Joint Space-Time Algebraically-Interleaved Turbo-Like Coded Incoherent MIMO Systems With Optimal and Suboptimal MAP Probability Decoders

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Abstract—The outstanding coding gains and also the practical complexities of the turbo-coded communication systems, heavily depend on the iterative soft-decision decoding algorithm used for the soft-input/output component decoders. The goal of this paper is to address the investigation on the energy gaps of the enhanced channel coding entity of [1], when original and also approximated logarithmic-based iterative MAP algorithms are employed for the iterative near maximum-likelihood (ML) decoding. As far as the large decoding delays of the turbo and turbo-like code structures constitute a major disadvantage for these types of channel-codes, the previously introduced high-performance channel-codes of [1] are developed for the original and also approximated iterative BCJR-MAP turbo decoders to enable the higher level of selection flexibility in error-rate versus computational complexity tradeoff. We examine the required energy per bit $E_b$ and the noise power spectral density $N_0$ ratio to attain the BER $10^{-4} - 10^{-5}$ for the differentially encoded phase-shift keying (D-PSK) MIMO Rayleigh fading channels.

Index Terms—BCJR algorithm, channel coding, maximum a posteriori (MAP), maximum-likelihood (ML) decoding, MIMO, space-time code, turbo code, wireless communications.

I. INTRODUCTION

The exceptional error correction capability of the turbo and turbo-like codes [2-4] made them part of many today’s wireless communication standards (see [2-6] of [5]) for suppressing the deleterious degradation of the system error-rate due to inherent multipath fading of wireless propagation environments. In addition in order to enhance the downlink performance without the requirements for multiple receive antennas at the terminals and also eliminate the necessity of channel state information (CSI) at the transmitting terminals, space-time codes are one of the key techniques which are robust against the certain non-ideal operating conditions of propagation environments, such as the frequency dispersion, antenna correlation, and channel estimation errors [6] hence the combined schemes i.e. encoding the turbo-coded sequences in space and time, and send them over the multi-antenna link, has attracted substantial research attention in academia and industry, in order to support the ever-growing capacity demands, and to provide top-quality systems and services for the modern digital wireless communications.

As one of the major concerns of communication theory, the received bit-error probability of the designed wireless system should be minimized by well-organized use of bandwidth and also power resources, while keeping the overall complexities feasible to reduce the buffering/decoder processing delays and consequential total equipment cost and deployment expenses. Specifically, for the case of turbo code family, it is well-known that greatest hindrance to their practical utilization lay in the complexities of the decoding algorithm. Although, far simpler than other codes of the similar performance, the structure still requires a high number of operations [7]. According to Table I, since incurred complexity for implementing the minimization of bit-error probability for BCJR-maximum a posteriori (MAP) decoding [8] (and for its variants) is significantly higher than the soft-output Viterbi algorithm (SOVA) [9], the vast body of research is dedicated for finding energy gaps or coding gain gaps (CGGs) between coded systems based on the utilization of the optimal and the suboptimal BCJR-MAP decoders; Table II tabulates the available results of the major contributions and the last findings on this subject. The MAP algorithm was firstly invented for estimating the a posteriori probabilities of the states, and the transitions of a Markov source observed in the memoryless noise. Berrou et al. invoked this algorithm for decoding the turbo codes [10].

<table>
<thead>
<tr>
<th>Operation</th>
<th>MAP</th>
<th>Log-MAP</th>
<th>Max-Log-MAP</th>
<th>SOVA</th>
</tr>
</thead>
<tbody>
<tr>
<td>Additions</td>
<td>$2^{n_{bits}}$ + 6</td>
<td>$6 \times 2^{n_{bits}}$ + 6</td>
<td>$4 \times 2^{n_{bits}}$ + 8</td>
<td>$2^{n_{bits}}$ + 9</td>
</tr>
<tr>
<td>Multiplications</td>
<td>$5 \times 2^{n_{bits}}$ + 8</td>
<td>$2^{n_{bits}}$</td>
<td>$2^{n_{bits}}$ + 1</td>
<td>$2^{n_{bits}}$</td>
</tr>
<tr>
<td>Maximizations</td>
<td>-</td>
<td>$4 \times 2^{n_{bits}}$ - 2</td>
<td>$4 \times 2^{n_{bits}}$ - 2</td>
<td>-</td>
</tr>
<tr>
<td>Lookups</td>
<td>-</td>
<td>$4 \times 2^{n_{bits}}$ - 2</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Exponentiations</td>
<td>$2^{n_{bits}+1}$</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
</tbody>
</table>

