \sum_{u=r_t}^{s_t} \left(\frac{s_t}{u}\right) z^u. \tag{73}$$

In the latter case, the local SSEF (MAP-SSEF) is given by

$$\Phi^{(t)}(z) = 1 + \sum_{u=r_t}^{s_t} \phi_u^\alpha z^u \tag{74}$$

where $\phi_u^\alpha \geq 0$ is the number of local stopping sets (under MAP decoding) of size $u$.

Furthermore, numerical examples are presented on the approximation of the parameter $\alpha$ for regular LDPC code ensembles, based on (74).

**Example 7.1:** [D-GLDPC ensembles with Hamming CNs and SPC VNs] In this first example, we design two ensembles with design rate $R = 1/2$ using Hamming $(7, 4)$ codes as generalized CNs and SPC codes as generalized VNs. Three representations of SPC VNs are considered, namely,
the cyclic (C), the systematic (S) and the antisystematic (A) representations.

Ensemble 1 is characterized by two CN types and two VN types. Specifically, we have $I_c = \{1, 2\}$, where $1 \in I_c$ denotes a $(7, 4)$ Hamming CN type and $2 \in I_c$ denotes a length-7 single parity check (SPC) CN type, and $I_v = \{1, 2\}$, where $1 \in I_v$ denotes a repetition-2 VN type and $2 \in I_v$ denotes a length-7 SPC CN type in cyclic form. Ensemble 2 is characterized by two CN types and four VN types. Specifically, we have $I_c = \{1, 2\}$, where $1 \in I_c$ denotes a $(7, 4)$ Hamming CN type and $2 \in I_c$ denotes a SPC-7 CN type, and $I_v = \{1, 2, 3, 4\}$, where $1 \in I_v$ denotes a repetition-2 VN type, $2 \in I_v$ denotes a length-7 SPC CN type in cyclic form, $3 \in I_v$ denotes a length-7 SPC CN type in antisystematic form, and $4 \in I_v$ denotes a length-7 SPC CN type in systematic form. The edge-perspective type distributions of the two ensembles are summarized in Table I.

Both Ensemble 1 and Ensemble 2 have been obtained by performing a decoding threshold optimization with differential evolution [27]. This is an evolutionary parallel optimization algorithm to find the global minimum of a real-valued cost function of a vector of continuous parameters, where the cost function may even be defined by a procedure (e.g., density evolution returning the threshold for an LDPC code ensemble over some channel and under some decoding algorithm). It is based on the evolution of a population of $N_p$ vectors, and its main steps are the same as typical evolutionary optimization algorithms (mutation, crossover, selection) [28]. A starting population of $N_p$ vectors is first generated. Then, a competitor (or trial vector) for each population element is generated by combining a subset of randomly chosen vectors from the same population. Finally, each element of the population is compared with its trial vector and the vector yielding the smallest cost function value is selected as the corresponding element of the evolved population [2]. These steps are iterated until a certain stopping criterion is fulfilled or until a maximum number of iterations is reached. Differential evolution was first proposed for the optimization of LDPC code degree profiles in [29].

In our experiments, each element of the population was a pair of polynomials $(\lambda, \rho)$ corresponding to given variable and check component codes, given VN representations, and given design rate $R$, while the cost function was the ensemble threshold over the BEC (returned by numerical procedure), under iterative decoding with MAP erasure decoding at the VNs and CNs.

For Ensemble 1 we have $C \cdot V = 1.19 > 1$, so the ensemble has bad growth rate behavior ($\alpha^* = 0$). Ensemble 2 has been obtained by imposing the further constraints $C \cdot V \leq 0.5$ and $\lambda_3 \geq 0.1$ on differential evolution optimization. Since in this case we have $C \cdot V = 0.5 < 1$, the ensemble has good growth rate behavior ($\alpha^* > 0$). The expected good or bad growth rate behavior of the two ensembles is reflected in the growth rate curves shown in Fig. 2. Using a standard numerical solver, it took only 5.1 s and 6.7 s to evaluate 100 points on the Ensemble 1 curve and on the Ensemble 2 curve, respectively. The relative minimum distance of Ensemble 2 is $d_{\text{min}} = 1$ with one codeword of weight 1.

7The $(k \times (k+1))$ generator matrix of a SPC code in A form is obtained from the generator matrix in S form by complementing each bit in the first $k$ columns. Note that a $(k \times (k+1))$ generator matrix in A form represents a SPC code if and only if the code length $q = k+1$ is odd. For even $k+1$ we obtain a $d_{\text{min}} = 1$ code with one codeword of weight 1.

