Four-Dimensional Coded Modulation with Bit-wise Decoders for Future Optical Communications

Alex Alvarado and Erik Agrell

Abstract—Coded modulation (CM) is the combination of forward error correction (FEC) and multilevel constellations. Coherent optical communication systems result in a four-dimensional (4D) signal space, which naturally leads to 4D-CM transceivers. A practically attractive design paradigm is to use a bit-wise decoder, where the detection process is (suboptimally) separated into two steps: soft-decision demapping followed by binary decoding. In this paper, bit-wise decoders are studied from an information-theoretic viewpoint. 4D constellations with up to 4096 constellation points are considered. Metrics to predict the post-FEC bit-error rate (BER) of bit-wise decoders are analyzed. The mutual information is shown to fail at predicting the post-FEC BER of bit-wise decoders and the so-called generalized mutual information is shown to be a much more robust metric. For the suboptimal scheme under consideration, it is also shown that constellations that transmit and receive information in each polarization and quadrature independently (e.g., PM-QPSK, PM-16QAM, and PM-64QAM) outperform the best 4D constellations designed for uncoded transmission. Theoretical gains are as high as 4 dB, which are then validated via numerical simulations of low-density parity check codes.

Index Terms—Bit-interleaved coded modulation, bit-wise decoders, channel capacity, coded modulation, fiber-optic communications, nonlinear distortion, low-density parity-check codes.

I. INTRODUCTION AND MOTIVATION

In coherent fiber-optic communication systems, both quadratures and both polarizations of the electromagnetic field are used. This naturally results in a four-dimensional (4D) signal space. To meet the demands for spectral efficiency, multiple bits should be encapsulated in each constellation symbol, resulting in multilevel 4D constellations. To combat the decreased sensitivity caused by multilevel modulation, forward error correction (FEC) is used. The combination of FEC and multilevel constellations is known as coded modulation (CM).

The most popular alternatives for CM are trellis-coded modulation (TCM) [1], multilevel coding (MLC) [2], and bit-interleaved coded modulation (BICM) [3]–[5]. TCM has been considered for optical communications in [6]–[10] and MLC in [11]–[15]. Regardless of the paradigm used at the transmitter (see [16, Fig. 3] for a schematic comparison), the optimum receiver structure is the maximum likelihood (ML) decoder. The ML decoder finds the most likely transmitted sequence, where the maximization is over all possible coded sequences. The ML solution is in general impractical1, and thus, suboptimal alternatives are preferred. One pragmatic and popular approach is BICM, which we study in this paper.

The key feature of BICM is a suboptimal decoder that operates on bits rather than on symbols. We refer to this receiver structure as a bit-wise (BW) decoder. In a BW decoder, the detection process is decoupled: first soft information on the bits (logarithmic likelihood ratios, LLRs) is calculated in a demapper and then a soft-decision FEC (SD-FEC) decoder is used. BW decoders are very flexible, where the flexibility is due to the use of off-the-shelf binary encoders and decoders. In the context of optical communications, a BW decoder for binary modulation and low-density parity check (LDPC) codes was studied in [17], where a finite-state machine and a histogram-based estimation of the channel was used to compute LLRs. A BW decoder with multilevel modulation and LDPC codes was considered in [18]. An LDPC-based BW decoder with a 24-dimensional constellation was experimentally demonstrated in [19]. Optimized mappings between code bits and constellation symbols for protograph-based LDPC codes were recently presented in [20].

To improve upon simple BW decoders, iterations between the binary FEC decoder and demapper can be included. In such a configuration, the FEC decoder and demapper iteratively exchange information on the code bits. This is usually known as BICM with iterative demapping (BICM-ID). BICM-ID for optical communications has been studied in [21]–[23], [24], Sec. 3, [25], Sec. 3, [26], Sec. 4. BICM-ID offers remarkable improvements with demapper iterations. These gains are typically obtained by custom-tailoring the constellation and its binary labeling to the channel and the encoder–decoder pair as well as the iteration scheduling [26]. In BICM-ID, iterations between the decoder and demapper are added to a possibly already iterative FEC decoder and to keep the number of iterations low, one can trade FEC decoder iterations for demapper iterations. However, this leads to nontrivial designs which reduce flexibility. On the positive side, BICM-ID is expected to perform very close to an ML sequence detector, and thus, to outperform BICM. To the best of our knowledge, no exact complexity-performance tradeoff analyses providing a clear-cut answer about BICM vs. BICM-ID exist. In this paper, we focus on BICM because of its simplicity and flexibility.

1A notable exception is TCM, where the FEC encoder is a convolutional encoder and the resulting CM code has a trellis structure, which allows an ML decoder based on the Viterbi algorithm to be implemented.
CM transceivers are typically based on quadrature amplitude modulation (QAM) or phase shift keying (PSK). Traditional constellations include polarization-multiplexed (PM) quadrature phase-shift keying (PM-QPSK), PM-16QAM, and PM-64QAM. However, recent years have seen an increased interest in formats that use the available four dimensions more efficiently than by pure multiplexing. Polarization-switched QPSK (PS-QPSK) was shown in [27, Fig. 1] to be the most power-efficient 8-ary 4D constellation. Power efficiency should here be understood as the energy per bit for a given minimum Euclidean distance between constellation points. This is the classical sphere packing problem, which has been used to optimize constellation formats for uncoded transmission since the 1970’s [28]–[30]. It arises when minimizing either the pre-FEC bit error rate (BER) or the symbol error rate for the additive white Gaussian noise (AWGN) channel at asymptotically high signal-to-noise ratio (SNR) [31], [32, Sec. 5.1]. 4D constellations optimized in this sense were compared in [31]. Spherically shaped 4D constellations based on the $D_4$ lattice were studied in, e.g., [23], [33]. Somewhat less power efficient, but easier to implement, are the cubically shaped constellations based on $D_4$, called set-partitioning QAM [33], [34]. Other irregular constellations include the amplitude phase-shift keying constellation optimized for channels with strong nonlinear phase noise in [35]–[37].

Of particular interest for this paper is the constellation $C_{4,16}$ introduced in [38], which is the most power efficient 16-ary 4D constellation known. Another constellation we will study in this paper is subset-optimized PM-QPSK (SO-PM-QPSK) introduced in [39] as an alternative to $C_{4,16}$ with lower complexity. In terms of power efficiency, $C_{4,16}$ and SO-PM-QPSK offer asymptotic gains over PM-QPSK of 1.11 dB and 0.44 dB, respectively. The asymptotic gains offered by $C_{4,16}$ have been experimentally demonstrated in [40], [41]. We also consider the power-efficient 4D constellations $C_{4,256}$ [42, Table. IV], [43, Table I] and $C_{4,4096}$ [30], which are, respectively, the best known 256-ary and 4096-ary constellations.

The performance of a BW decoder based on hard decisions (HDs) can be accurately characterized by the pre-FEC BER. In this paper, we study SD-FEC, i.e., when LLRs are passed to the soft-input FEC decoder, and thus, we question the optimality of constellations designed in terms of pre-FEC BER. Furthermore, we show that a different metric is more relevant for capacity-approaching SD-FEC encoder–decoder pairs: the so-called generalized mutual information (GMI).

