Raptor Codes for Higher-Order Modulation Using a Multi-Edge Framework

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Abstract—In this paper, we represent Raptor codes as multi-edge type low-density parity-check (MET-LDPC) codes, which gives a general framework to design them for higher-order modulation using MET density evolution. We then propose an efficient Raptor code design method for higher-order modulation, where we design distinct degree distributions for distinct bit levels. We consider a joint decoding scheme based on belief propagation for Raptor codes and also derive an exact expression for the stability condition. In several examples, we demonstrate that the higher-order modulated Raptor codes designed using the multi-edge framework outperform previously reported higher-order modulation codes in literature.

Index Terms—Density evolution, Multi-edge type LDPC codes, Modulation, Raptor codes

I. INTRODUCTION

The design of Raptor codes with binary modulation has been well investigated in literature [1], [2]. In contrast, there has been little progress on universal design methods for Raptor codes with bandwidth efficient higher-order modulation, and a complete analysis is still missing. Existing work applying Raptor codes to higher-order modulation [3], [4] have considered the bit-interleaved coded modulation scheme (BICM) [5], [6], which uses a single binary code to protect all bits in the signal constellation. However, it has been shown [7] that the BICM scheme, which averages over the different bit-channel reliabilities, may not show the optimal performance if the degree distribution is irregular, which is the case for Raptor codes. In contrast, the multilevel coded (MLC) modulation [6], [8] protects different bit levels using different binary codes. Zhang et al. [7] designed multi-edge type low-density parity-check (MET-LDPC) codes for MLC modulation for 16-quadrature amplitude modulation (16-QAM) and showed that the spectral efficiency can be improved when the differences in the bit-channel reliabilities are incorporated into the code design. Further, Hou et al. [6] showed that the non-binary higher-order modulation channel can be decomposed into equivalent binary-input component channels for each individual bit level in the signal constellation, which simplifies the analysis and design of the MLC and BICM modulation schemes. Density evolution [9] can then be used to design codes for each equivalent binary-input component channel.

In this work, we propose a general design framework for Raptor codes for higher-order modulation using a multi-edge framework. Since the multi-edge framework allows for differences in the bit channel reliabilities in the code ensemble, this enables us to design Raptor codes using the MLC scheme and also to perform a comprehensive analysis of the performance of the Raptor code with higher-order modulation using MET density evolution (MET-DE). The innovation of this work is that we simultaneously optimize distinct Raptor code degree distributions for distinct equivalent binary-input component channels using the multi-edge framework. This helps to improve the performance in higher-order modulation systems as we are incorporating the difference in the bit-channel reliabilities into the code design.

II. TARGET SYSTEM MODEL

In this work, we focus on Gray-labelled 16-QAM, which is equivalent to Gray-labelled 4-pulse-amplitude modulation (4-PAM) mapped independently to in-phase and quadrature components, and consider a higher-order MLC modulation scheme and parallel independent decoding strategy [6].

Consider a binary codeword, c with code length n. We can partition c into q-tuples. Let the bit position in each q-tuple, a_i, where a \in A and A \in {0,1}^q, be indexed by i, i = 1, ..., q, where i = 1 represents the left-most bit position. A_i^t represents the subset of all elements of A having value b \in {0,1} in bit position i. Then a Q-ary modulation scheme, where Q = 2^q, maps each a \in A, to a modulated symbol, x_a, from a signal constellation, X \subset \mathbb{R}, according to the Gray labelling scheme, and transmits over an additive white Gaussian noise (AWGN) channel. Let y denote the received symbol, which is the noisy version of x_a. At the receiver side, an a posteriori probability (APP) module uses y to compute the log-likelihood ratio (LLR) for the coded bits a_i, i = 1, ..., q as follows:

\[ v_i = \log \frac{P_r(a_i = 0|Y = y)}{P_r(a_i = 1|Y = y)} = \log \frac{\sum_{a \in A_i^t} p(Y = y|X = x_a)}{\sum_{a \in A_i} p(Y = y|X = x_a)} \]

The conditional density function, p(Y = y|X = x_a), for an AWGN channel output y is given by

\[ p(Y = y|X = x_a) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp \left( -\frac{(y - x_a)^2}{2\sigma^2} \right) \]

where \sigma^2 is the variance of the AWGN.