8Usually, the algorithm’s “greediness” is reduced by selecting the vector yielding the smallest cost with a probability smaller than one, instead of systematically selecting it.
Fig. 4. Weight spectral shape of the check-hybrid GLDPC code ensemble in Example 7.4. Critical exponent codeword weight ratio: $\alpha^* = 0.028179$.

Fig. 5. Weight spectral shape of the Tanner code ensemble in Example 7.4.

rate behavior). The WEF of a Hamming $(7, 4)$ CN is given by $A(z) = 1 + 7z^3 + 7z^4 + z^7$, while its local MAP-SSEF and BD-SSEF are given by $\Phi(z) = 1 + 7z^3 + 10z^4 + 21z^5 + 7z^6 + z^7$ and $\Psi(z) = 1 + 35z^3 + 35z^4 + 21z^5 + 7z^6 + z^7$ respectively. Note that we have $M = \frac{2}{7} = 1$ in all three cases. A plot of $G(\alpha), G_{\Phi}(\alpha)$ and $G_{\Psi}(\alpha)$ obtained by implementation of (38) is depicted in Fig. 5. We observe that $A(z)$ satisfies the conditions of Theorem 4.7. This is reflected by the fact that the weight spectral shape $G(\alpha)$ is symmetric with respect to $\alpha = 1/2$.

Example 7.3: [Check-hybrid ensemble] Consider a rate $R = 1/3$ check-hybrid GLDPC code ensemble where all VNs are repetition codes of length $q = 3$ and whose CN set is composed of a mixture of two linear block code types ($I_c = \{1, 2\}$). CNs of type $1 \in I_c$ are length-7 SPC codes with WEF $A^{(1)}(z) = [(1 + z)^7 + (1 - z)^7]/2$ and $\gamma_1 = 0.722$, while CNs of type $2 \in I_c$ are $(7, 4)$ codes with WEF $A^{(2)}(z) = 1 + 5z^2 + 7z^3 + 3z^6$ and $\gamma_2 = 0.278$. The weight spectral shape of this ensemble, obtained from (27), is depicted in Fig. 4. Note that for this ensemble $M = 6/7$, and also that the weight spectral shape does not exhibit any symmetry property (the CN WEFs are not symmetric).

Example 7.4: [Ensemble with bad growth rate behavior] Consider a rate $R = 1/5$ Tanner code ensemble where all VNs are repetition codes of length $q = 2$ and where all CNs are (5, 3) linear block codes with WEF $A(z) = 1 + 3z^2 + 3z^3 + z^5$. This ensemble is known to have bad growth rate behavior ($\alpha^* = 0$) since we have $A'(0)/C = 6/5 > 1$, where $\lambda(x) = x$ and $C = 2A_2/s$ [30], [31]. A plot of the weight spectrum for this ensemble, obtained from (38) is depicted in Fig. 5. We observe that the plot of $G(\alpha)$ is symmetric, due to the fact that $A(z)$ is symmetric ($M = 1$). As expected, the derivative of $G(\alpha)$ at $\alpha = 0$ is positive and hence $\alpha^* = 0$.

Finally, note that for the ensembles of Figures 4 and 5 the weight spectral shape has in each case a maximum value of $R \log 2$ which occurs at $\alpha = 1/2$, in accordance with Theorem 4.4.

Example 7.5: Fig. 6 illustrates a comparison between the actual value of the parameter $\alpha^*$ for some high-rate regular LDPC code ensembles (solid curves) and the corresponding approximation (21) obtained via growth rate analysis in the small-$\alpha$ case (dashed curves). Each curve corresponds to a value of the design rate $R$, so the CN degree $d_c$ for each point is given by $d_c = d_v/(1 - R)$ where the VN degree $d_v$ is reported in abscissa. From the figure we see that the approximation is quite good for small values of $d_v$, which are usually the ones of interest in practical applications. Moreover, the higher the design rate $R$ the larger the range of $d_v$ over which the approximation is satisfactory. The actual values of $\alpha^*$ and their corresponding approximated values are reported in Table II for several regular LDPC code ensembles with $d_v = 3$. A very good match can be observed. Similar examples may be developed, for instance, for Tanner codes through (20).