Achievable rates provide an upper bound on the number of bits per symbol that can be reliably transmitted through the channel. From an information-theoretic point of view, a BW decoder does not implement the ML rule, and thus, a penalty in terms of achievable rates is expected. While the mutual information (MI) is the largest achievable rate for any receiver, for a BW decoder, this quantity is replaced by the GMI [5, Sec. 3]. Although the MI and the GMI coincide when the SNR tends to infinity, for any nontrivial case, the MI is strictly larger than the GMI for any finite SNR. This penalty, which depends on the constellation and its binary labeling, can be very large [4, Fig. 4]. The MI has been considered as the figure of merit for optical communications in [14], [38], [40], [45]–[49]. To the best of our knowledge, however, the GMI has been considered in optical communications only in [24].

One problem often overlooked when designing 4D-CM with a BW decoder is the problem of choosing an appropriate binary labeling for the constellation. Finding good labelings based on brute force approaches quickly fails, as the number of binary labelings grows factorially with the constellation size. For example, for the relatively simple case of 16 constellation points, there are about $2 \cdot 10^{13}$ different binary labelings. When regular constellations (QAM, PSK, etc.) are considered, a Gray code is typically used, as Gray codes have been proven to be asymptotically optimum in terms of pre-FEC BER [51]. This conclusion holds only in the regime of asymptotically large SNR and only for the AWGN channel. The problem is considerably more difficult when the GMI is the cost function. Although results in the asymptotic regimes exist (see [52]–[56] and [57] for low and high SNR, respectively), finding the optimal binary labeling in terms of GMI for a finite SNR remains as an open research problem.

In this paper, achievable rates for 4D constellations with a BW decoder in the context of future generation coherent optical communication systems are studied. It is shown that the GMI is the correct metric to predict the post-FEC BER for a BW decoder. It is also shown that constellations that are good for uncoded systems are also good in terms of MI if the SNR is sufficiently high. These constellations, however, are not the best choice for coded systems based on a BW decoder. Numerical results based on LDPC codes confirm the theoretical analysis.

The remainder of this paper is organized as follows. In Sec. II, the system model is introduced and achievable rates are reviewed. Post-FEC BER prediction based on the GMI is studied in Sec. III and numerical results on achievable rates are shown in Sec. IV. Conclusions are drawn in Sec. V.

II. System Model and Achievable Rates

In Fig. 1, a generic structure of the CM transceiver we study in this paper is shown. At the transmitter, a rate-$R_c$ binary FEC encoder encodes a binary input sequence $U$ into $m$ binary sequences $\mathbf{B}_1, \ldots, \mathbf{B}_m$, where $\mathbf{B}_k = [B_{1,k}, \ldots, B_{N_c,k}]$ for $k = 1, \ldots, m$ and $N_c$ is the symbol block length.\footnote{The term GMI was coined by Martinez et al. in [44], where the BW decoder was recognized as a mismatched decoder. The GMI is known in the literature under different names such as “parallel decoding capacity”, “receiver constrained capacity”, and “BICM capacity.”}

\footnote{Throughout this paper, vectors are denoted by boldface letters $\mathbf{x}$, sequences of vectors by underlined boldface letters $\underline{x}$ and sets by calligraphic letters $\mathcal{X}$. Random variables, vectors, and sequences are denoted by uppercase letters and their outcomes by the same letter in lowercase. Probability density functions and conditional probability density functions are denoted by $f_{\mathbf{x}}(\mathbf{y})$ and $f_{\mathbf{Y} \mid \mathbf{X}}(\mathbf{y} | \mathbf{x})$, respectively. Expectations are denoted by $\mathbb{E}[\cdot]$.} A $3$Also known in the literature as dual-polarization QPSK (DP-QPSK) and polarization-division-multiplexed QPSK (PDM-QPSK).
memoryless mapper then maps $B_1, \ldots, B_m$ into a sequence of symbols $X = [X_1, X_2, \ldots, X_M]$, one symbol at a time. After transmission over the physical channel, the received symbols $Y = [Y_1, Y_2, \ldots, Y_M]$ are processed by the CM decoder, which gives an estimate of the transmitted information sequence $\hat{U}$.

We consider the discrete-time, memoryless, vectorial AWGN channel

$$Y_n = X_n + Z_n$$

where $X_n, Y_n, Z_n$ are 4D real vectors and $n = 1, 2, \ldots, N_s$ is the discrete-time index. The components of the noise vector $Z_n$ are independent, zero-mean, Gaussian random variables with variance $N_0/2$ in each dimension, and thus,

$$f_{Y_n|X_n}(y|x) = \frac{1}{(\pi N_0)^2} \exp\left(-\frac{\|y - x\|^2}{N_0}\right).$$  \hspace{1cm} (2)

The communication channel in Fig. 1 encompasses all the transmitter digital signal processing (DSP) used after the bit-to-symbol mapping (i.e., pulse shaping, polarization multiplexing, filtering, electro-optical conversion, etc.), the physical channel (the fiber, amplifiers, regenerators, etc.), and the receiver DSP (optical-to-electrical conversion, filtering, equalization, digital back-propagation, matched filtering, etc.). The use of the AWGN channel in (1) to model all these blocks can be justified in amplified spontaneous emission noise dominated links where chromatic dispersion and polarization mode dispersion are perfectly compensated. The AWGN assumption also holds for uncompensated coherent systems where the so-called GN model has been widely used (see [47] and references therein).

At each time instant $n$, the transmitted vector $X_n$ is selected with equal probability from a constellation $\mathcal{S} \triangleq \{s_1, s_2, \ldots, s_M\}$, where $M = 2^m$. The average symbol energy is $E_s = \mathbb{E}[\|X\|^2] = (1/M) \sum_{i=1}^{M} \|s_i\|^2$ and the SNR is defined as $\gamma \triangleq E_s/N_0$. For a rate $R_c$ FEC encoder, the spectral efficiency in bits/symbol is $\eta = R_c m$. The length of the information sequence $U$ is $N_b = \eta N_s$ and the average bit energy is $E_b = E_s/\eta$.

The transmitter in Fig. 1 is a one-to-one mapping between the information sequence $U \in \{0, 1\}^{N_b}$ and the coded sequence $X \in \mathcal{C} \subseteq \mathcal{S}^{N_s}$, where $|\mathcal{C}| = 2^{N_b}$. The set $\mathcal{C}$ is called the codebook, and the mapping between the $2^{N_b}$ information sequences and the code $\mathcal{C}$ is called the CM encoder. At the receiver side, a CM decoder (see Fig. 1) uses the mapping rule used at the transmitter (as well as the channel characteristics) to give an estimate of the information sequence. The triplet codebook, encoder, and decoder forms a so-called coding scheme. Practical coding schemes are designed so as to minimize the probability that $\hat{U}$ differs from $U$, while at the same time keeping the complexity of both encoder and decoder low.

