

Unveiling Near-Capacity Code Design: The Realization of Shannon’s Communication Theory for MIMO Channels

Soon Xin Ng, Jin Wang and Lajos Hanzo

School of Electronics and Computer Science, University of Southampton, SO17 1BJ, United Kingdom.

Email: {sxn,lh}@ecs.soton.ac.uk.

Abstract—In this contribution we show how Shannon’s coding theory could be realized for Multiple-Input Multiple-Output (MIMO) channels with the aid of EXtrinsic Information Transfer (EXIT) charts and the Maximum-Aposteriori Probability (MAP) algorithm. We review the relationship between the channel capacity, EXIT charts and the MAP algorithm, outlining the principles of designing near MIMO-channel capacity coding schemes. Both serial and parallel concatenated coding schemes are designed based on these principles and near MIMO-channel capacity performance is achieved.

I. INTRODUCTION

The history of channel coding dates back to Shannon’s pioneering work [1] in 1948, where he showed that it is possible to design a communication system having an indefinitesimally low probability of error, whenever the rate of transmission is lower than the capacity of the channel. This motivated the search quest for codes that would produce arbitrarily small probability of error at a transmission rate close to the channel capacity. The classic coding design aims to approach the channel capacity by optimum channel codes which requires a huge number of trellis states or code memory, using a non-iterative decoder.

A breakthrough in the history of error control coding was the invention of turbo codes by Berrou *et al.* [2] in 1993. Convolutional codes were used as the component, which were combined with iterative decoders employing the Maximum-Aposteriori Probability (MAP) algorithm [3]–[5]. The results demonstrated that a performance close to the Shannon limit can be achieved in practice with the aid of binary codes using Binary Phase Shift Keying (BPSK) modulation. Turbo codes were later designed also for higher-order modulation. More specifically, bit-based Turbo Coded Modulation (TuCM) [6] and symbol-based Turbo Trellis Coded Modulation (TTCM) [7] were designed for attaining a higher spectral efficiency. The classic design of iterative decoders was based on analysing the associated distance spectrum, which influenced the Bit-Error Ratio (BER) floor of the code [8].

In order to analyze the convergence behaviour of an iterative decoding/detection scheme, density evolution techniques [9] and EXtrinsic Information Transfer (EXIT) charts [10], [11] were proposed. More explicitly, as suggested by the terminology, density evolution tracks the density distribution of the extrinsic probabilities, as the number of decoding iterations increases [12]. Similarly, EXIT charts track the mutual information exchange between the component decoders in consecutive iterations. It has been shown in [12], [13] that the area between the EXIT curves of the components in an iterative decoder is characteristic of the iterative decoder’s ability to approach the channel’s capacity. This is often referred to as the ‘area-property’ of the EXIT charts, which has been used by the binary EXIT chart [10] for rendering a near-capacity binary code design to a curve fitting problem.

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In this contribution, we present a unified portrayal of the channel capacity and of the symbol-based EXIT charts [11], based on the mutual information formula. We show that the area under the symbol-based EXIT charts is related to the channel capacity. The outline of the paper is as follows. The system model is described in Section II and the EXIT charts is detailed in Section III. Two near-capacity coding schemes are designed in Section IV and our conclusion is offered in Section V.

II. SYSTEM MODEL

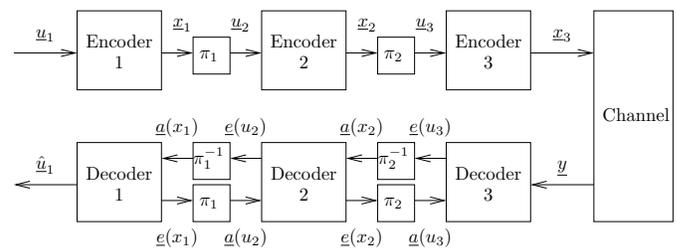


Fig. 1. The block diagram of a 3-stage serially concatenated coding scheme.