VIII. CONCLUSION

A general expression for the weight and stopping set size spectral shapes of irregular D-GLDPC ensembles has been presented. Evaluation of the expression requires solution of a $4 \times 4$ polynomial system, irrespective of the number of VN and CN types in the ensemble. A compact expression was developed for the special case of check-hybrid GLDPC codes, and both a necessary and a sufficient condition for symmetry of the weight spectral shape was developed. Simulation results
TABLE II
COMPARISON BETWEEN THE ACTUAL VALUES OF THE PARAMETER \( \alpha^* \)
FOR SOME REGULAR LDPC ENSEMBLES WITH VN DEGREE \( d_v = 3 \) AND
THE CORRESPONDING APPROXIMATED VALUES BASED ON \( 23 \).

| \( R, d_c \) | \( \alpha^* \), exact | \( \alpha^* \), approx. |
|----------|-----------------|-----------------|
| 1/4, 4   | 0.112159        | 0.100677        |
| 2/5, 5   | 0.045365        | 0.042473        |
| 1/2, 6   | 0.022733        | 0.021746        |
| 4/7, 7   | 0.012993        | 0.012585        |
| 5/8, 8   | 0.008117        | 0.007925        |
| 2/3, 9   | 0.005410        | 0.005309        |
| 7/10, 10 | 0.003785        | 0.003729        |

were presented for two example optimized irregular D-GLDPC code ensembles as well as a number of check-hybrid GLDPC code ensembles.

APPENDIX A
SOME USEFUL LEMMATA

The following results are special cases of [12 Corollary 16].

**Lemma A.1:** Let \( A(x) = 1 + \sum_{u=c}^{d} A_u x^u \), where \( 1 \leq c \leq d \), be a polynomial satisfying \( A_c > 0 \) and \( A_u \geq 0 \) for all \( c < u \leq d \). Then, as \( \ell \to \infty \),

\[
\text{Coeff} \left( A(x) \right)^\ell_x^{\ell} = \exp \left[ \ell \log \left( \frac{A(z)}{z} \right) \right]
\]

where \( z \) is the unique positive real solution to

\[
\frac{A'(z)}{A(z)} \cdot z = \xi.
\]

**Lemma A.2:** Let

\[ B(x, y) = 1 + \sum_{u=1}^{k} \sum_{v=e}^{d} B_{u,v} x^u y^v \]

where \( k \geq 1 \) and \( 1 \leq e \leq d \), be a bivariate polynomial satisfying \( B_{u,v} \geq 0 \) for all \( 1 \leq u \leq k, \ c \leq v \leq d \). Then, as \( \ell \to \infty \),

\[
\text{Coeff} \left[ (B(x, y))^\ell \right] \cdot x^{\ell \xi} y^{\ell \rho} = \exp \left[ \ell \log \left( \frac{B(x_0, y_0)}{x_0^{\xi_0} y_0^{\rho_0}} \right) \right]
\]

where \( x_0 \) and \( y_0 \) are the unique positive real solutions to the pair of simultaneous equations

\[
\frac{\partial B}{\partial x}(x_0, y_0) \cdot x_0 = \xi
\]

and

\[
\frac{\partial B}{\partial y}(x_0, y_0) \cdot y_0 = \theta.
\]

APPENDIX B
SOLUTION FOR SMALL LINEAR-WEIGHT CODEWORDS

In this appendix, we analyze Theorem 3.1 for the case of small \( \alpha \). Specifically, we prove Corollary 5.2, which is a slightly weaker form of [23 Theorem 4.1]. The proof consists of first obtaining expressions for \( x_0, y_0 \) and \( x_0 \) in terms of \( \beta \), and for \( \alpha \) in terms of \( \beta \), and then exploiting these in [9].

First we develop an expression for \( x_0 \) in terms of \( \beta \). Considering \( 11 \), we have that its left-hand side must be \( o(1) \) because so is its right-hand side. The only possibility is that \( x_{0i} y_{0j} = o(1) \) for some admissible choices of \((i, j, t)\), and \( x_{0i} y_{0j} = o(\alpha) \) for all other admissible choices. Since the only difference between the left-hand side of \( 11 \) and the left-hand side of \( 12 \) is represented by the coefficients of the \( x_{0j} y_{0j} \) terms, it follows that

\[
\lim_{\alpha \to 0} \beta = 0.
\]