### A. CM Decoder Structures

Fig. 2 shows two possible receiver structures for the CM encoder in Fig. 1 together with the AWGN channel in (2): the optimal ML decoder and the (suboptimal) BW decoder. The ML decoder operates on the sequence of symbols $Y$ and finds the most likely coded sequence, i.e., it performs $\hat{u}_k = \arg \max_u f_{Y|X}(y|x)$. On the other hand, the BW decoder computes soft information on the code bits $B_1, \ldots, B_m$ on a symbol-by-symbol basis. This soft information is typically represented in the form of LLRs $\Lambda_{k,1}, \ldots, \Lambda_{k,4}$, where $\Lambda_k = [\Lambda_{1,k}, \ldots, \Lambda_{4,k}]$ for $k = 1, \ldots, N_s$. These LLRs are then passed to a binary SD-FEC decoder.

Assuming perfect knowledge of $N_0$, at each discrete-time instant $n$, $m$ LLRs are calculated as

$$\Lambda_{n,k} \triangleq \log \frac{f_{Y_n|B_{n,k}}(y|1)}{f_{Y_n|B_{n,k}}(y|0)} \hspace{1cm} (3)$$

$$= \log \frac{\sum_{s \in S_{k,0}} \exp\left(-\frac{1}{N_0} \|y - s\|^2\right)}{\sum_{s \in S_{k,0}} \exp\left(-\frac{1}{N_0} \|y - s\|^2\right)} \hspace{1cm} (4)$$

$$\approx \frac{1}{N_0} \min_{s \in S_{k,0}} \|y - s\|^2 - \frac{1}{N_0} \min_{s \in S_{k,1}} \|y - s\|^2 \hspace{1cm} (5)$$

where (4) follows from (2) and $S_{k,1,0} \subseteq \mathcal{S}$ is the set of constellation symbols labeled with a bit $b \in \{0, 1\}$ at bit

---

6Alternatively, an HD demapper can be combined with an HD-FEC decoder. In this paper, we only consider SD-FEC decoders.
position \(k \in \{1, \ldots, m\}\). The approximation in (5) follows from using the so-called max-log approximation [58].

Alternatively, the LLRs in (3) can be defined as

\[
\Lambda_{n,k} = \log \frac{f_{Y_n|B_n,k}(y|0)}{f_{Y_n|B_n,k}(y|1)},
\]

which could have some advantages in practical implementations. For example, in a popular complement-to-two binary format, the most significant bit carries the sign, i.e., when an MSB is equal to zero (0), it means that a number is positive, and when an MSB is equal to one (1), it means that the number is negative. Then, if (6) is used, the transmitted bit obtained via HDs can be recovered directly from the MSB.

Without loss of generality, in this paper we use the definition in (3). Furthermore, since the mapper, channel, and demapper are all memoryless, the time index \(n\) is dropped from now on.

Throughout this paper we denote the pre-FEC BER and the post-FEC BER by \(\text{BER}_{\text{pre}}\) and \(\text{BER}_{\text{pos}}\), respectively. \(\text{BER}_{\text{pre}}\) can be obtained from the max-log LLRs in (5) as [59, Theorem 1]

\[
\text{BER}_{\text{pre}} = \frac{1}{m} \sum_{k=1}^{m} \frac{1}{2} \sum_{b \in \{0,1\}} \int_0^\infty \Lambda_{b|B_n,k} ((-1)^b \lambda) \mathrm{d}\lambda
\]

and depends only on the constellation, its binary labeling and the communication channel.\(^7\) On the other hand, \(\text{BER}_{\text{pos}}\) also depends on the choice of FEC code.

The BW decoder in Fig. 2 is usually known as a BICM receiver/decoder, owing its name to the original works [3], [4], where a bit-level interleaver was included between the FEC encoder and mapper. We refrain from using such a name because the interleaver might or might not be included, and if included, it can be assumed to be part of the FEC encoder.

### B. Achievable Rates

A rate \(R\) (in bits/symbol) is said to be achievable at block length \(N_b\) and average error probability \(\epsilon\) if there exists a coding scheme, consisting of a codebook \(C\), an encoder, and a decoder, such that \(|C| = 2^{RN_b}\) and \(\Pr\{U \neq \hat{U}\} \leq \epsilon\).

The largest achievable rate at given \(N_b\) and \(\epsilon\) is denoted by \(R^*(N_b, \epsilon)\). The channel capacity \(C\) is the largest achievable rate for which a coding scheme with vanishing error probability exists, in the limit of large block length [60, Sec. 1 and 14], i.e.,

\[
C \triangleq \lim_{\epsilon \to 0} \lim_{N_b \to \infty} R^*(N_b, \epsilon).
\]

The channel capacity is often defined subject to an average power constraint \(P\), which means that every codeword \(X = [X_1, \ldots, X_{N_b}] \in C\) must satisfy \(\sum_n \|X_n\|^2 \leq P\).

For memoryless channels and a given constellation \(S\), the largest achievable rate is the MI between \(X\) and \(Y\) defined as

\[
I(X;Y) \triangleq \mathbb{E} \left[ \log_2 \frac{f_{Y|X}(Y|X)}{f_Y(Y)} \right].
\]

By Shannon’s channel coding theorem, the channel capacity of a discrete-time memoryless channel with an average power constraint can be calculated as [60], [61, Ch. 7]

\[
C = \sup_{f_x, E_x \leq P} I(X;Y)
\]

where \(I(X;Y)\) is the MI in (9) and the maximization in (10) is over all distributions\(^3\) of \(X\) that satisfy the average power constraint \(E_x \leq P\), for a given channel \(f_{Y|X}\).

For the 4D channel in (1), (10) gives

\[
C = \frac{N}{2} \log_2 \left( 1 + \frac{2}{N} \gamma \right) = 2 \log_2 \left( 1 + \frac{\gamma}{2} \right)
\]

which is attained by a zero-mean Gaussian input distribution \(f_X\) with a diagonal covariance matrix with all diagonal entries equal to \(E_x/4 = P/4\).

In this paper, we consider equally likely symbols and discrete constellations \(S\), and thus, \(f_X\) is a uniform distribution over \(S\). In this case, the MI in (9) becomes

\[
I(X;Y) = \frac{1}{M} \sum_{s \in S} \int_{\mathbb{R}^4} f_{Y|X}(y|s) \log_2 \frac{f_{Y|X}(y|s)}{f_Y(y)} \, dy.
\]

The MI \(I(X;Y)\) in (12) is the largest achievable rate for the optimum ML decoder and a given constellation \(S\). Thus, for the optimal ML decoder, reliable transmission with arbitrarily low error probability is possible if \(\eta < I(X;Y)\). By Shannon’s random coding paradigm, the rate in (12) is achievable using a codebook \(C\) consisting of \(2^{RN_b}\) codewords of length \(N_b\), each symbol drawn independently and uniformly from \(S\).

When the BW decoder in Fig. 2 is considered, due to the fact that this decoder is not ML, the largest achievable rate is unknown. The most popular achievable rate for the BW decoder is the GMI\(^\dagger\) [44]

\[
I_{\text{GMI}}^m = \sum_{k=1}^{m} I(B_k;Y)
\]

where

\[
I(B_k;Y) = \mathbb{E} \left[ \log_2 \frac{f_{Y|B_k}(Y|B_k)}{f_Y(Y)} \right].
\]

In analogy with (12), we consider in this paper independent, equally likely bits, in which case the GMI in (13) becomes

\[
I_{\text{GMI}}^m = \sum_{k=1}^{m} \sum_{b \in \{0,1\}} \int_{\mathbb{R}^4} f_{Y|B_k}(y|b) \log_2 \frac{f_{Y|B_k}(y|b)}{f_Y(y)} \, dy.
\]

This rate is achievable with a BW decoder, without iterative decoding, using the same codebook \(C\) that achieves (12) with an optimum decoder. Note that designing a codebook by drawing symbols independently and uniformly from \(S\) corresponds to independent and equally likely bits \(B_k\).