One of the difficulties in the design of bandwidth efficient systems with higher-order modulations is that equivalent binary input channels are not necessarily symmetric. Therefore the performance analysis and the code design using density evolution is complicated in this case, as it is not valid to assume that the all-zero codeword is transmitted at each bit level. To address this problem, Hou et al. [6] introduced the concept of an independent and identically distributed (iid) channel adapter, which produces an equivalent symmetric binary input channel with the same capacity. We incorporate this technique into the target system and design Raptor codes for higher-order modulation using the standard MET-DE method [9].

Note that the extension of the proposed method to other higher-order modulations is straightforward.

III. REPRESENTATION OF RAPTOR CODES IN THE MULTI-EDGE FRAMEWORK FOR 16-QAM

Consider a Raptor code, which is a serial concatenation of an inner Luby transform (LT) code with degree distribution...
The design of 16-QAM modulated Raptor codes is described using the multi-edge framework. For a Raptor code, the k bit information vector is encoded by a precode of rate $R_{\text{pre}} = k/n$ to generate n coded symbols, referred to as input bits. Then an LT encoder, linearly transforms this input bits into an infinite stream of LT coded symbols, referred to as output bits, which is transmitted through a channel. The realized rate of the LT code is defined as $R_{\text{LT}} = n/m$, where m is the average number of LT coded symbols required for a successful decoding.

Now we describe the new representation of 16-QAM modulated Raptor codes as MET-LDPC codes using a multi-edge framework. Since we consider the PLC scheme, we assign one edge-type to each distinct bit level and design distinct LT degree distributions, $\Omega(i)(x)$, $i = 1, 2, \ldots, q$, for distinct bit levels using the multi-edge framework. Since a Gray-labelled 16-QAM is equivalent to two Gray-labelled 4-PAM, there are only two distinct bit-channel densities. The Tanner graph of a 16-QAM modulated Raptor code ensemble in the multi-edge framework is shown in Fig. 1. This Tanner graph has 4 edge-types, where $\Pi_1$ is the precode and $\Pi_2$ represents the LT code. $\Pi_3$ and $\Pi_4$ are used to input two distinct bit-channel densities of 16-QAM into the LT code.

The 16-QAM modulated Raptor code ensemble using the multi-edge framework can be represented by two node-perspective multinomials associated with variable nodes and check nodes as follows:

$$L(r, x) = r_0 \sum_{i=1}^{i_{\text{max}}} \sum_{j_{v}=1}^{j_{\text{max}}} L_{[iv \ jv \ 0 \ 0]} x_{iv}^{jv} + r_1 L_{[0 \ 0 \ 1 \ 0]} x_3 + r_2 L_{[0 \ 0 \ 0 \ 1]} x_4,$$  \hspace{1cm} (1)

$$R(x) = \sum_{i=2}^{j_{c1\text{max}}} R_{[icj \ 0 \ 0 \ 0]} x_{icj}^{j_{c1}} + \sum_{j_{c1}=1}^{j_{c1\text{max}}} R_{[0 \ 0 \ 1 \ 0]} x_2^{j_{c1}} x_3 + \sum_{j_{c2}=1}^{j_{c2\text{max}}} R_{[0 \ 0 \ 0 \ 1]} x_2^{j_{c2}} x_4,$$  \hspace{1cm} (2)

where $x = [x_1, x_2, x_3, x_4]$ corresponds to each edge-type in the Tanner graph and $x_{icj}^{j_{c1}}$ is used to indicate the number of edges of the ith edge-type connected to a particular node. The vector $r = [r_0, r_1, r_2]$ associated with each variable node is used to denote bit-channel reliabilities. In the multi-edge framework, input bits of the Raptor code are considered as punctured variable nodes (i.e., which are not transmitting over the channel) and denoted by $r_0$. The output bits of the Raptor code are considered as unpunctured variable nodes (i.e., which are transmitted over two different channels) and denoted by $r_1, r_2$. Finally, $L_{[iv \ jv \ 0 \ 0]}$ and $R_{[icj \ 0 \ 0 \ 0]}$ are used to represent the fraction of variable nodes of type $[iv \ jv \ 0 \ 0]$ and the fraction of check nodes of type $[icj \ 0 \ 0 \ 0]$, where fractions are relative to the number of transmitted variable nodes and $[i \ j \ k \ m]$ denote the number of edges, from edge-types 1 to 4, connected to that node.