For the moment, the notation \( o(1) \) will be intended as \( \beta \to 0 \). (We will show later that \( \beta \) is proportional to \( \alpha \) to the first order, so that the notation \( o(\beta) \) is equivalent to the notation \( o(\alpha) \).) Because of the above discussion, the left-hand side of \( 10 \) must also be \( o(1) \), a condition which can be satisfied if and only if \( z_{0i} = o(1) \) for some admissible choices of \((u, t)\) and \( z_0 = o(\beta) \) for all other admissible choices of \((u, t)\). Consider now \( 10 \) written as an equality of polynomials. Due to \( 80 \), its left-hand side is dominated by the terms corresponding to \( u = r \), and the equation can be written in the form

\[
z_{0i} \sum_{t: r_i=r} \frac{\partial r_i A(t)}{s_t} = \beta \frac{\lambda}{C} (1 + o(1)) ,
\]

i.e.

\[
z_0 = \left( \frac{\beta \frac{\lambda}{C}}{1 - \beta \frac{\lambda}{C}} \right)^{1/\rho} (1 + o(1)) .
\]

We next develop an expression for \( y_0 \) in terms of \( \beta \); from \( 13 \) we have

\[
y_0 = \frac{1}{z_0} \cdot \beta \frac{\lambda}{1 - \beta \frac{\lambda}{C}}
\]

which, combined with \( 81 \) and \( 80 \) respectively, yields

\[
y_0 = C^{1/\rho} (\beta \frac{\lambda}{C})^{1/\rho} (1 + o(1)) .
\]

Next, we develop an expression for \( x_0 \) in terms of \( \beta \). To this purpose, since \( x_{0i} y_{0j} = o(1) \) for some admissible choices of \((i, j)\), \( 12 \) (when written as an equality of polynomials) can be expressed as

\[
\sum_{t \in T_0} \frac{\lambda_j}{q_j} \sum_{(i,j) \in S_i} j B_{i,j} x_0 y_0 = \beta (1 + o(1))
\]

where \( S_i^- = \{(i, j) \neq (0, 0): B_{i,j}^{(t)} > 0\} \). Using \( 82 \), this latter equation can be written in the form

\[
\sum_{t \in T_0} \frac{\lambda_j}{q_j} \sum_{(i,j) \in S_i^-} j B_{i,j}^{(t)} C^{\lambda_j} (x_0 (\beta \frac{\lambda}{C})^{r_{i,j}})^i = 1 + o(1) ,
\]

Here (and often throughout the proof) we use the property \((1 + o(1))^k = 1 + o(1)) for rational \( k \).
where $T_{i,j} = (j - i)/i \geq 0$. It follows that, as $\alpha \to 0$, the left-hand side of (83) is dominated by summands corresponding to admissible choices of $(i, j)$ for which $x_0 \beta^{T_{i,j}/\psi}$ tends to a constant. Letting $T = \max T_{i,j}$, we have $T_{i,j} \in S^{-}_{\psi}$ for all $i, j \in S$ since $T_{i,j} \in S^{-}_{\psi}$ for all $i, j \in S$. Then, for all choices of $(i, j)$ such that $T_{i,j} = T$, we can write $x_0 \beta^{T_{i,j}/\psi}$ as $c + o(1)$ for some $T_{i,j} > T$ and constant $c$. Hence, we have

$$
\sum_{t \in I_v} \frac{\lambda_t}{q_t} \sum_{(i,j) \in S^-} j B^{(t)}_{i,j} C^{j/r} \left( x_0 (\beta f \lambda)^{2} \right)^i = 1 + o(1),
$$
i.e.,

$$Q_1 \left( x_0 (\beta^{T_{i,j}}) \right) = 1 + o(1)$$

and therefore,

$$x_0 = Q_1^{-1}(1) e^{T_{i,j}} \beta^{-1}(1 + o(1)).$$

Next, we develop an expression for $\beta$ in terms of $\alpha$. Similarly to (11) (when written as an equality of polynomials) can be expressed as

$$\sum_{t \in I_v} \frac{\lambda_t}{q_t} \sum_{S^-} j B^{(t)}_{i,j} x_0^{j} y_0^{i} = \alpha(1 + o(1))$$

where $o(1)$ is intended as $\alpha \to 0$. Using (82) and dividing each side by $\beta$ we obtain

$$\sum_{t \in I_v} \frac{\lambda_t}{q_t} \sum_{S^-} i B^{(t)}_{i,j} C^{j/r} \left( x_0 (\beta f \lambda)^{2} \right)^i = \frac{\alpha}{\beta}(1 + o(1)).$$

Reasoning in the same way as we did for (83), we can write the previous equation in the form

$$Q_2 \left( x_0 (\beta^{T_{i,j}}) \right) = \frac{\alpha}{\beta}(1 + o(1))$$

which, using (84), becomes

$$Q_2(Q_1^{-1}(1)) = \frac{\alpha}{\beta}(1 + o(1)).$$

This yields

$$\beta = \frac{\alpha}{Q_2(Q_1^{-1}(1))}(1 + o(1)).$$

We conclude from (83) that $\beta$ is proportional to $\alpha$ to the first order, so that $o(\beta)$ and $o(\alpha)$ are equivalent notations.