\(^7\)Note that HDs on the exact LLRs in (4) give slightly worse pre-FEC BER results in the low-SNR regime.

\(^3\)In general, the capacity-achieving distribution can be discrete, continuous, or mixed.

\(^\dagger\)The GMI is not necessarily the largest achievable rate for the receiver in Fig. 2. Other achievable rates include the so-called LM rate [62, Part 1] and the newly derived rate for unequally likely constellation points (i.e., probabilistic shaping) [63, Theorem 1].
When the LLRs are calculated using (4), it can be shown that [64, Theorem 4.21]
\[ I_{\text{gmi}} = \sum_{k=1}^{m} I(B_k; \Lambda_k) = \sum_{k=1}^{m} \mathbb{E} \left[ \log_2 \frac{f_{\Lambda_k|B_k}(\Lambda_k|B_k)}{f_{\Lambda_k}(\Lambda_k)} \right]. \tag{16} \]

When the LLRs are calculated using (5), the resulting achievable rate is smaller than \( I_{\text{gmi}} \) in (16). Under certain conditions, this loss can be recovered by correcting the max-log LLRs, as shown in [65], [66] (see also [64, Ch. 7]).

Achievable rates for BW decoders were first analyzed in [4]. The BW decoder was later recognized in [44] as a mismatched decoder, where it was shown that the GMI in (13) is an achievable rate. It was also shown in [44] that in terms of achievable rates, the interleaver plays no role, and that the key element is the suboptimal (mismatched) decoder.

The GMI in (13) is an achievable rate for BW decoders but has not been proven to be the largest achievable rate. Finding the largest achievable rate remains as an open research problem. Despite this cautionary statement, the GMI has been shown to predict very well the performance of BW decoders based on capacity-approaching FEC encoder–decoder pairs. This has been shown for example in [67, Sec. V], [68, Sec. V-D], and [69, Sec. IV]. Generally speaking, when good turbo or LDPC codes are used, the gap between the coded system and the GMI prediction is usually less than 1 dB.

The mapper is one-to-one, and thus, \( I(B; Y) = I(X; Y) \). The chain rule of MI [61, Sec. 2.5] gives
\[ I(B; Y) \geq \sum_{k=1}^{m} I(B_k; Y) \tag{17} \]
and thus,
\[ I_{\text{gmi}} \leq I(X; Y). \tag{18} \]

The difference \( I(X; Y) - I_{\text{gmi}} \) can be understood as the loss in terms of achievable rates caused by the use of a BW decoder. Furthermore, the GMI (unlike the MI) is highly dependent on the binary labeling. Gray codes are known to be good for high SNR [4, Fig. 4], [55], [70, Sec. IV], but for many constellations, they do not exist.

Closed-form expressions for the MI and GMI are in general unknown, and thus, numerical methods are needed. For the AWGN channel, both MI and GMI can be efficiently calculated based on Gauss–Hermite quadrature. To this end, the ready-to-use expressions in [70, Sec. III] can be used. The GMI can also be calculated using the approximation recently introduced in [69]. This approximation is particularly useful to find good binary labelings in terms of GMI. When the channel is unknown or when the dimensionality of the constellation grows, Monte Carlo integration is preferred.

III. POST-FEC BER PREDICTION VIA GMI

In this section, we consider the problem of predicting the decoder’s performance for a given code rate. To this end, we first introduce the concept of the BICM channel (see [71, Fig. 1], [72, Fig. 1]). The BICM channel\footnote{Also called “modulation channel” in [68, Fig. 1].} encompasses all the elements that separate the encoder and decoder (see Figs. 1 and 2), i.e., the mapper and demapper, transmitter and receiver DSP, fiber, amplifiers, filtering, equalization, etc. The BICM channel is then what the encoder–decoder pair “sees”.

In principle, to predict the post-FEC BER of a given encoder over different BICM channels (e.g., different constellations, different amplification schemes, different fiber types, etc.), the whole communication chain should be re-simulated. To avoid this, one could try to find an easy-to-measure metric that characterizes the BICM channel and hope that different channels with the same metric result in the same BER\(_{\text{post}}\). Here we consider four different metrics and argue that the GMI in (13) is the most appropriate one.

Consider the irregular repeat-accumulate LDPC codes proposed by the second generation satellite digital video broadcasting standard [73] and the 6 code rates \( R_c \in \{1/3, 2/5, 1/2, 3/5, 3/4, 9/10\} \) \tag{19} which correspond to the FEC overheads \( \{200, 150, 100, 66.6, 33.3, 11.1\} \% \). Each transmitted block consists of 64 800 code bits which are randomly permuted before being cyclically assigned to the binary sequences \( \vec{B}_1, \ldots, \vec{B}_n \). At the receiver, LLRs \( \Lambda_k \) are calculated using (4) and passed to the SD-FEC decoder, which performs 50 iterations.

Fig. 3 shows the performance of the LDPC decoder with PM-QPSK, PM-16QAM, PM-64QAM, and PM-256QAM as a function of SNR. There are 24 different coding and modulation pairs, leading to 24 spectral efficiencies \( \eta = R_c m \). The results in this figure show that, for any given code rate, different modulations have very different SNR requirements. For example, for \( R_c = 3/5 \) and a target post-FEC BER of 10\(^{-4}\), the SNR thresholds are 5.1 dB, 10.8 dB, 15.5 dB and 20 dB for PM-QPSK, PM-16QAM, PM-64QAM, and PM-256QAM, respectively. This leads to the obvious conclusion that SNR cannot be used to predict the post-FEC BER performance of a given code when used with different constellations.

Under some assumptions on independent errors within a block,\footnote{This can be guaranteed by properly interleaving the code bits.} the pre-FEC BER in (7) can be used to predict the post-FEC BER of HD-FEC decoders. Based on such relations, the conventional design paradigm in optical communications is to design systems for a certain required pre-FEC BER, the so-called FEC limit or FEC threshold, which is typically in the range 10\(^{-4}\) – 10\(^{-3}\). The HD-FEC decoder is then assumed to bring down the post-FEC BER to, say, 10\(^{-12}\) or 10\(^{-15}\), without actually including any coding in simulations or experiments.

For a given (fixed) BICM channel, the pre-FEC BER can also be used to predict the post-FEC BER of an SD-FEC decoder. This has been done for example for some of the SD-FEC decoders in the G.975.1 standard [74], where post-FEC BER values are given as a function of pre-FEC BER. There is nothing fundamentally wrong with presenting post-FEC BER as a function of pre-FEC BER. However, more often than not, reported uncoded experiments or simulations rely on these tabulated values and claim (without encoding and
decoding information) the existence of an SD-FEC decoder that can deal with the measured pre-FEC BER. The caveat with this approach is that it relies on the strong assumption that the same SD-FEC encoder and decoder pair will perform identically for two different BICM channels which happen to have the same pre-FEC BER.