II. DESIGN OF CHANNEL-CODE ENTITIES

A. Encoder Structure and Transmission

The principle of the turbo codes based on the parallel concatenation suffers from the fact that at medium-to-high SNRs, an increase in transmit power does not lead to reduction...
of bit-errors. Therefore, the turbo coding community dedicated
a vast body of research for raising the minimum distance of
the code; hence, improving the flare performance. One way to
increase the minimum distance is by configuring the same
ingredients of turbo codes (i.e. component encoders/decoders
and interleaver/deinterleaver) in serial concatenation fashion as
proposed in [22]. The serial concatenation scheme has shown
to yield the high-performance comparable, and in some cases,
superior to the parallel concatenation schemes [3].

<table>
<thead>
<tr>
<th>System Specification</th>
<th>MAP (in dBs)</th>
<th>Log-Map (in dBs)</th>
<th>Max-Log-Map (in dBs)</th>
<th>BER</th>
</tr>
</thead>
<tbody>
<tr>
<td>[12] DVB-RCS application</td>
<td>N/A¹</td>
<td>1.4</td>
<td>1.65</td>
<td>10⁻³</td>
</tr>
<tr>
<td>[13] N = 1000-bit PCC² over AWGN channel with BPSK</td>
<td>1.825</td>
<td>1.81</td>
<td>2</td>
<td>10⁻⁵</td>
</tr>
<tr>
<td>[14]</td>
<td>1.75</td>
<td>1.75</td>
<td>1.93</td>
<td>10⁻⁴</td>
</tr>
<tr>
<td>[15] UMTS application</td>
<td>N/A</td>
<td>0.417</td>
<td>0.819</td>
<td>10⁻⁵</td>
</tr>
<tr>
<td>[16]</td>
<td>N/A</td>
<td>0.417</td>
<td>1.533</td>
<td>10⁻⁵</td>
</tr>
<tr>
<td>[17] 3GPP application</td>
<td>4.37</td>
<td>N/A</td>
<td>4.45</td>
<td>10⁻⁵</td>
</tr>
<tr>
<td>[18] OCDMA networking</td>
<td>Kₗ=90</td>
<td>N/A</td>
<td>Kₗ = 100</td>
<td>10⁻⁵</td>
</tr>
<tr>
<td>[19]</td>
<td>12.5</td>
<td>N/A</td>
<td>15</td>
<td>10⁻⁵</td>
</tr>
<tr>
<td>[20] Duo-binary Turbo Code used in Wi-MAX standard</td>
<td>N/A</td>
<td>0.9</td>
<td>1.25</td>
<td>10⁻⁵</td>
</tr>
<tr>
<td>[21] Rate ½ N=1024-bit PCC FF/FB² = (17, 15)net</td>
<td>1.9</td>
<td>N/A</td>
<td>2.1</td>
<td>10⁻⁴</td>
</tr>
</tbody>
</table>

¹N/A: not computed in this investigation.
²PCC: parallel concatenated codes.
³L: component code’s constraint length.
⁴N: interleaver size.
⁵Kₗ: average number of photons per symbol.
⁶R: code rate.
⁷FF/FB: feedforward/feedback.

The encoder of interest is shown in Fig. 1. It is composed of two cascaded digital IIR (infinite impulse response) filters. The 4-state outer convolutional component code with the rate Rₒ = 1/2, and the inner convolutional component code with the rate Rᵢ = 2/4, with generating matrices given by:

\[
C_{\text{out}}^{\text{int}} = \begin{bmatrix} 1 & 1 + D^2 \\ 1 + D^2 & 1 \end{bmatrix}
\]

(1)

\[
C_{\text{inner}}^{\text{(D)}} = \begin{bmatrix} 1 & 1 + D^2 \\ 1 + D^2 & 0 \end{bmatrix}
\]

(2)