Fig. 1 shows a 3-stage serially concatenated coding scheme, where $\underline{u}_i = \{u_{i,1}, u_{i,2}, \dots\}$ and $\underline{x}_i = \{x_{i,1}, x_{i,2}, \dots\}$ denote the sequences of information symbols and coded symbols of ‘Encoder i ’, respectively. The sequence of channel-contaminated received symbols is denoted as \underline{y} and the estimate of \underline{u} is denoted as $\hat{\underline{u}}$. The interleaver and deinterleaver are represented by π and π^{-1} , respectively. The notations $\underline{a}(x_i) = \{a(x_{i,1}), a(x_{i,2}), \dots\}$ and $\underline{e}(x_i) = \{e(x_{i,1}), e(x_{i,2}), \dots\}$ represent the sequences of the *a priori* and *extrinsic* probability, respectively, for the coded symbols of ‘Decoder i ’. Similarly, $\underline{a}(u_i)$ and $\underline{e}(u_i)$ denote the corresponding sequences of *a priori* and *extrinsic* probability of the information symbols of ‘Decoder i ’.

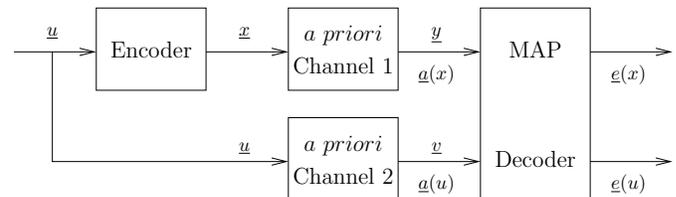


Fig. 2. The encoding and decoding of an intermediate encoder whose input symbols, \underline{u} , emanate from an outer encoder and whose output symbols, \underline{x} , are fed to an inner encoder. The *a priori* channels are used to model the outer and inner decoders.

Fig. 2 illustrates the decoding model of the intermediate decoder, i.e. the ‘Decoder 2’ of Fig. 1. We may use two *a priori* channels to model the outer and inner decoders following the approach of [12], Fig. 1]. More explicitly, the top *a priori* channel in Fig. 2 is used to model an inner decoder which has a transition probability of $P(y_k|x_k)$ signifying the probability of producing y_k at its output when the input is x_k . Similarly, the *a priori* channel at the bottom

of Fig. 2 is used to model an outer decoder, which has a transition probability of $P(v_k|u_k)$, representing the probability of producing v_k at its output, when the input is u_k .

Let us consider a Multiple-Input Multiple-Output (MIMO) system, which invokes N_t transmit antennas and N_r receive antennas. When complex-valued \mathcal{M} -ary PSK/QAM is employed, the received signal vector of the MIMO system can be written as:

$$\mathbf{y} = \mathbf{H}\mathbf{x} + \mathbf{n}, \quad (1)$$

where $\mathbf{y} = [y_1, \dots, y_{N_r}]^T$ is an N_r -element vector of the received signals, \mathbf{H} is an $(N_r \times N_t)$ -element channel matrix, $\mathbf{x} = [x_1, \dots, x_{N_t}]^T$ is an N_t -element vector of the transmitted signals and $\mathbf{n} = [n_1, \dots, n_{N_r}]^T$ is an N_r -element noise vector. Each elements of \mathbf{n} is an Additive White Gaussian Noise (AWGN) process having a zero mean and a variance of $N_0/2$ per dimension. There are $M = \mathcal{M}^{N_t}$ number of possible \mathcal{M} -ary PSK/QAM phasor combinations in the transmitted signal vector \mathbf{x} . Let us denote \mathbf{x} as an M -ary PSK/QAM signal vector.

The conditional probability of receiving a signal vector \mathbf{y} , given that an M -ary PSK/QAM signal vector $\mathbf{x}^{(m)}$, $m \in \{1, \dots, M\}$, was transmitted over Rayleigh fading channels is determined by the Probability Density Function (PDF) of the noise, yielding:

$$p(\mathbf{y}|\mathbf{x}^{(m)}) = \frac{1}{(\pi N_0)^{N_r}} \exp\left(-\frac{\|\mathbf{y} - \mathbf{H}\mathbf{x}^{(m)}\|^2}{N_0}\right), \quad (2)$$

where $\|(\cdot)\|$ is the Frobenius norm of vector (\cdot) .