To study the robustness of the pre-FEC BER as a metric to predict post-FEC BER, we show in Fig. 4 BER$_{pre}$ as a function of BER$_{pre}$ for the same 24 combinations of constellations and codes as in Fig. 3. Ideally, all lines corresponding to the same code rate should fall on top of each other, indicating that measuring BER$_{pre}$ is sufficient to predict the post-FEC BER when the BICM channel changes (in this case, due to the change in modulation format). The results in this figure show that the curves get “grouped” for the same code rate, and thus, BER$_{pre}$ is a better metric than SNR (cf. Fig. 3). The results in Fig. 4 also show that BER$_{pre}$ is a good metric for very high code rates. For low and moderate code rates, however, BER$_{pre}$ fails to predict the performance of the decoder. The implication of this is that measuring pre-FEC BER cannot be used to predict the post-FEC BER of an encoder–decoder pair across different BICM channels. The FEC-limit design paradigm fails.

In Fig. 5, we consider BER$_{pos}$ as a function of the (normalized) MI \(I(X;Y)/m\) for the 24 cases in Fig. 3 (same markers). Based on the analysis in Sec. II-B, we propose here to study BER$_{pos}$ as a function of the GMI. The information-theoretic rationale behind this idea is that a SD-FEC decoder is fed with this approach is that it relies on the strong assumption that the same SD-FEC encoder and decoder pair will perform identically for two different BICM channels which happen to have the same pre-FEC BER.

To study the robustness of the pre-FEC BER as a metric to predict post-FEC BER, we show in Fig. 4 BER$_{pre}$ as a function of BER$_{pre}$ for the same 24 combinations of constellations and codes as in Fig. 3. Ideally, all lines corresponding to the same code rate should fall on top of each other, indicating that measuring BER$_{pre}$ is sufficient to predict the post-FEC BER when the BICM channel changes (in this case, due to the change in modulation format). The results in this figure show that the curves get “grouped” for the same code rate, and thus, BER$_{pre}$ is a better metric than SNR (cf. Fig. 3). The results in Fig. 4 also show that BER$_{pre}$ is a good metric for very high code rates. For low and moderate code rates, however, BER$_{pre}$ fails to predict the performance of the decoder. The implication of this is that measuring pre-FEC BER cannot be used to predict the post-FEC BER of an encoder–decoder pair across different BICM channels. The FEC-limit design paradigm fails.

In Fig. 5, we consider BER$_{pos}$ as a function of the (normalized) MI \(I(X;Y)/m\) for the 24 cases in Fig. 3 (same markers). Based on the analysis in Sec. II-B, we propose here to study BER$_{pos}$ as a function of the GMI. The information-theoretic rationale behind this idea is that a SD-FEC decoder is fed with this approach is that it relies on the strong assumption that the same SD-FEC encoder and decoder pair will perform identically for two different BICM channels which happen to have the same pre-FEC BER.

To study the robustness of the pre-FEC BER as a metric to predict post-FEC BER, we show in Fig. 4 BER$_{pre}$ as a function of BER$_{pre}$ for the same 24 combinations of constellations and codes as in Fig. 3. Ideally, all lines corresponding to the same code rate should fall on top of each other, indicating that measuring BER$_{pre}$ is sufficient to predict the post-FEC BER when the BICM channel changes (in this case, due to the change in modulation format). The results in this figure show that the curves get “grouped” for the same code rate, and thus, BER$_{pre}$ is a better metric than SNR (cf. Fig. 3). The results in Fig. 4 also show that BER$_{pre}$ is a good metric for very high code rates. For low and moderate code rates, however, BER$_{pre}$ fails to predict the performance of the decoder. The implication of this is that measuring pre-FEC BER cannot be used to predict the post-FEC BER of an encoder–decoder pair across different BICM channels. The FEC-limit design paradigm fails.

In Fig. 5, we consider BER$_{pos}$ as a function of the (normalized) MI \(I(X;Y)/m\) for the 24 cases in Fig. 3 (same markers). Based on the analysis in Sec. II-B, we propose here to study BER$_{pos}$ as a function of the GMI. The information-theoretic rationale behind this idea is that a SD-FEC decoder is fed with this approach is that it relies on the strong assumption that the same SD-FEC encoder and decoder pair will perform identically for two different BICM channels which happen to have the same pre-FEC BER.

To study the robustness of the pre-FEC BER as a metric to predict post-FEC BER, we show in Fig. 4 BER$_{pre}$ as a function of BER$_{pre}$ for the same 24 combinations of constellations and codes as in Fig. 3. Ideally, all lines corresponding to the same code rate should fall on top of each other, indicating that measuring BER$_{pre}$ is sufficient to predict the post-FEC BER when the BICM channel changes (in this case, due to the change in modulation format). The results in this figure show that the curves get “grouped” for the same code rate, and thus, BER$_{pre}$ is a better metric than SNR (cf. Fig. 3). The results in Fig. 4 also show that BER$_{pre}$ is a good metric for very high code rates. For low and moderate code rates, however, BER$_{pre}$ fails to predict the performance of the decoder. The implication of this is that measuring pre-FEC BER cannot be used to predict the post-FEC BER of an encoder–decoder pair across different BICM channels. The FEC-limit design paradigm fails.

In Fig. 5, we consider BER$_{pos}$ as a function of the (normalized) MI \(I(X;Y)/m\) for the 24 cases in Fig. 3 (same markers). Based on the analysis in Sec. II-B, we propose here to study BER$_{pos}$ as a function of the GMI. The information-theoretic rationale behind this idea is that a SD-FEC decoder is fed with this approach is that it relies on the strong assumption that the same SD-FEC encoder and decoder pair will perform identically for two different BICM channels which happen to have the same pre-FEC BER.

To study the robustness of the pre-FEC BER as a metric to predict post-FEC BER, we show in Fig. 4 BER$_{pre}$ as a function of BER$_{pre}$ for the same 24 combinations of constellations and codes as in Fig. 3. Ideally, all lines corresponding to the same code rate should fall on top of each other, indicating that measuring BER$_{pre}$ is sufficient to predict the post-FEC BER when the BICM channel changes (in this case, due to the change in modulation format). The results in this figure show that the curves get “grouped” for the same code rate, and thus, BER$_{pre}$ is a better metric than SNR (cf. Fig. 3). The results in Fig. 4 also show that BER$_{pre}$ is a good metric for very high code rates. For low and moderate code rates, however, BER$_{pre}$ fails to predict the performance of the decoder. The implication of this is that measuring pre-FEC BER cannot be used to predict the post-FEC BER of an encoder–decoder pair across different BICM channels. The FEC-limit design paradigm fails.

In Fig. 5, we consider BER$_{pos}$ as a function of the (normalized) MI \(I(X;Y)/m\) for the 24 cases in Fig. 3 (same markers). Based on the analysis in Sec. II-B, we propose here to study BER$_{pos}$ as a function of the GMI. The information-theoretic rationale behind this idea is that a SD-FEC decoder is fed with this approach is that it relies on the strong assumption that the same SD-FEC encoder and decoder pair will perform identically for two different BICM channels which happen to have the same pre-FEC BER.
with LLRs, and thus, the GMI is a better metric (see (16)). The values of BERpos as a function of the GMI are shown in Fig. 6. These results show that for any given code rate, changing the constellation does not greatly affect the post-FEC BER prediction if the GMI is kept constant. More importantly, and unlike for the pre-FEC BER, the prediction based on the GMI appears to work across all code rates.