We next use the derived expressions to analyze (9). Using the Taylor series of $\log(1 + x)$ around $x = 0$, we have

$$\log(1 - \beta f \lambda) = -\beta(1 + o(1)).$$

Consider now the term $\frac{\rho}{f\lambda} \sum_{t \in I_v} \gamma_t \log(A^{(t)}(z_0))$. Because of (81), we have

$$A^{(t)}(z_0) = \begin{cases} 1 + A^{(t)}(\beta f \lambda)^{(1 + o(1))} & \text{if } r_t = r, \\ 1 + o(\beta) & \text{if } r_t > r \end{cases}$$

and, using the Taylor series of $\log(1 + x)$ around $x = 0$,

$$\log A^{(t)}(z_0) = \begin{cases} A^{(t)}(\beta f \lambda)^{(1 + o(1))} & \text{if } r_t = r, \\ o(\beta) & \text{if } r_t > r \end{cases}$$

This yields

$$\frac{\rho}{f\lambda} \sum_{t \in I_v} \gamma_t \log(A^{(t)}(z_0)) = \sum_{t : r_t = r} \frac{\rho}{f\lambda} A^{(t)}(\beta f \lambda)^{(1 + o(1))}$$

and finally, recalling (17),

$$\frac{\rho}{f\lambda} \sum_{t \in I_v} \gamma_t \log(A^{(t)}(z_0)) = \frac{\beta}{r}(1 + o(1)).$$

We then have

$$\frac{\rho}{f\lambda} \sum_{t \in I_v} \gamma_t \log(A^{(t)}(z_0)) + \log(1 - \beta f \lambda) = -\frac{\beta}{\psi}(1 + o(1)).$$

Next, the term $-\alpha \log x_0$ may be developed as follows,

$$-\alpha \log x_0 = -\alpha \log \left( \frac{Q_1^{-1}(1)}{e^{\psi f \lambda}} \right)(1 + o(1))$$

$$= -\alpha \log \left( \frac{Q_1^{-1}(1)}{e^{\psi f \lambda}} \right) Q_2(Q_1^{-1}(1))^{-\psi}(1 + o(1))$$

$$= -\alpha \left[ \log Q_1^{-1}(1) + T \log Q_2(Q_1^{-1}(1)) - T \log \frac{Q_2(Q_1^{-1}(1))}{\psi}(1 + o(1)) \right]$$

where we have used (84) in (a), (85) in (b), and the Taylor series of $\log(1 + x)$ around $x = 0$ in (c). Hence, we conclude that

$$-\alpha \log x_0 = \alpha \left( \log \frac{1}{Q_1^{-1}(1)} + T \frac{T}{\psi} \log Q_2(Q_1^{-1}(1)) \right)$$

$$+ T \frac{T}{\psi} \alpha \log \alpha + T \frac{T}{\psi} Q_2(Q_1^{-1}(1)) \beta + o(\alpha)$$

(89)

where we have again used (85).

Finally, we analyze the term $\sum_{t \in I_v} \delta_t \log B^{(t)}(x_0, y_0)$. Using (82) and (84), we obtain

$$B^{(t)}(x_0, y_0) = 1 + \sum_{(i,j) \in S^-} B^{(t)}_{i,j} x_0^{j} y_0^{i}$$

$$= 1 + \sum_{(i,j) \in S^-} B^{(t)}_{i,j} \left( \frac{Q_1^{-1}(1)}{e^{\psi f \lambda}} \right)^{j} C^{j}(f \lambda)^{-\psi} \beta^{i}(1 + o(1)).$$