The capacity of Discrete-input Continuous-output Memoryless Channel (DCMC) [14] for the MIMO system using M -ary signalling can be derived as [15], [16]:

$$C = \log_2(M) - \frac{1}{M} \sum_{m=1}^M \mathbb{E} \left[\log_2 \sum_{n=1}^M \exp(\Psi_{m,n}) \mid \mathbf{x}^{(m)} \right], \quad (3)$$

where the exponent $\Psi_{m,n}$ is given by [16]:

$$\Psi_{m,n} = \frac{-\left\| \mathbf{H}(\mathbf{x}^{(m)} - \mathbf{x}^{(n)}) + \mathbf{n} \right\|^2 + \|\mathbf{n}\|^2}{N_0}. \quad (4)$$

III. EXIT CHARTS

The EXIT charts [10], [11] visualize the input/output characteristics of the constituent MAP decoders in terms of the average mutual information transfer. Let us use the term ‘mutual information’ as the ‘per-symbol mutual information’, unless otherwise stated. In the context of the intermediate decoder of Fig. 2, the EXIT chart visualises the following mutual information exchange:

- 1) average mutual information of \underline{u} and $\underline{a}(u)$:

$$I_A(u) = \frac{1}{N_u} \sum_{k=1}^{N_u} I(u_k; a(u_k)); \quad (5)$$

- 2) average mutual information of \underline{x} and $\underline{a}(x)$:

$$I_A(x) = \frac{1}{N_x} \sum_{k=1}^{N_x} I(x_k; a(x_k)); \quad (6)$$

- 3) average mutual information of \underline{u} and $\underline{e}(u)$:

$$I_E(u) = \frac{1}{N_u} \sum_{k=1}^{N_u} I(u_k; e(u_k)); \quad (7)$$

- 4) average mutual information of \underline{x} and $\underline{e}(x)$:

$$I_E(x) = \frac{1}{N_x} \sum_{k=1}^{N_x} I(x_k; e(x_k)), \quad (8)$$

where the number of symbols in the sequences \underline{u} and \underline{x} are given by N_u and N_x , respectively. Since the intermediate decoder is associated with four mutual information transfers according to Eqs. (5)–(8), two three-dimensional EXIT charts [17], [18] are required for visualising the four-dimensional mutual information transfer between the intermediate decoder (two-input, two-output) and the outer decoder (one-input, two-output) as well as between the intermediate decoder and the inner decoder (two-input, one output) of Fig. 1.

Provided that a MAP decoder is used, the average *extrinsic* mutual information of \underline{u} may be computed as [11]:

$$\begin{aligned} I_E(u) &= \frac{1}{N_u} \sum_{k=1}^{N_u} H(u_k) - H(u_k|e(u_k)) \\ &= \log_2(\mathcal{M}_u) - \frac{1}{N_u} \sum_{k=1}^{N_u} \mathbb{E} \left[\sum_{m=1}^{\mathcal{M}_u} e(u_k^{(m)}) \log_2(e(u_k^{(m)})) \right] \end{aligned}$$

where $e(u_k^{(m)}) = P(u_k^{(m)}|\underline{y}, \underline{v}_{[k]})$ is the extrinsic probability of the hypothetically transmitted symbol $u_k^{(m)}$, for $m \in \{1, \dots, \mathcal{M}_u\}$, which is provided by the MAP decoder and the expectation may be removed, when N_u is sufficiently large, yielding:

$$I_E(u) = \log_2(\mathcal{M}_u) - \frac{1}{N_u} \sum_{k=1}^{N_u} \sum_{m=1}^{\mathcal{M}_u} e(u_k^{(m)}) \log_2(e(u_k^{(m)})). \quad (9)$$

Similarly, we have [11]:

$$I_E(x) = \log_2(\mathcal{M}_x) - \frac{1}{N_x} \sum_{k=1}^{N_x} \sum_{m=1}^{\mathcal{M}_x} e(x_k^{(m)}) \log_2(e(x_k^{(m)})), \quad (10)$$

where $e(x_k^{(m)}) = P(x_k^{(m)}|\underline{y}_{[k]}, \underline{v})$ is the extrinsic probability of the hypothetically transmitted symbol $x_k^{(m)}$, for $m \in \{1, \dots, \mathcal{M}_x\}$, generated by the MAP decoder and N_x is assumed to be sufficiently large.