The results in Fig. 6 suggest that measuring the GMI of the BICM channel is the correct quantity to characterize the post-FEC BER of a (capacity-achieving) SD-FEC decoder. Although for high code rates the results in Fig. 6 are somehow similar to those in Figs. 4 and 5, we have no theoretical justification the use BERpre or MI as a metric to predict the performance of a SD-FEC. More importantly, having a metric like the GMI that works for all code rates is very important. Considering only high code rates—as is usually done in the optical community—is an artificial constraint that reduces flexibility in the design, as correctly pointed out in [76, Sec. II-B].

IV. ACHIEVABLE RATES

In this section, we focus on cases where the number of bits per dimension is an integer, due to their practical relevance. The examples studied have 1, 2, and 3 bits/dimension, which corresponds to, respectively, 4, 8, and 12 bits/symbol or $M = 16, 256, 4096$ constellation points.

A. Achievable Rates for $M = 16$

We consider three 4D constellations with $M = 16$: PM-QPSK, $C_{4,16}$, and SO-PM-QPSK. While $C_{4,16}$ is asymptotically the best constellation in terms of BERpre, PM-QPSK and SO-PM-QPSK have the advantage of a lower implementation complexity. On the other hand, the results in [40], [41] show that $C_{4,16}$ gives higher MI than PM-QPSK at all SNRs. This indicates that $C_{4,16}$ is the best choice among these constellations for capacity-approaching CM transmitters with ML decoding.

In terms of binary labelings, we use the unique Gray code for PM-QPSK, which assigns a separate bit to each dimension. Thus, PM-QPSK becomes the Cartesian product of four binary shift keying (BPSK) constellations. $\sum_{k=1}^{4} I(B_k; Y) = I(X; Y)$, and thus, (18) holds with equality. In other words, PM-QPSK causes no penalty in terms of achievable rates if a BW decoder is used. For SO-PM-QPSK, we use the labeling proposed in [39], while for $C_{4,16}$ we use a labeling (found numerically) that gives high GMI for a wide range of SNR.

In Fig. 7, the MI and GMI for the three constellations under consideration are shown. For PM-QPSK, the GMI and the MI coincide. This is not the case for the two other constellations. The results in Fig. 7 show that $C_{4,16}$ gives a high MI at all SNRs; however, a large gap between the MI and GMI exists (more than 1 dB for low code rates). Therefore, $C_{4,16}$ will not work well with a BW decoder. The situation is similar for SO-PM-QPSK, although in this case the losses are smaller. Interestingly, when comparing the GMIs for $C_{4,16}$ and SO-PM-QPSK, we observe that they cross at around $\eta \approx 3.25$ bits/symbol. This indicates that a capacity-approaching transmitter with a BW decoder will perform better with $C_{4,16}$ than SO-PM-QPSK at high SNR. However, PM-QPSK is the best choice at any SNR.

To show that the conclusions above correspond to gains in terms of BERpos, we consider the LDPC codes defined in Sec. III and the additional code rate $R_c = 1/4$ (also defined in [73]). The obtained BER results for 4 different code rates are shown in Fig. 8. Among the three constellations, PM-QPSK always gives the lowest BERpos. The gains offered by PM-QPSK with respect to $C_{4,16}$ for low code rates are about 1 dB. More importantly, these gains are obtained by using a very simple demapper that computes four BPSK LLRs, one in each constellations for capacity-approaching CM transmitters with ML decoding.

12The MIs and GMIs were estimated using Monte Carlo integration.

13Calculated numerically via Monte Carlo integration.
dimension. These results also show that the GMI curves in Fig. 7 predict the coded performance of the system well. For example, the GMI curves indicate that at high code rates, $C_{4,16}$ is better than SO-PM-QPSK, which is exactly what happens in terms of BER$_{\text{pos}}$ (i.e., for $R_c = 9/10$, $C_{4,16}$ gives a lower BER$_{\text{pos}}$ than SO-PM-QPSK).

B. Achievable Rates for $M = 256$

For $M = 256$ (i.e., 2 bits/dimension) we consider two constellations. The first one is PM-16QAM, which is a straightforward generalization of PM-QPSK formed as the Cartesian product of four 4-ary pulse amplitude modulation (PAM) constellations. The labeling problem for PM-16QAM then boils down to labeling a 4-PAM constellation. Here we then consider the three nonequivalent binary labelings for 4-PAM: the BRGC [50], [51], the natural binary code (NBC) [55, Sec. II-B], and the anti-Gray code (AGC) [57, Sec. IV-E].

The second constellation we consider is a lattice-based constellation which we denote by $C_{4,256}$. It consists of all points with integer coordinates, such that the coordinate sum is odd and the Euclidean norm is 3 or less.\footnote{The same construction used with norm 1 gives PS-QPSK.} In lattice terminology, $C_{4,256}$ consists of the five first spherical shells of the $D_4$ lattice centered at a hole. The constellation was first characterized in [42, Table. IV] and [43, p. 822] and it corresponds to a point on the solid line in [33, Fig. 1 (a)] (4 bits/symbol/pol).

To label this constellation, we use a numerically optimized labeling obtained using the binary-switching algorithm (BSA) and the GMI approximation in [69]. The BSA was executed 300 times, and every time initialized with a randomly generated seed. A labeling was obtained, optimized for an SNR of $\gamma = 5$ dB (i.e., for MI around 3 bits/symbol). Binary labelings that give a slightly higher GMI can be obtained when optimizing at lower SNR; however, the gains are marginal.

The obtained results are shown in Fig. 9 and are quite similar to the ones in Fig. 7. When compared to PM-16QAM, the constellation $C_{4,256}$ gives higher MI but lower GMI. We thus conclude that $C_{4,256}$ is unsuitable for a BW decoder. A major advantage with PM-16QAM is the existence of a Gray code, which not only offers good performance but also lets the LLRs be calculated in each dimension separately, thus reducing complexity. The results in Fig. 9 also show a quite large gap between the MI for $C_{4,256}$ and PM-16QAM in the high-SNR regime. This is explained by the increase in minimum Euclidean distance of $C_{4,256}$ with respect to PM-16QAM [33, Fig. 1 (a)].

To show that the performance of a BW decoder based on LDPC codes follows the GMI prediction, we simulated 7 different code rates: 1/4 and the ones in (19) (all of them defined in [73]), PM-16QAM labeled by the BRGC, and $C_{4,16}$ using the numerically optimized binary labeling. For each of the 14 coding and modulation pairs, we measured the minimum value of $E_b/N_0$ needed to guarantee BER$_{\text{pos}} = 10^{-4}$. The obtained results are shown with circles in Fig. 9, where the vertical position of the marker is given by the achieved spectral efficiency (i.e., $\eta = R_c m$). The obtained results clearly show that the BW decoder based on LDPC codes follow the GMI curve quite well. The SNR penalty of this particular family of LDPC codes with respect to the GMI is between 1 and 0.5 dB for low and high code rates, respectively.