The Cᵢ, Cᵦᵢ, Cᵧᵢ, and Cᵦᵢ denote the \((\frac{N_{\text{INT}}}{2}) - \text{bit source node} \) (i.e. a sequence \(m_t\) of binary message where \(0 \leq i \leq (\frac{N_{\text{INT}}}{2} - 1)\) and \(m_t \in \{0, 1\}\), \(N_{\text{INT}} - \text{bit outer coded sequence, 2N}_{\text{INT}}- \text{bit inner coded node} \) (i.e. coded bits \(x_t\), \(0 \leq t \leq (\frac{N_{\text{INT}}}{2R_{\text{block}}}) - 1\) ), and \(2N_{\text{INT}}- \text{bit differentially coded baseband sequence. Together with the constituent codes, it is mainly the permulator that determines the distance spectrum and the minimum distance of the overall code. That is one of the major issues of theoretical and practical turbo code design since it essentially determines the error-rate performance of the code. In particular, the performance of short-length turbo codes is highly determined by the permutation structure [23]. Permutation is a process of interleaving the order of a data sequence in a one-to-one deterministic format in order to create statistical independence between disturbances on parity symbols related to the same information symbol entering the channel encoders [24]. Many permutation algorithms have been investigated to improve the distance properties of the code while maintain the simple and practical hardware implementations. The most used interleaver type in communication systems is block interleaver [25], which has been extensively investigated and employed in turbo coded systems. For the spreading factor of permutation function \(\pi\), one way is to express in terms of the displacement vector \((\Delta a, \Delta b) \in Z^2\) of a pair of the input values \(i \leq \frac{i}{\bar{a}}\), a given permutation has spreading factor \(s_1, s_2\) if where \(\Delta a < s_1, \\text{then } |\Delta b| \geq s_2\) [26]. To study the randomness of the permutor, the set of displacement vectors of a block interleaver can be used [26]: \(D_i = \{((\Delta a, \Delta b) \in Z^2) | \Delta a = \omega - x, \Delta b = \pi(\omega - \pi(x)), 0 \leq x < \omega < N_{\text{INT}} + 1\}\); the largest set of the displacement vectors could be attained for Costas arrays [26-27] which are square arrays of dots/1s and blanks/0s such that: a) there exists exactly one dot/1 in each row and column; b) no four dots/1s form a parallelogram; and c) no tree dots/1s lying on a straight line are equidistant [28]; i.e. let \(n \in N^*\) and \(A\) be a permutation matrix, if \(a_{ij} = 1\), set \(f(j) = i, \text{ with } i,j \in \{1,2,...,n\}\); we call \(A\) a Costas array of dimension \(n\) iff [29]:

Figure 1. Bit-level turbo-like encoder with DPSK and space-time coding.
\[ \forall i_1, i_2, i_3, i_4 \in \{1, 2, \ldots, n\}, i_1 \leq i_2, i_3 \leq i_4; (i_1 - i_2, f(i_2) - f(i_1) - f(i_2)(i_3) - f(i_1)) \in \{i_2, i_3\}, i_2 = i_4. \] (3)

Costas arrays appeared for the first time in connection with describing a novel frequency hopping pattern for the Sonar systems and have been an active topic of Radar/Sonar research since [30-31]. They have also been applied in the other areas of the communications engineering; such as [32]. There is a website [32-33]. Costas arrays appeared for the first time in connection with describing a novel frequency hopping pattern for the Sonar systems and have been an active topic of Radar/Sonar research since [30-31]. They have also been applied in the other areas of the communications engineering; such as [32]. There is a website [32-33].

In order to investigate the fundamental Costas permutation mapping, let us consider a bijection \( f: [n] \rightarrow [n] \); \( f \) is a Costas array iff \([34-35])\; let \( n = \{1, \ldots, n\} \), \( n \in \mathbb{N}^{N_{\text{INT}}} \), \( \forall i, j, k \) such that \( 1 \leq i, j, k \leq n; f(i + k) - f(i) = f(j + k) - f(j) \Rightarrow i = j \) or \( k = 0 \). Based on the Welch's algebraic construction [28,36], and according to [37], let \( p \) be a prime and \( \phi \) be a primitive root of the finite field \( F(p) \) of \( p \) elements, and let \( d \in [p - 1] \) be a constant then the function \( f: [p - 1] \rightarrow [p - 1] \) where \( f(i) = \phi^{d \cdot i} \mod p \) is a Costas permutation of order \( p - 1 \); hence the output of permuter is the bit-level randomized outer-coded bit sequence, and is inserted into the inner code. The functionality of permuter is described by invertible permutation function \( \pi \) on the numbers \( \{0,1, \ldots, N_{\text{INT}}\} \):

\[ \pi: \{0,1, \ldots, N_{\text{INT}}\} \rightarrow \{0,1, \ldots, N_{\text{INT}}\} \] (4)