We add the following constraints into (1) and (2) to impose the Raptor code structure into the multi-edge parametrization:

1. $\sum_{i=1}^{i_{\text{max}}} \sum_{j_{v}=1}^{j_{\text{max}}} L_{[iv \ jv \ 0 \ 0]} = R_{\text{LT}},$
2. $\sum_{i=2}^{j_{c1\text{max}}} R_{[icj \ 0 \ 0 \ 0]} = R_{\text{LT}}(1 - R_{\text{pre}}),$
3. $L_{[0 \ 0 \ 1 \ 0]} = L_{[0 \ 0 \ 0 \ 1]} = 0.5,$
4. $\sum_{j_{c1}=1}^{j_{c1\text{max}}} R_{[0 \ 0 \ 1 \ 0]} = \sum_{j_{c2}=1}^{j_{c2\text{max}}} R_{[0 \ 0 \ 0 \ 1]} = 0.5.$

Constraints (2) and (3) are used to impose the rates of LT and precode components into the MET-LDPC code [10]. Constraints (4) and (5) are used to satisfy constraints on the total number of transmitted bits as fractions of code length.

In this work, we design distinct LT degree distributions for distinct bit levels for the 16-QAM modulated Raptor codes using the multi-edge framework. We can compute these LT degree distributions from (5) as follows:

$$\Omega^{(1)}(x) = \sum_{j_{c1}=1}^{j_{c1\text{max}}} R_{[0 \ 0 \ 1 \ 0]} x_{j_{c1}},$$  \hspace{1cm} (7)

$$\Omega^{(2)}(x) = \sum_{j_{c2}=1}^{j_{c2\text{max}}} R_{[0 \ 0 \ 0 \ 1]} x_{j_{c2}},$$  \hspace{1cm} (8)

where $\Omega^{(i)}(x)$ is the LT degree distribution related to bit level i. Note that these polynomials, $\Omega^{(1)}(x)$ and $\Omega^{(2)}(x)$, are not valid probability mass functions as their coefficients do not sum to one.

The rate efficiency of a Q-ary modulated Raptor code ensemble at signal-to-noise ratio (SNR), $\gamma$, can be computed as

$$\eta(\gamma) = \frac{R(\gamma)}{C(\gamma)} \log_2 Q,$$  \hspace{1cm} (9)

where $C(\gamma)$ is the capacity of the AWGN channel with Q-ary modulation at SNR, $\gamma$ [6] and $R(\gamma)$ is the designed rate of the Raptor code, which can be computed using [9, page 383]

$$R(\gamma) = L(1, 1) - H(1),$$  \hspace{1cm} (10)

where 1 denotes a vector of all 1’s with the length determined by the context.

IV. DESIGN OF 16-QAM MODULATED RAPTOR CODES USING THE MULTI-EDGE FRAMEWORK

A. Decoding of Raptor codes in the multi-edge framework

Generally, the decoding process of Raptor codes consists of a series of decoding attempts, which runs the LT decoder for a predetermined number of belief propagation (BP) decoding iterations, which is followed separately by the precode...
decoder. In this work, we instead consider a joint decoding scheme based on the BP decoding for the decoding of Raptor codes, where both component codes (LT code and precode) are decoded in parallel and extrinsic information (i.e., the information from previous BP decoding iteration) is exchanged between the decoders [10]. We consider the stability of Q-ary modulated Raptor codes and derived an exact expression for the stability condition when Raptor codes are decoded with joint decoding using the multi-edge framework, which is given in Theorem 1.