The average *a priori* mutual information of \underline{u} and \underline{x} may be model using the following assumptions [11], [19]:

- 1) the LLRs of the bits are Gaussian distributed: the LLR of a bit b , which can be either from the sequence \underline{u} or \underline{x} , is given by [10]:

$$z = h_A b + n_A, \quad (11)$$

where the variance of the AWGN n_A is σ_A^2 per dimension and the equivalent ‘fading factor’ is given by $h_A = \sigma_A^2/2$ [10];

- 2) the bits in a symbol are assumed to be independent of each other and uniformly distributed: the average *a priori* mutual information of a symbol sequence \underline{u} (or \underline{x}), where each symbol u_k (or x_k) consists of L_u (or L_x) bits, is L_u (or L_x) times the average *a priori* mutual information of a bit in the symbol.

The average *a priori* mutual information of a certain bit denoted as $b \in \{b^{(1)} = +1, b^{(2)} = -1\}$ and its LLR z may be expressed similar to Eq. (3) as:

$$I(b; z) = 1 - \frac{1}{2} \sum_{m=1}^2 \mathbb{E} \left[\log_2 \sum_{n=1}^2 \exp(\Psi_{m,n}^A) \mid b^{(m)} \right], \quad (12)$$

where $\exp(\Psi_{m,n}^A) = p(z|b^{(n)})/p(z|b^{(m)})$ and the conditional Gaussian PDF is given by:

$$p(z|b) = \frac{1}{\sqrt{2\pi\sigma_A^2}} \exp\left(-\frac{(z - h_A b)^2}{2\sigma_A^2}\right), \quad (13)$$

while the exponent is given by:

$$\Psi_{m,n}^A = \frac{-\left|(\sigma_A^2/2)(b^{(m)} - b^{(n)}) + n_A\right|^2 + |n_A|^2}{2\sigma_A^2}. \quad (14)$$

Note that another interpretation of Eq. (12) was given in [10, Eq. (14)]. We have a function $I_A = I(b; z) = J(\sigma_A)$, with $J(\sigma_A)$ being monotonically increasing and therefore invertible. Hence, at a given I_A we may find the corresponding σ_A value from $J^{-1}(I_A)$. Finally one may compute the corresponding LLR value z from Eq. (11). The *a priori* mutual information of a L_u -bit symbol u_k is given by:

$$I(u_k; z_{(k)}) = \sum_{i=1}^{L_u} I(b_{(k,i)}^u; z_{(k,i)}^u), \quad (15)$$

where $z_{(k)}^u = \{z_{(k,1)}^u, \dots, z_{(k,L_u)}^u\}$ is the LLR sequence, which is related to the L_u bits of u_k and $z_{(k,l)}^u$ is the LLR of $b_{(k,l)}^u$, which is the l th bit in the k th symbol u_k .

IV. NEAR-CAPACITY CODE DESIGN

It is clear from Eqs. (3), (9), (10) and (12) that both the computation of the channel capacity and that of the EXIT chart is based on the the same mutual information formula. It was shown in [11], [12] that when a MAP decoder is used, the *extrinsic* probability $e(u_k)$, which is computed at the decoder's output, contains the same amount of information as the sequences \underline{y} and $\underline{v}_{[k]}$ at the decoder's input, i.e. we have:

$$I(u_k; e(u_k)) = I(u_k; \underline{y}, \underline{v}_{[k]}). \quad (16)$$

Similarly, the *extrinsic* probability $e(x_k)$ contains the same amount of information as the sequences $\underline{y}_{[k]}$ and \underline{v} :

$$I(x_k; e(x_k)) = I(x_k; \underline{y}_{[k]}, \underline{v}). \quad (17)$$

Hence, the area under the EXIT curve I_A is the same as the area under the EXIT curve I_E when the MAP algorithm is used in all decoders, although the shape of the curve may change. In other words, the MAP decoder acts like a lossless filter where its two outputs represent a full statistical characterisation of the two sequences observed at its input.