The fundamental permutation \([38]) is given by \((5)\) as:

\[ \pi = \begin{pmatrix} 0 & 1 & \cdots & N_{\text{INT}} - 1 \end{pmatrix} \] (5)

Its inverse permutation is denoted by \( \pi^{-1} \), which maps the permuted sequence back to the original order i.e. we have \( \pi^{-1}(\pi(i)) = i \) with:

\[ \pi = \begin{pmatrix} 0 & 1 & \cdots & N_{\text{INT}} - 1 \\ \pi^{-1}(0) & \pi^{-1}(1) & \cdots & \pi^{-1}(N_{\text{INT}} - 1) \end{pmatrix} \] (6)

Therefore, as an instance, the fundamental Costas interleave arrays for permutation size \( N_{\text{INT}} = 18 \) are given by:

\[ \pi^2 = \begin{pmatrix} 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 & 11 & 12 & 13 & 14 & 15 & 16 & 17 & 18 \\ \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow \end{pmatrix} \] (7)

\[ \pi^3 = \begin{pmatrix} 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 & 11 & 12 & 13 & 14 & 15 & 16 & 17 & 18 \\ \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow \end{pmatrix} \] (8)

\[ \pi^{13} = \begin{pmatrix} 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 & 11 & 12 & 13 & 14 & 15 & 16 & 17 & 18 \\ \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow \end{pmatrix} \] (9)

where, the index is the primitive element of GF(19). Before forming constellations from the \( C_{\text{int}} \), the differential coding (DC) is performed by taking the difference between the present information bit and the past encoded bit; by taking advantage of the high level of correlation between neighboring samples in the data sequence by the encoding differences; hence, at time \( k \), \( C_{\text{diff}, k} = C_{\text{int}, k} \oplus C_{\text{diff}, k-1} \) (\( \oplus \) stands for the modulo-2 addition).

In digital wireless communications, there are a great number of scenarios over fading channels in which the transmission has to be performed without having reliable carrier phase estimation at the reception side. Hence, phase ambiguity is likely to be present in the received signal due to imperfect knowledge of the carrier’s phase and fading effects. These phase errors can be pre-compensated by DC at the transmitter and incoherent detection at the receiver [39-40]. DC is computationally-simple technique widely used to enable non-coherent detection, when the carrier phase acquisition and the tracking are impossible/unreliable, or prohibitively complex due to successive phase noise, commonly happened in cases where noise and errors of local oscillators are considerable [41]. This is a great advantage when the waveform is passed through many circuits in a communication system and network changes (e.g. switching between data paths), result in occasional change in constructive sense of the output [42]. Moreover, that is well recognized for protecting bit sequence against unintentional polarity reversals [43]. Hence they are useful in many bad channel conditions e.g. for mobile users at cell boundary who experience bad quality-of-service (QoS), and for estimating channel gains significant amount of training data is required [44]. Although they result in deterioration of coding gain for \( \approx 2dB \) since whenever an error occurs on a single bit in the received data sequence that tends to propagate as double error, their robustness in the fading environments, and also their easier implementations made them widely used technique in the digital communication systems.

The signal constellations are formed from \( C_{\text{diff}} \) by using phase-shift keying (PSK) i.e. \( I_n = e^{j\theta_n} \). Since the modulation is differential, at each \( k \)th transmission interval, the data phasor \( \phi_k = \phi_{k-1} + \Delta\phi_k \) is sent, where \( \Delta\phi_k \in (0, 2\pi/Q, ..., 2\pi(Q - 1)/Q) \) is determined by the sequence of \( \log_2 Q \) input bits using the Gray and Binary mapping rule [45-46]. The transmission sequences for space-time encoding for twin-antenna and triplet-antenna systems are given in (32) and (37) of [47]. The Rayleigh-faded MIMO wireless channel with the PSK input is described by:

\[ y = Hx + n \]

where, \( H \in C^{N_t \times N_t} \), and

\[ x \in X = \left\{(x_1 x_2 \cdots x_N) \in X_N : 1 \leq m \leq N_t, 0 \leq q \leq Q - 1 \right\} \]

where, \( y, n \in C^{N_t \times 1} \), \( Q \) is the number of points in the PSK constellation, and \( N_t \) and \( N_t \) are the number of transmit and receive antennas, respectively (superscript \( T \) denotes transpose) [48]. The definition of \( X \) essentially forces the \( Q \)-ary PSK input constraint and also implies uniform power allocation at the transmitter; the values \( x \) can take guarantee that the average energy from each transmit antenna is \( 1/N_t \) and elements of \( H \) are i.i.d. complex Gaussian \( -N(0,1) \) [47]. Those of noise \( n \) are i.i.d. complex Gaussian \( -N(0, \sigma^2) \), where \( \sigma^2 = 1/\gamma_n \) [48] and \( \gamma_n \) is the symbol SNR \( E_s/N_0 \) [48].