**Theorem 1:** Consider a Q-ary modulated Raptor code with joint decoding based on the BP decoding using the multi-edge framework. On the AWGN channel, the stability condition is given by

\[ \sum_{j \geq 1} \lambda_{[2, j]} (y_2)^j \rho_1(1) \leq 1, \]

where \( \lambda_{[2, j]} \) gives the fraction of degree-two variable nodes in the precod component and \( \rho_1(1) \) is the derivative of \( \rho_1(x) \), which is the edge-perspective check node degree distribution for the edge-type 1 (\( \Pi_1 \)). When \( x = 1 \), \( y_2 = \frac{1}{q} \sum_{i=1}^{q} x_0^{(i)} \), where \( x_0^{(i)} \) is the Bhattacharya constant [9, page 202] associated with \( i \) and \( q = \log_2 Q \).

**Proof:** See Appendix A.

Note that the stability condition of a Q-ary modulated Raptor code given in [10] having multiple channel inputs.

### B. Raptor code optimization for 16-QAM

The aim of the Raptor code optimization in the multi-edge framework is to find the largest possible realized rate, \( R(\gamma) \), for a given SNR, \( \gamma \), over the check node degree distribution, \( R(x) \), under the constraints given in (1) to (3). Then we can find the corresponding variable node degree distribution, \( L(r, x) \), using the rate constraint given in (10) and the socket count equality constraint which is given by

\[ L_{x_1}(1, 1) = \sum_{i_1=1}^{i_{\text{max}}} \sum_{j_0=1}^{j_{\text{max}}} L[i_1 j_0 0 0] = R_{\text{LT}}, \]

\[ R_{x_1}(1) = \frac{dR(r, x)}{dx}, \]

The optimization problem for 16-QAM modulated Raptor codes for a given SNR, \( \gamma \), in the multi-edge framework can be summarized as follows:

**Maximize** \( R(\gamma) \)

**s.t.**

1. \( \sum_{i_1=1}^{i_{\text{max}}} \sum_{j_0=1}^{j_{\text{max}}} L[i_1 j_0 0 0] = R_{\text{LT}} \)
2. \( \sum_{i_1=1}^{i_{\text{max}}} R[i_1 0 0 0] = R_{\text{LT}}(1 - R_{\text{pre}}) \)
3. \( L[0 0 0 1] = 0.5 \)
4. \( \sum_{j_1=1}^{j_{\text{max}}} R[0 j_1 1 0] = 0.5 \)
5. \( L_{x_1}(1, 1) = R_{x_1}(1) \)
6. \( \max(\epsilon_j) \leq \epsilon^* \)

where \( \epsilon_j \) and \( \epsilon^* \) respectively denote the bit error rate (BER) on the jth variable node and target maximum BER. We use combined optimization method proposed in [11] to optimize the node fractions and differential evolution [12] to optimize Raptor code ensembles, which are represented in the multi-edge framework, and compute the realized rate \( R(\gamma) \) using MET-DE.

### V. RESULTS AND DISCUSSION

We formulate the optimization problem as per Section IV-B and optimize Raptor codes for 16-QAM using MLC modulation. For the Raptor code design we set the precode to a (3,60)-regular LDPC code with a rate of 0.95, and set the maximum output node degree to be 50. The optimized LT degree distributions and their rate efficiencies are shown in Table 1. In Fig. 2 we evaluate the rate efficiency performance for the optimized Raptor code ensembles for SNRs above and below the designed SNR using MET-DE. The rate efficiency performance for two reference Raptor codes from the literature [3][4] are also shown in the same figure for comparison. The reference code 1 [5] was designed using BICM for Graylabelled 16-QAM with a precode of (3,30)-regular LDPC code with a rate of 0.9. Venkiah [4] designed Raptor codes for a binary input AWGN channel with a precode of (3,60)-regular LDPC code and use those codes to evaluate the performance with 16-QAM. Thus we chose the Raptor degree distribution from [4] as shown in the same figure for comparison. Note that the stability condition of a Raptor-MLC code ensemble in the multi-edge framework is given in [10], which is similar to the stability condition of binary modulated Raptor codes given in [10] having multiple channel inputs.

### VI. CONCLUSION

In this paper, we represented Raptor codes as multi-edge type low-density parity-check (MET-LDPC) codes, to design higher-order modulated Raptor codes using multi level coded (MLC) modulation. The MET representation enables us to design distinct degree distributions for distinct bit channel densities using MLC modulation and achieve better rate efficiency performance compared to the previously reported Raptor codes in the literature. Moreover, in contrast to 16-QAM modulated MET-LDPC codes [7] Table 4.2, which are fixed-rate codes, Raptor codes can achieve good rate efficiency performance in the entire SNR range. This shows the benefit of considering Raptor codes for the higher-order modulation compared to traditional fixed-rate coding schemes.