According to the properties of EXIT charts [12], the area under the EXIT curve of the inner code equals to the capacity of the communication channel (the upper *a priori* channel in Fig. 2), when the communication channel's input is given by equiprobable M -ary symbols and the *a priori* channel (the lower *a priori* channel in Fig. 2) is modeled using a Binary Erasure Channel (BEC) [12]. This area property was formally shown to be valid for arbitrary inner codes and communication channels, provided that the *a priori* channel is modeled by a BEC [12]. Furthermore, the area under the EXIT curve of the outer code equals to $(1-R_1)$, where R_1 is the outer code's rate.

Let us consider a PSK/QAM MIMO mapper as the inner encoder ('Encoder 3' of Fig. 1) and assume that 'Decoder 3' is a MIMO soft-demapper. The decoder can be modelled using Fig. 2 where the upper *a priori* channel is an uncorrelated Rayleigh fading MIMO channel and the lower *a priori* channel models 'Decoder 2' of Fig. 1. Since the MIMO channel is non-dispersive and uncorrelated in time, the *a posteriori* probability of the soft MIMO demodulator contains only the channel transition term $\gamma_k(\hat{s}, s)$ of the MAP algorithm:

$$o(u_k) = \gamma_k(\hat{s}, s) = a(x_k) \cdot a(u_k). \quad (18)$$

Hence, the associated symbol-based *extrinsic* probability is given by $e(u_k) = o(u_k)/a(u_k) = a(x_k)$, which is independent of the sequence \underline{v} . Hence, based on Eq. (16) we have:

$$I(u_k; e(u_k)) = I(u_k; y_k). \quad (19)$$

In this case, $I_E(u)$ is independent of $I_A(u)$, where $I_E(u)$ is a constant across $I_A(u) = \{0, \dots, I_{A,\max}\}$, with $I_{A,\max}$ being the

capacity of the lower *a priori* channel. Since 'Encoder 3' represents a one-to-one mapper, we have $I(u_k; y_k) = I(x_k; y_k)$. Hence, the area under the EXIT curve of this soft demodulator can be computed using Eqs. (8) and (19) as:

$$\mathcal{A} = \int_0^{I_{A,\max}} \frac{1}{N} \sum_{k=1}^N I(u_k; e(u_k)) dI_A \quad (20)$$

$$= \int_0^{I_{A,\max}} dI_A \frac{1}{N} \sum_{k=1}^N I(x_k; y_k) \quad (21)$$

$$= I_{A,\max} I(x; y) = I_{A,\max} C, \quad (22)$$

where C is the capacity of the MIMO channel (including the case when $N_t = N_r = 1$) given by Eq. (3). Hence, it is shown that the area under the symbol-based EXIT curve of a soft-demapper is given by the product of the communication channel's capacity C and the capacity of the *a priori* channel $I_{A,\max}$. This area property derived for the related symbol-based EXIT chart also indicates that when the intermediate code (or the outer code of a 2-stage scheme) is also symbol-based, all the mutual information gleaned from the communication channel has been transferred to the soft-demapper's extrinsic symbol probability sequence. In this case, the link from \underline{x}_2 of 'Encoder 2' to $\underline{a}(x_2)$ of 'Decoder 2' in Fig. 1 may be considered to be a transformed MIMO channel, namely $I_A(x_2) = I(x_2; a(x_2)) = C$. Hence, if we employ a symbol-based 'Encoder 2' and a MAP algorithm for all decoders, then the area under the EXIT curve of $e(u_2)$ at the output of 'Decoder 2' can be shown to be:

$$\mathcal{A} = C \cdot I_A(u_2). \quad (23)$$

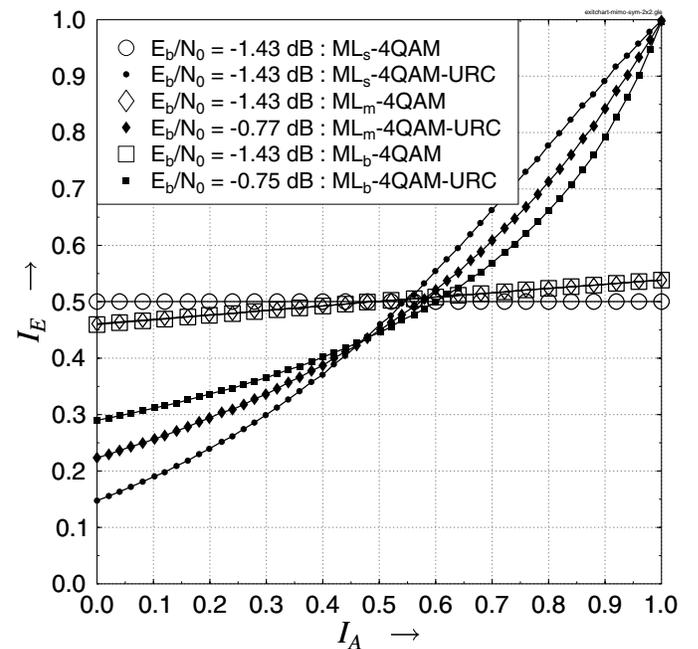


Fig. 3. Normalized EXIT charts where the area under each EXIT curve is 0.5. The different soft-demapper types that compute the extrinsic probability of 4-bit, 2-bit or 1-bit symbols are represented as ML_s , ML_m and ML_b , respectively.

Let us now consider a $N_t = N_r = 2$ MIMO scheme using $\mathcal{M} = 4$ -level QAM transmissions. The MIMO channel's capacity computed from Eq. (3) at a throughput of $\eta = 2$ bit/s/Hz is $E_b/N_0 = -1.43$ dB. The MIMO mapper is considered to be the inner encoder (Encoder 3 of Fig. 1), where $L = N_t \times \log_2(\mathcal{M}) = 4$ bits are mapped to $N_t = 2$ transmit antennas, each employing 4QAM transmissions.

Note that we may represent the input symbol of the MIMO mapper as N number of $L = 4$ -bit symbols, $2N$ number of $L/2 = 2$ -bit symbols or $4N$ number of $L/4 = 1$ -bit symbols. At the receiver, the soft-demapper is used to compute the extrinsic probability of these 4-bit, 2-bit or 1-bit symbols. The EXIT curves recorded for the various soft-demapper types are shown in Fig. 3 which were normalized to unity by dividing the I_A (or I_E) values by $I_{A,\max} = L = 4$ bits. The different soft-demapper types, which compute the extrinsic symbol probability as that of 4-bit, 2-bit or 1-bit symbols are represented as ML_s , ML_m and ML_b , respectively. The area under each of the EXIT curves in Fig. 3 equals 0.5 (or $\mathcal{A} = 0.5 \times L^2 = 2L$ before the normalization). It is shown in Fig. 3 that the EXIT curve for ML_s -4QAM is a horizontal line at $I_E = 0.5$, while that for ML_m -4QAM and ML_b -4QAM emerges from $I_E < 0.5$ and terminates at $I_E > 0.5$.

A. Serial Concatenated Scheme

We employed a recursive symbol-based unit-memory Unity-Rate Code (URC) [18] as the intermediate encoder, where each input/output symbol has $L = N_t \times \log_2(\mathcal{M}) = 4$ bits and each URC encoded symbol is mapped to $N_t = 2$ transmit antennas, where 4QAM transmission is used at each antenna. The symbol-based URC encoder employs a modulo- $M = 2^L$ adder and it requires M number of trellis states. At the receiver a soft-demapper was used to compute the $M = 2^L = 16$ -valued extrinsic symbol probability for each of the URC encoded symbol $x_{2,k}$ for $k \in \{1, \dots, N\}$, where $N = N_x = N_u$ is the number of symbols. These symbol probabilities are fed to the URC decoder in order to compute the extrinsic probability of the URC's input information symbols $u_{2,k}$ for $k \in \{1, \dots, N\}$. The EXIT curve of the ML_s -4QAM-URC scheme is shown in Fig. 3, which is based on the URC's input information symbol u_2 while assuming the absence of iterative information exchange with the soft-demapper. Observe in Fig. 3 that the ML_s -4QAM-URC scheme requires the same E_b/N_0 value as the ML_s -4QAM arrangement in order to maintain an area of 0.5 under it.