**B. Reception and Channel Decoder**

At reception side, the decision criteria expressed in (17)-(20) of [49] are used for detecting symbols of the system with
two transmitting antennas; meanwhile the maximum-likelihood (ML) detector amounts to minimize the decision metrics given in appendix of [50] for detecting symbols of the system with three transmitting antennas. Then, the detected symbols are differentially decoded (i.e. \( \tilde{c}_k = c_k \otimes \tilde{c}_{k-1} \)), where \( c_k \) stands for the decisions taken on the transmitted information bits and \( \tilde{c}_k \) presents the received bits), and demodulated; and then the demodulated signals are fed into the iterative turbo decoder. The turbo decoder consists of the soft-input/soft-output constituent decoders for corresponding convolutional codes, linked by an interleaver and a deinterleaver (depicted in Fig. 2). Owing to the presence of permutor, optimum maximum-likelihood (ML) decoding algorithms for the turbo codes are very complex. However, their iterative decoding in which soft information is exchanged between sub-blocks of decoder is quite feasible and leads to near capacity performance. This information could be regarded as a type of diversity in that it can refine constituent decoder’s outputs in each iteration [19]. A single decoding is performed in each iteration using the channel observations and reliability information delivered by other component decoder that has acted before [19]. This iterative process can be applied to the BCJR-MAP decoding invoked by Berrou et al. for decoding turbo codes, and also to SOVA algorithm advocated in [51-53]. A fundamental difference between the BCJR-MAP and SOVA algorithms for decoding turbo coded sequences is that while the former minimizes the probability of bit-error, the latter minimizes the probability of sequence-error which yields considerable performance penalty but with the lower amount of the computational complexities. In SOVA, in order to find the most probable information sequence, only two paths in trellis diagram are considered but the competing path may not be the best competing path. As shown in Fig. 3, only one competing path at every decoding step is computed, in each decoding iteration; compared to the SOVA, where the sequence error probability is minimized, in the MAP, all the paths in the trellis diagram are considered at each decoding step; therefore, in order to approach the capacity performance, the computation complexity grows significantly. The soft-input and soft-output constituent decoders put out not only the various bits, but also the confidence it has in a specific decision which is described by the Log-Likelihood Ratio (LLR). The LLR is the ratio of the \( a \ posteriori \) probability (APP) of each information bit value. Suppose \( u \cdot y \cdot \pi (\omega) \cdot \), and \( 2(\cdot) \) are denoting the information sequence, noise-corrupted version of codeword corresponding to \( u \), permuted version of \( \omega \), and the map from the information bits to parity check bits. For codeword \( x = (u, 2(u), 2(\pi (u))) \) corresponding to the information sequence \( u \), the decoder must compute the APP for each data bit \( u_k \), which is given by \( L_k = f(u_k = 1|y)/f(u_k = 0|y) \). If \( \mathbb{L}_k \geq 1 \), the decoder estimates that \( u_k \) is equal to one, otherwise, it is then estimated to be equal to zero. Hence the turbo decoding process can be regarded as the iterative improvement of the APPs since the LLR is the ratio of the APPs of each information bit value. The APP is given by the equation (12) as [53]:

\[
L(u_k) = \log \frac{\sum_{\alpha_k \beta_k} r(y_k|u_k = l)}{\sum_{\alpha_k \beta_k} \alpha_k \beta_k}
\]

(12)

where, \( \alpha_k (\cdot) \), \( \beta_k (\cdot) \), and \( y_k (\cdot) \) are the forward, backward, and branch transition parameters, respectively; given by:

\[
\alpha_k (\tilde{S}_k - 1) = \frac{\sum_{\beta_k} r(y_k|u_k = l)}{\sum_{\beta_k} \beta_k}
\]