C. Achievable Rates for $M = 4096$

For $M = 4096$, we consider PM-64QAM labeled by the BRGC and by the NBC. This choice is motivated by the fact that the BRGC and the NBC are good labelings for the constituent 8-PAM constellation in terms of GMI for high and low SNR, respectively. We also consider $C_{4,4096}$, which is the
best known 4096-point constellation for uncoded transmission at high SNR [30], it is a subset of the $D_4$ lattice, found by extensive numerical search, and its binary labeling was also numerically optimized.\textsuperscript{15} The obtained results are shown in Fig. 10 and indicate that 4D optimized constellation offer gains in terms of MI, however, when the GMI is considered, it performs suboptimally.\textsuperscript{16} For example, at $\eta = 6$ bits/symbol, the losses caused by using $C_{4,4096}$ and a BW decoder with respect to PM-64QAM with the BRGC are about 4 dB. Similarly to Fig. 9, Fig. 10 also shows the achieved spectral efficiencies for a target BER$_{pos} = 10^{-4}$ and the same code rates used in Sec. IV-B. The results show that the penalties caused by using $C_{4,4096}$ with respect to PM-64QAM are much larger than the corresponding penalties in Fig. 9.

For $M = 4096$, the problem of selecting the binary labeling is very challenging. Although good labelings in the low- and high-SNR regimes can be found, these labelings are not necessarily suitable for the practically relevant medium-SNR regime. On the other hand, using 8-PAM in each dimension simplifies the search for labelings and results in penalties (with respect to the MI) tending to zero for medium and high SNR values.

To conclude, we selected the constellations and labelings that give the highest MI and GMIIs in Figs. 7, 9, and 10. The results are presented in Fig. 11 and show that the best constellation in terms of MI, regardless of the targeted spectral efficiency, is $C_{4,4096}$.\textsuperscript{17} The gap to the channel capacity for $\eta \leq 10$ bits/symbol is less than 1 dB, which makes us believe that changing the shape of a constellation with large cardinality is enough to make the MI to be close to the channel capacity.

\textsuperscript{15}The numerically optimized labelings obtained for $C_{4,16}$ and $C_{4,4096}$ have no regular structure and will be made available online as supplementary data to this paper.

\textsuperscript{16}Due to the large number of constellation points and dimensions, the MI and GMI for $C_{4,256}$ and $C_{4,4096}$ was estimated via Monte Carlo integration.

\textsuperscript{17}This is of course ignoring practical problems that would arise by using large constellations at low SNR.

When the GMI is considered, the results in Fig. 11 indicate that for $\eta \leq 3$ bits/symbol, PM-QPSK should be the preferred alternative, for $3 \leq \eta \leq 6$ bits/symbol, PM-16QAM labeled by the BRGC should be used, and for $\eta \geq 6$ bits/symbol, PM-64QAM with the BRGC should be used. For $3 \leq \eta \leq 6$ bits/symbol and PM-16QAM, the optimum FEC overheads should then vary between 33.3% and 166%, which is good agreement with the code rates considered in Sec. III (see (19)). The results in this figure also show that for $\eta \geq 3$ bits/symbol, the loss from using a BW decoder instead of an ML decoder is typically less than 1 dB.

V. CONCLUSIONS

In this paper, we studied achievable rates for coherent optical coded modulation transceivers where the receiver is based on a bit-wise structure. It was shown that the generalized mutual information is the correct metric to study the performance of capacity-approaching coded modulation transceivers based on this paradigm. We conjecture that the correct metric for a bit-wise receiver with iterative demapping is the mutual information.

For the suboptimal bit-wise structure under consideration, both analytical and numerical results show that simply transmitting and receiving independent data in each quadrature of each polarization is the best choice. Multidimensional constellations optimized for uncoded systems were shown to give high MI, and are thus good for ML decoders; these constellations, however, are not well-suited for bit-wise decoders. On top of the weaker performance and higher demapper complexity, such constellation also carry the design challenge of selecting a good binary labeling.

We did not try to increase the generalized mutual information by changing the shape of the constellation (geometrical shaping) or the probability of the transmitted symbols (probabilistic shaping). Constellation shaping and the effect of the nonlinear optical channel using the GMI as a figure of merit.

---

Fig. 10. MI and GMI for two constellations with $M = 4096$. Simulations results are shown with markers for seven code rates at BER$_{pos} = 10^{-4}$.

Fig. 11. MI and GMI for the best constellations with $M = 16$, 256, and 4096 from Figs. 7, 9, and 10, respectively.
are left for future work. The intriguing connection between the generalized mutual information and the pre-FEC BER (see Figs. 4 and 6) is also left for further investigation.

Acknowledgments

The authors would like to thank Dr. Domen h Lavery (University College London) and Tobias Fehenberger (Technische Universit t Munchen) for fruitful discussions regarding different parts of this manuscript.