### APPENDIX A

**PROOF OF THEOREM 1**

As indicated in (1), the variable node degree distribution of a Raptor code ensemble in the multi-edge framework has degree-one variable nodes. Thus we consider the special case given in [9] pages 396-397 and define nodes connected to \( \Pi_1 \) as the core LDPC graph for the stability analysis. The edge-perspective multinomial of the core LDPC graph with edges connected to \( \Pi_2 \) can be computed using (1) and (2) as follows:
TABLE I

| γd | Degree distribution | η(γd) |
|----|---------------------|------|
| 4 dB | \( Ω^{(1)}(x) = 0.101 x + 0.1856 x^2 + 0.0948 x^3 + 0.1425 x^4 + 0.00878 x^5 + 0.0078 x^6 + 0.0184 x^7 + 0.03005 x^8 + 0.0024 x^9 \) | 0.9345 |
| 6 dB | \( Ω^{(1)}(x) = 0.01 x + 0.1015 x^2 + 0.1275 x^3 + 0.02195 x^4 + 0.00175 x^5 + 0.00153 x^6 + 0.00012 x^7 + 0.00002 x^8 \) | 0.9519 |
| 8 dB | \( Ω^{(1)}(x) = 0.01375 x + 0.3361 x^2 + 0.1348 x^3 + 0.0566 x^4 + 0.0019 x^5 + 0.00027 x^6 + 0.0118 x^7 + 0.00175 x^8 \) | 0.9759 |
| 10 dB | \( Ω^{(1)}(x) = 0.21975 x + 0.05065 x^2 + 0.2076 x^3 + 0.0146 x^4 + 0.0066 x^5 + 0.00005 x^6 + 0.003 x^7 + 0.0003 x^8 \) | 0.9646 |

where \( a_0(i) \) is the bit channel density corresponding to bit level \( i \) and \( q = \log_2 Q \). Then by applying Lemma 4.63 in [9] pages 202) to (13), we get the final update rule for \( x_1^{(ℓ+1)} \) as follows:

\[
x_1^{(ℓ+1)} = \left( 1 - \sum_{j=1}^{q} \lambda[j] (1 - x_1^{(ℓ)})^{k-1} \right) \left( \frac{y_2^{(ℓ)}}{y_1^{(ℓ)}} \right)^j,
\]

where \( y_2^{(ℓ)} = \frac{1}{q} \sum_{i=1}^{q} x_0(i) \) and \( x_0(i) \) is the bit channel density associated with \( a_0(i) \). Furthermore, to ensure that the BER decreases throughout the BP decoding iterations condition, \( x_1^{(ℓ+1)} < x_1^{(ℓ)} \) must hold. Thus for a successful decoding under MET-DE, we need to guarantee that

\[
x_1^{(ℓ)} > \sum_{j=2}^{q} \lambda[j] (1 - x_1^{(ℓ)})^{k-1} \left( \frac{y_2^{(ℓ)}}{y_1^{(ℓ)}} \right)^j.
\]

For the decoding to be successful, inequality (15) needs to be true around \( x_1^{(ℓ)} = 0 \). Thus taking the derivative of (15) with respect to \( x_1^{(ℓ)} = 0 \) gives us

\[
\frac{d}{dx} \left( \sum_{j=2}^{q} \lambda[j] (1 - x_1^{(ℓ)})^{k-1} \left( \frac{y_2^{(ℓ)}}{y_1^{(ℓ)}} \right)^j \right) \bigg|_{x_1^{(ℓ)}=0} = \sum_{j=2}^{q} \lambda[j] (y_2^{(ℓ)})^j \frac{d y_1^{(ℓ)}}{dx_1^{(ℓ)}} \bigg|_{x_1^{(ℓ)}=0} = \rho_1 \left( \frac{d y_1^{(ℓ)}}{dx_1^{(ℓ)}} \bigg|_{x_1^{(ℓ)}=0} \right) \leq 1.
\]

This completes the proof.

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