(13)

and,

\[
\beta_k (\tilde{S}_k) = \frac{\sum_{\alpha_k} r(y_k|u_k = 1)}{\sum_{\alpha_k} \alpha_k}
\]

(14)

and,

\[
y_k (C_k) = \text{Pr}(y_k^u|u_k = l) \cdot \text{Pr}(y_k^v|u_k = 1, \tilde{S}_k = \theta, \tilde{S}_{k-1} = \theta) \cdot \text{Pr}(u_k = l|\tilde{S}_k = \theta, \tilde{S}_{k-1} = \theta) \cdot \text{Pr}(\tilde{S}_k = \theta|\tilde{S}_{k-1} = \theta)
\]

(15)

where, \( \tilde{S}_k \) is the state of the encoder at time \( k \), \( \theta \) is \( S_k \) with \( S_k \) the set of transitions \( \tilde{S}_{k-1} = \theta \rightarrow \tilde{S}_k = \theta \) that are due to \( u_k = l \), and \( y_k^u \) and \( y_k^v \) are noise corrupted channel outputs for \( d_k \) and parity bits at time \( k \) respectively, and \( C_k \) denotes the collection \( (y_k^u, y_k^v, \tilde{S}_k, \tilde{S}_{k-1}, \tilde{S}_k) \). Hence, evaluation of APP could be summarized as the algorithm presented in Fig. 4. As far as the MAP algorithm is very complex, due to the amount of the computations, it could be then simplified by performing the operations in the logarithmic domain [54-56]. The forward, backward, and transition metrics could be calculated in the logarithmic domain using the Jacobian logarithm [14]. Using this approach, the multiplication in equation (13) and (14) is replaced by the addition in order to save the computations: \( \alpha_k (\cdot) = \max \{ \alpha_{k-1}(\cdot) + y_k^u (\cdot, \cdot) \} \) and \( \beta_k (\cdot) = \max \{ \beta_{k-1}(\cdot) + y_k^v (\cdot, \cdot) \} \); where, \( \alpha_k (\cdot), \beta_k (\cdot), \) and \( y_k (\cdot, \cdot) \) are logarithmic values of the forward, backward, and transition metrics, respectively.
In order to preserve the original MAP algorithm and yield equivalent performance, all maximizations can be augmented over the two values with the correction function. The Jacobian logarithm has been computed as: 
\[
\log(1 + e^{-|s_1 - s_2|}) = \max(x_1, x_2) + f_c(y - x)
\]
i.e. maximum of the function’s two arguments plus a nonlinear correction function, that is only a function of the absolute difference between the two arguments; i.e. \(f_c(y - x)\) depicted in Fig. 5. Thus, the extrinsic information bit in the equation (12) can be derived in terms of the new metrics [14] as:
\[
L(u_k) = \max_{(ξ_1, ξ_2, u=+1)}[\hat{δ}_{k-1}(ξ) + \hat{γ}_k(ξ, s) + \hat{β}_k(s)] - \max_{(ξ_1, ξ_2, u=-1)}[\hat{δ}_{k-1}(ξ) + \hat{γ}_k(ξ, s) + \hat{β}_k(s)]
\]

The significant computations saving also arise since division operation is replaced by the subtraction in logarithmic domain. This algorithm which tries to preserve the original MAP by calculating the bit-error probability with less the computation complexity is known as Log-MAP algorithm. If we avoid using the correction function term to further reduce the amount of the complexity, the resultant decoding algorithm is known as Max-Log-MAP algorithm [55], which is only based on the selection and comparison operations; the Jacobian logarithm is applied recursively; e.g. considering three values yields: 
\[
\max(x, y, z) = \max[\max(x, y), z]
\]
Typically, for more than two input values, the Jacobian logarithm is applied recursively; e.g. considering three values yields: \(\max(x, y, z)\). The Max-Log-MAP which consists of the two Viterbi algorithms (one running in the forward and another in the backward direction) is the least complex of the true MAP and Log-MAP, (note that it has twice the complexity of the SOVA for each half-decoding iteration) but offers the worst BER performance [16]. The Log-MAP avoids approximation in the Max-Log-MAP algorithm and presents the performance equivalent to the original MAP, while does not manifest the major disadvantage of the true MAP (i.e. high computational complexity). The correction only requires an additional one-dimensional dimensional lookup table (LUT), and also an addition per max operation [57]. The relationship among the MAP, Log-MAP, Max-Log-MAP, and SOVA, has been depicted in Fig. 6.
Algorithm 1. Simplified algorithm of the actual code used in the conducted computer simulations.