References

[1] G. Ungerboeck, “Channel coding with multilevel/phase signals,” IEEE Trans. Inf. Theory, vol. 28, no. 1, pp. 55–67, Jan. 1982.
[2] H. Imai and S. Hirakawa, “A new multilevel coding method using error-correcting codes,” IEEE Trans. Inf. Theory, vol. IT-23, no. 3, pp. 371–377, May 1977.
[3] E. Zehavi, “8-PSK trellis codes for a Rayleigh channel,” IEEE Trans. Commun., vol. 40, no. 3, pp. 873–884, May 1992.
[4] G. Caire, G. Taricco, and E. Biglieri, “Bit-interleaved coded modulation,” IEEE Trans. Inf. Theory, vol. 44, no. 3, pp. 927–946, May 1998.
[5] A. Guillén i Fàbregas and J. M. Kahn, “Bit-interleaved coded modulation,” Foundations and Trends in Communications and Information Theory, vol. 5, no. 1–2, pp. 1–153, 2008.
[6] S. Benedetto, G. Olmo, and P. Poggiolini, “Trellis coded polarization shift keying modulation for digital optical communications,” IEEE Trans. Commun., vol. 43, no. 2/3/4, pp. 1591–1602, Feb./Mar./Apr. 1995.
[7] H. Bülow, G. Thielecke, and F. Buchali, “Optical trellis-coded modulation (oTCM),” in Proc. Optical Fiber Communication Conference (OFC), Los Angeles, CA, Mar. 2004.
[8] H. Zhao, E. Agrell, and M. Karlsson, “Trellis-coded modulation in PSK and DPSK communications,” in Proc. European Conference on Optical Communication (ECOC), Cannes, France, Sep. 2006.
[9] M. S. Kumar, H. Yoon, and N. Park, “Performance evaluation of trellis code modulated oDQPSK using the KLSE method,” IEEE Photon. Technol. Lett., vol. 19, no. 16, pp. 1245–1247, Aug. 2007.
[10] M. Magarini, R.-J. Essiambre, B. E. Basch, A. Ashikhmin, G. Kramer, and A. J. de Lind van Wijngaarden, “Concatenated coded modulation for optical communications systems,” IEEE Photon. Technol. Lett., vol. 22, no. 16, pp. 1244–1246, Aug. 2010.
[11] I. B. Djordjevic and F. Vasic, “Multilevel coding in M-ary DPSK/Differential QAM high-speed optical transmission with direct detection,” J. Lightw. Technol., vol. 24, no. 1, pp. 420–428, Jan. 2006.
[12] C. Gong and X. Wang, “Multilevel LDPC-Coded high-speed optical transmission with direct detection,” J. Lightw. Technol., vol. 25, no. 11, pp. 3619–3625, Nov. 2007.
[13] L. Beygi, E. Agrell, and M. Karlsson, “Novel multilevel coded modulation scheme for fiber optical channel with nonlinear phase noise,” in IEEE Global Telecommunications Conference (GLOBECOM), Miami, Dec. 2010.
[14] B. P. Smith and R. F. Kschischang, “A pragmatic coded modulation scheme for high-spectral-efficiency fiber-optic communications,” J. Lightw. Technol., vol. 30, no. 13, pp. 2047–2053, July 2012.
[15] R. Farhoudi and L. A. Rusch, “Multi-level coded modulation for 16-ary constellations in presence of phase noise,” J. Lightw. Technol., vol. 32, no. 6, pp. 1159–1167, Mar. 2014.
[16] L. Beygi, E. Agrell, J. M. Kahn, and M. Karlsson, “Coded modulation for fiber-optic networks: Toward better tradeoff between signal processing complexity and optical transparent reach,” IEEE Signal Processing Magazine, vol. 31, no. 2, pp. 93–103, Mar. 2014.
[17] I. B. Djordjevic, S. Sankaranarayanan, S. K. Chilappagari, and B. Vasic, “Low-density parity-check codes for 40-Gb/s optical transmission systems,” IEEE J. Quantum Electron., vol. 12, no. 4, pp. 555–562, July/Aug. 2006.
[18] H. Bülow and T. Rankl, “Soft coded modulation for sensitivity enhancement of coherent 100-Gbit/s transmission systems,” in Proc. Optical Fiber Communication Conference (OFC), San Diego, CA, Mar. 2009.
[19] D. S. Millar, T. Koike-Akino, R. Maher, D. Lavery, M. Paskov, K. Kojima, K. Parsons, B. C. Thomsen, S. J. Savory, and P. Bayvel, “Experimental demonstration of 24-dimensional extended Golay coded modulation with LDPC,” in Proc. Optical Fiber Communication Conference (OFC), San Francisco, CA, Mar. 2014.
[20] C. Häger, A. Graell i Amat, F. Brännström, A. Alvarado, and E. Agrell, “Improving soft FEC performance for higher-order modulations via optimized bit channel mappings,” Opt. Express, vol. 22, no. 12, pp. 14 544–14 558, June 2014.
[21] I. B. Djordjevic, M. Cvijetic, L. Xu, and T. Wang, “Using LDPC-coded modulation and coherent detection for ultra highspeed optical transmission,” J. Lightw. Technol., vol. 25, no. 11, pp. 3619–3625, Nov. 2007.
[22] H. Batson, I. B. Djordjevic, L. Xu, and T. Wang, “Multidimensional LDPC-Coded modulation for beyond 400 Gb/s per wavelength transmission,” IEEE Photon. Technol. Lett., vol. 21, no. 16, pp. 1139–1141, Aug. 2009.
[23] H. Bülow, X. Lu, L. Schmalen, A. Kleenkamp, and F. Buchali, “Experimental performance of 4D optimized constellation alternatives for PM-8QAM and PM-16QAM,” in Proc. Optical Fiber Communication Conference (OFC), San Francisco, CA, Mar. 2014.
[24] H. Bülow, U. Abay, A. Schenk, and J. B. Huber, “Coded modulation of polarization- and space-multiplexed signals,” in Asia Communications and Photonics Conference and Exhibition (ACP), Shanghai, China, Nov. 2011.
[25] H. Bülow and E. Masalkina, “Coded modulation in optical communications,” in Proc. Optical Fiber Communication Conference (OFC), Los Angeles, CA, Mar. 2014.
[26] L. Schmalen, “Energy efficient FEC for optical transmission systems,” in Proc. Optical Fiber Communication Conference (OFC), San Francisco, CA, Mar. 2014.
[27] M. Karlsson and E. Agrell, “Which is the most power-efficient modulation format in optical links?” Opt. Express, vol. 13, no. 17, pp. 10 814–10 819, Apr. 2009.
[28] M. K. Simon and J. G. Smith, “Hexagonal multiple phase-and-shift-keyed signal sets,” IEEE Trans. Commun., vol. COM-21, no. 10, pp. 1144–1155, Oct. 1973.
[29] G. J. Foschini, R. D. Gitlin, and S. B. Weinstein, “Optimization of two-dimensional signal constellations in the presence of Gaussian noise,” IEEE Trans. Commun., vol. 22, no. 1, pp. 28–38, Jan. 1974.
[30] E. Agrell, “Database of sphere packings,” Online: http://codes.cs.washington.edu/packings, 2014.
[31] E. Agrell and M. Karlsson, “Power-efficient modulation formats in coherent transmission systems,” J. Lightw. Technol., vol. 27, no. 22, pp. 5115–5126, Nov. 2009.
[32] S. Benedetto and E. Biglieri, Principles of Digital Transmission with Wireless Applications, Kluwer Academic, 1999.
[33] M. Karlsson and E. Agrell, “Spectrally efficient four-dimensional modulation,” in Proc. Optical Fiber Communication Conference (OFC), Los Angeles, CA, Mar. 2012.
[34] L. D. Coelho and N. Hanik, “Global optimization of fiber-optic communication systems using four-dimensional modulation formats,” in Proc. European Conference on Optical Communication (ECOC), Geneva, Switzerland, Sep. 2011.
[35] T. Pirsa, X. Liu, and S. Chandrasekhar, “Optimization of 16-ary quadra- ture amplitude modulation constellations for phase noise impaired channels,” in Proc. European Conference on Optical Communication (ECOC), Geneva, Switzerland, Sep. 2011.
[36] L. Beygi, E. Agrell, and M. Karlsson, “Optimization of 16-point ring constellations in the presence of nonlinear phase noise,” in Proc. Optical Fiber Communication Conference (OFC), Los Angeles, CA, Mar. 2011.
[37] H. Battiston, I. B. Djordjevic, L. Xu, and T. Wang, “Multidimensional LDPC-Coded modulation for beyond 400 Gb/s per wavelength transmission,” J. Lightw. Technol., vol. 25, no. 11, pp. 3619–3625, Nov. 2007.
[38] H. Bülow, U. Abay, A. Schenk, and J. B. Huber, “Coded modulation of polarization- and space-multiplexed signals,” in Asia Communications and Photonics Conference and Exhibition (ACP), Shanghai, China, Nov. 2011.
[39] H. Bülow and E. Masalkina, “Coded modulation in optical communications,” in Proc. Optical Fiber Communication Conference (OFC), Los Angeles, CA, Mar. 2014.
[40] L. Schmalen, “Energy efficient FEC for optical transmission systems,” in Proc. Optical Fiber Communication Conference (OFC), San Francisco, CA, Mar. 2014.