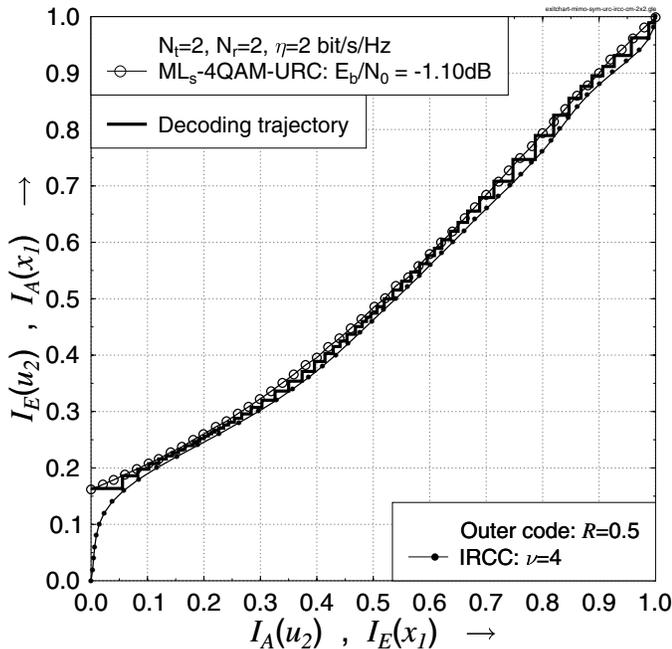


Fig. 4. Normalized EXIT charts for ML_s -4QAM-URC and IRCC.

The EXIT curves of the ML_m -4QAM-URC and ML_b -4QAM-URC schemes are also shown in Fig. 3, where the number of bits per

symbol entered into the URC encoder is given by $L/2 = 2$ and $L/4 = 1$, when employing the ML_m and ML_b scheme, respectively. When there is no iteration between the soft-demapper and the URC decoder, the EXIT curve of the URC decoder depends on the first I_E value of the soft-demapper. Hence, both the ML_m -4QAM-URC and ML_b -4QAM-URC schemes require a higher E_b/N_0 value in order to maintain an area of 0.5 under their EXIT curves, as shown in Fig. 3. In other words, this loss of information (or E_b/N_0 value) cannot be recovered if there is no iteration between the URC decoder and the soft-demapper. In this case, we only need a 2D EXIT chart in order to analyze the decoding convergence between the ML_s -4QAM-URC scheme and the outer code.

We employ the IRregular Convolutional Codes (IRCCs) of [20] as the outer encoder, where 17 subcodes were constructed from a memory-four mother code. We design the IRCC to match its EXIT curve to that of the ML_s -4QAM-URC scheme fed with the information symbols $u_{2,k}$ for $k \in \{1, \dots, N\}$. The resultant normalized EXIT curves are shown in Fig. 4, where an EXIT tunnel leading to the maximum $I_E(x_1)$ value was created for E_b/N_0 values in excess of -1.10 dB, which is as close as within 0.33 dB from the MIMO channel's capacity. The corresponding IRCC weighting coefficients are given by: $\alpha_o = [0.172469z^3 + 0.252871z^5 + 0.214169z^9 + 0.209495z^{13} + 0.15104z^{17}]$, where the superscript $s = \{3, 5, 9, 13, 17\}$ of z^s denotes the index of the subcodes used.

B. Parallel Concatenation Scheme

In contrast to the serial concatenated scheme of the previous section, here we assume that 'Encoder 1' and 'Encoder 2' are parallel concatenated as an outer code, which is the classic Turbo Code (TC) arrangement. The 'Encoder 3' is the same MIMO mapper that was used in Fig. 3. As argued earlier, it is impossible to recover the information loss, when a bit-based outer encoder (or when a bit-based interleaver) is used unless there are iterations between the outer decoder and the inner decoder (soft-demapper), when communicating over MIMO channels. However, a symbol-based TC that exhibits a horizontal EXIT curve, which matches the ML_s -4QAM EXIT curve would require non-identical and symbol-based component codes in the TC.

Let us consider a scheme where apart from having inner iterations in the outer TC decoder, there are also outer iterations between the TC decoder and the MIMO soft-demapper. We consider having identical memory $\nu = 3$ component codes for the outer TC, where the component code's generator polynomial is given by $G = [13 \ 6]_8$, which has a similar structure to the TCCM scheme of [7] except that a bit-interleaver is used before the MIMO mapper ('Encoder 3'). The normalized EXIT curves of the ML_b -4QAM and TC decoder (which uses 16 TC iterations) is shown in Fig. 5, where the notation $x_{1\&2}$ is used to denote the TC-encoded symbol. An open EXIT tunnel leading to the maximum value of $I_E(x_{1\&2}) = 1$ was created at $E_b/N_0 = -0.85$ dB, which is only 0.58 dB away from the MIMO channel's capacity.