1: while (decision criterion based on outer decoder’s output) do
2:   If (targeted \( \Delta S \)) then;
3:     for (targeted 2;)
4:       set symbol alphabets for \( Q = ary \) PSK given in Fig. 1;
5:     select BCR – based algorithm
6:   if (MAP )
7:     (eq.12 – 15)
8:     else if (Log – MAP )
9:     \( L(u_k) = \max_{(\xi_{S},\xi_{M} \leq 1)} [\hat{\epsilon}_{k-1}(\xi) + \hat{\gamma}_{k}(\xi,s) + \hat{\beta}_{k}(s)] - \max_{(\xi_{S},\xi_{M} \leq 1)} [\hat{\epsilon}_{k-1}(\xi) + \hat{\gamma}_{k}(\xi,s) + \hat{\beta}_{k}(s)] \)
10: else if (Max – Log – MAP )
11:   selection and comparison \( \max (x_1 + \cdots + x_n) = \)
12: end for
13: end if
14: end while

<table>
<thead>
<tr>
<th>TABLE III. THE REQUIRED ( E_b/N_0 ) TO ACHIEVE BER ( = 10^{-4} ) FOR ( N = 600 ) (AND PRIMITIVE ELEMENT 373).</th>
</tr>
</thead>
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<th>TABLE IV. THE REQUIRED ( E_b/N_0 ) TO ACHIEVE BER ( = 10^{-5} ) FOR ( N = 600 ) (AND PRIMITIVE ELEMENT 373).</th>
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If \( \Delta S \) is the memory order of the system encoder; meanwhile, as our numerical results suggest, the deterioration of coding gain falls below \( 0.5 \text{ dB} \); hence, this slightly higher \( E_b/N_0 \) requirements for yielding the same error-rate translates into a considerable decrease in the complexity and sensitivity. Hence, the obtained results recommend that for the discussed channel-code entity, the iterative turbo decoders utilize Max-Log-MAP algorithm are good compromises between system performance and complexity.

In addition, for the case of \( Q = 4\) and 8-PSK constellation formations with the gray mapping where the labels of adjacent symbols differ only in one bit; i.e. \{00,01,11,10\} and \{000,001,011,101,110,111,100,101,111\} for \( Q = \{4,8\} \), respectively; additional coding gains of up to 1.5 and 3 \text{ dB} \) attained. Hence, the results confirm to use then gray mapping rather than the binary-coded bit assignment; i.e. \{00,01,10,11\} and \{000101101111101011111\} for \( Q = \{4,8\} \) in order to form signal constellations; which yields additional coding gains without incurring additional difficulties.

Moreover it is worth mentioning here that we defer to make comparisons for cases that the Monte Carlo simulation results may seem to be useless and/or the result of the investigations is predictable based on available literature for generic turbo coded system model with of 90-99% confidence level; the two major ones have been discussed briefly as follows:

1. **Comparison with block interleavers:** block interleaver is the most used interleaver type in communication systems [25]. In [1], the authors already suggest that serial turbo-like codes employing the Welch-constructed Costas interleavers of length 100-bit and 600-bit result in additional coding gains of up to 3 and 4 \text{ dB} \) for attaining the BER \( = 10^{-3} \) and \( 10^{-6} \), respectively; in case of 3\times4 MIMO wireless transmissions with the coherent BPSK digital modulation.

2. **Investigation for different interleaver sizes:** It may seem to be required to characterize the performance of generic serial turbo-like codes for AWGN and fading channels for varying interleaver lengths since in design of conventional serial turbo-like codes [3], Benedetto et al. introduced that large coding gains of up to 2 \text{ dB} \) could be achieved by increasing the interleaver size of 200-bit to 2000-bit (at the expense of further buffering delays which is equal to \( N_{\text{int}} \times f_b \) where \( f_b \) is the bit rate [59]). In response to this potential idea of investigation, careful attention must be paid into the interleaver lengths to be selected in wireless systems. Modern wireless applications tend to be short-frame (less than 1000-bit). As the original work on serial turbo-like codes and also many other works on this class of channel-codes suggest, the system performance predictions and resultant coding gains for the in-range interleavers lengths (i.e. \( 100 \leq N_{\text{int}} \leq 1000 \) ) remain the same; hence, the important interleaver length range is already investigated and available on literature since 100-bit and 600-bit codes of [1] are the good representative lengths; higher sizes (i.e. \( 1000 \leq N_{\text{int}} \leq 65536 \) ) are not of particular interest in modern digital communications applications due to the resultant long delays.