From $T_{i,j} = \frac{\lambda_t}{q_t} \geq T$ for all $t \in I_v$, $(i, j) \in S^-$, it follows that $\frac{\lambda_t}{q_t} \geq 1$ for all $t \in I_v$, $(i, j) \in S^-$, with equality if and only if $t \in Y_v$ and $(i, j) \in P_t$. Then we have

$$B^{(t)}(x_0, y_0) = 1 + \sum_{(i,j) \in P_t} B^{(t)}_{i,j} \left( \frac{Q_1^{-1}(1)}{e^{\psi f \lambda}} \right)^{j} C^{j}(f \lambda)^{-\psi} \beta^{i}(1 + o(1))$$

Note: that, using the Taylor series of $Q_1^{-1}(1)$ around $x = 0$, we have

$$Q_1^{-1}(1 + o(1)) = Q_1^{-1}(1) (1 + o(1)).$$

Note: that $Q_2(Q_1^{-1}(1) + o(1)) = Q_2(Q_1^{-1}(1)) + o(1)$ using the Taylor series of $Q_2(Q_1^{-1}(1)) + o(1)$ around $x = 0$. 

10Note that, using the Taylor series of $Q_1^{-1}(1)$ around $x = 0$, we have

$$Q_1^{-1}(1 + o(1)) = Q_1^{-1}(1) + o(1).$$

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15
when \( t \in Y_v \), and \( B^t(x_0, y_0) = 1 + o(\beta) \) otherwise. This implies that

\[
\log B^t(x_0, y_0) = \sum_{(i,j) \in P_t} B_{i,j}^t \left( \frac{Q_{1}^{-1}(1)}{e^{T/\psi}} \right)^i C^t \left( f(\lambda) \right)^j \beta(1+o(1))
\]

(90)

when \( t \in Y_v \), and \( \log B^t(x_0, y_0) = o(\beta) \) otherwise. This yields

\[
\sum_{t \in I_v} \delta_t \log B^t(x_0, y_0) = \beta(1+o(1)) \sum_{t \in Y_v} \delta_t \sum_{(i,j) \in P_t} B_{i,j}^t \left( \frac{Q_{1}^{-1}(1)}{e^{T/\psi}} \right)^i C^t \left( f(\lambda) \right)^j
\]

(91)

\[
= \beta(1+o(1)) \sum_{t \in Y_v} \delta_t \sum_{(i,j) \in P_t} j B_{i,j}^t \left( \frac{Q_{1}^{-1}(1)}{e^{T/\psi}} \right)^i C^t \left( f(\lambda) \right)^j
\]

\[
- \beta T \psi (1+o(1)) \sum_{t \in Y_v} \delta_t \sum_{(i,j) \in P_t} i B_{i,j}^t \left( \frac{Q_{1}^{-1}(1)}{e^{T/\psi}} \right)^i C^t \left( f(\lambda) \right)^j
\]

\[
= \beta(1+o(1)) \sum_{t \in Y_v} \frac{\lambda_t}{\psi} q_t \sum_{(i,j) \in P_t} j B_{i,j}^t C^t \left( \frac{\lambda}{e} \right)^j \left( Q_{1}^{-1}(1) \right)^i
\]

Finally, substituting (88), (89), and (91) into (90), we obtain (14), as desired.

APPENDIX C

CLOSED FORM EXPRESSIONS FOR THE SPECTRAL SHAPE

It is worthwhile to note that in some cases, (88) can be expressed in closed form because \( f^{-1}(\alpha) \) can be expressed analytically. This is the case, for instance, for the \((3, 6)\) regular LDPC ensemble, for which \( f(z) = \alpha \) becomes \( ax^3 + bx^2 + cx + d = 0 \), where \( x = z^\alpha \) and \( \alpha = 1 - 10, 10, 15, 20 \). The cubic equation in \( z \) may be solved by Cardano’s method (see, e.g., [32] p. 17); the discriminant \( \Delta = \mu^2 + 4\rho^3 \) is negative for every \( \alpha \in (0, 1) \), where

\[
\rho = \frac{3ac - b^2}{9a^2} \quad \mu = \frac{9abc - 27a^2d - 2b^3}{54a^3}.
\]

The required solution is then uniquely and analytically identified as that given by [35] where \( q = 3, s = 6 \) and \( f^{-1}(\alpha) = z = \sqrt[3]{-\Delta/\mu} \) where \( x = 2\sqrt{-\rho \cos(\theta/3)} - b/3 \), \( \theta = \tan^{-1}(\sqrt{-\Delta}/\mu) \).

Similarly, the weight spectral shape of a \((4, 8)\) regular LDPC ensemble may be expressed in closed form through the solution of a quartic equation.

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