The normalized EXIT curves of the *optimal* maximal-distance Non-Systematic Convolutional Codes (NSCCs) [21] having a code polynomial of $G = [15 \ 17]_8$ (code memory $\nu = 3$) and $G = [10533 \ 17661]_8$ (code memory $\nu = 12$) are also shown in Fig. 5. Note that approximately seven outer iterations are required between the TC decoder and the ML_b -4QAM for attaining $I_E \approx 1$. Each TC component code has $2^\nu = 8$ trellis states, hence after 16 TC iterations and 7 outer iterations, the ML_b -4QAM-TC scheme will have traversed through $2 \times 2^\nu \times 16 \times 7 = 1792$ number of trellis states per decoded bit. By contrast, the memory-12 maximal-distance NSCC decoder

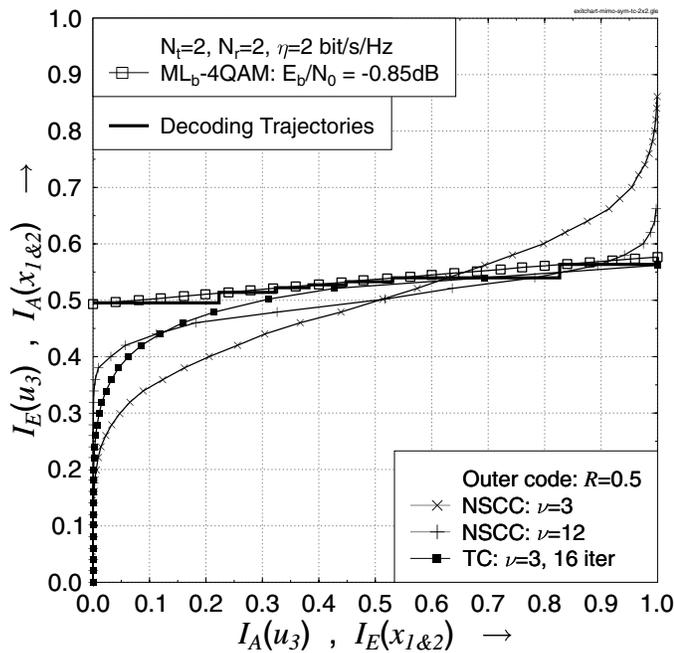


Fig. 5. Normalized EXIT charts for ML_b-4QAM, TC and NSC.

evaluated $2^{12} = 4096$ number of trellis states per decoded bit. Despite its higher complexity, there was no open EXIT tunnel leading to $I_E(x_{1\&2}) = 1$ between the EXIT curves of either of the two NSCC decoders and the ML_b-4QAM scheme at $E_b/N_0 = -0.85\text{dB}$. Hence, the memory-12 *optimal* NSCC requires about $4096/1792 \approx 2.3$ times higher complexity than the ML_b-4QAM-TC scheme, and yet fails to achieve decoding convergence to infinitesimally low BER at $E_b/N_0 = -0.85\text{dB}$.

V. CONCLUSIONS

We have shown the relationship between the computation of the channel capacity, EXIT charts and the MAP algorithm. As seen in Fig. 4, the classic *optimal* coding design (i.e. NSCC) was inefficient as it is very difficult to achieve a near-horizontal EXIT curve for matching the EXIT curve of the soft-demapper. It was shown in Fig. 3 that when communicating over MIMO channels, we need outer iterations exchanging extrinsic information with the soft-demapper unless a symbol-based decoder is employed, if approaching the MIMO channel's capacity is desired. It was shown that Shannon's communication theory could be realized for MIMO channels with the aid of the MAP algorithm as well as by using an EXIT charts aided code design and iterative decoding between two or three sub-optimal codes. Hence, achieving near-capacity performance is feasible, when the associated encoding/decoding complexity and decoding delay is affordable. The new challenge for code design is to approach the attainable rate at the lowest encoding/decoding complexity and the lowest possible interleaving delay.

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