**IV. CONCLUSION AND PERSPECTIVES**

In an attempt to find the efficient and high-performance (in terms of the error-rate and overall complexity) algorithms to be employed in designing the permutor and scramblers of serial turbo-like codes (which are of particular interest due to better
performance in flare region rather than classic schemes), in [1], it has been shown that space-time turbo-like codes employing the Welch-constructed Costas interleavers with representative \( N_{\text{INT}} \leq 1000 \) lengths result in the coding gains of 1-2 dB for attaining BER \( < 10^{-3} \) for 2x2 Rayleigh-faded MIMO systems, and 2-5 dB for achieving BER \( < 10^{-5} \) for 3x4 Rayleigh-faded MIMO systems, when they are compared with the same system model based on the block interleavers of 10x10 and 24x25 dimensions. Hence, as in many other related key investigations (i.e. listed in Table II), a great deal of interest which motivated us in order to study the coding gains associated with exploiting optimal and suboptimal MAP decoding algorithms in space-time turbo-like codes of [1], to demonstrate that to what extent we can save energy while tolerating the higher complexities. The numerical results and comparisons have been all extended for \( Q \)-ary PSK with gray and binary bit assignments, over 2x2 and 3x3 MIMO wireless systems (enjoying the STBCs [47, 49, 50], in conjunction with designed channel codes). Meanwhile, during this comprehensive investigation, we further detailed the potential analysis concerned with design of the channel encoder/decoder of [1], by presenting fundamental permutation arrays for the Welch-constructed Costas interleavers.

The results of this article can be extended in a number of ways; the following outlined issues are of particular interest:

1. As we pointed out briefly, the delay of turbo processing is considered as the main bottleneck of their application and utilization in the modern hard and firm real-time wireless data transmissions [60]. Hence finding strategies for preserving their performance while keeping the complexities lower are still important open research issues concentrated by the information theory community; In [15] it is suggested to use the Maclaurin series expansion instead of the logarithmic term in the Jacobian logarithmic function in order to enable direct implementation of Jacobian logarithmic function by comparators and adders, as this is the case for Max-Log-MAP algorithm. The outstanding high-performance very close to Log-MAP decoding while also outperforming the Max-Log-MAP algorithm in the waterfall region of the BER performance curves are presented in [15], when the communication channel is AWGN. Developing this decoding algorithm for channel-codes presented in this article over SISO/MIMO fading channels is of particular theoretical and practical importance.

2. In [61], Wang et al. proposed two new matrix transform algorithms for Max-Log-MAP decoding of parallel turbo codes which considerably speed up the decoding operations and also lowers the computational system complexities which facilitates the implementation of the Max-Log-MAP algorithm in special-purpose VLSI hardware for the parallel processing. Hence, this algorithm could be extended in the case of Log-MAP and also Talakoub et al. decoder [15], and employ the resultant iterative turbo decoder for the space-time turbo-like codes presented in this investigation.

3. As in the case of [1], in this article, we have assumed transmitters are uninformed about the channel state information (CSI). As far as the CSI affects the capacity of MIMO systems significantly, developing the detailed performance analysis and evaluations for the case of discussed channel codes employed in conjunction with the MIMO technologies with the available CSI at the Tx(s) is an important open issue.

4. Analog circuits may need lower power and the area compared with digital implementations in high-speed iterative applications; hence, in [62] design and test results for a fully-integrated translinear tail-biting MAP decoder with a goal of verifying the feasibility of the CMOS translinear circuits for the MAP decoders implementation is developed. Developments for the specific turbo codes (as in our scheme) may result in the revolutionary changes from key practical design aspects, i.e. power consumption and occupied space.

5. In order to take advantage of the trade-offs between the BER, code rate, spectral efficiency, and decoder complexity; the introduced FECs have potential advantage of enjoying the Khandani et al.’s turbo coding method [63], which parses the message frame into n-bit symbols and permutes on a symbol-by-symbol basis.

6. In [64], the turbo-coded OFDM transmissions have been analyzed; developing the combination of SCQICs with clipping is of particular interest in order to reduce the peak-to-average power ratio (PAPR), and achieve the provisioning performance with the reasonable redundancy.

REFERENCES


