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Research Article

Log-Likelihood Ratio Calculation for Iterative Decoding on Rayleigh Fading Channels Using Padé Approximation

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Approximate calculation of channel log-likelihood ratio (LLR) for wireless channels using Padé approximation is presented. LLR is used as an input of iterative decoding for powerful error-correcting codes such as low-density parity-check (LDPC) codes or turbo codes. Due to the lack of knowledge of the channel state information of a wireless fading channel, such as uncorrelated flat Rayleigh fading channels, calculations of exact LLR for these channels are quite complicated for a practical implementation. The previous work, an LLR calculation using the Taylor approximation, quickly becomes inaccurate as the channel output leaves some derivative point. This becomes a big problem when higher order modulation scheme is employed. To overcome this problem, a new LLR approximation using Padé approximation, which expresses the original function by a rational form of two polynomials with the same total number of coefficients of the Taylor series and can accelerate the Taylor approximation, is devised. By applying the proposed approximation to the iterative decoding and the LDPC codes with some modulation schemes, we show the effectiveness of the proposed methods by simulation results and analysis based on the density evolution.

1. Introduction

In recent years, iterative decoding techniques based on message passing algorithm such as turbo decoding [1] or belief-propagation (BP) decoding [2–4] have been attracted by their significant performance which attain close to the Shannon limit. The BP decoding algorithm, a well-known iterative decoding algorithm for LDPC codes [2, 3], has been widely studied for the binary erasure channel or the additive white Gaussian noise (AWGN) channel [2–8]. The algorithms firstly derive channel log-likelihood ratios (LLR) where the messages in the decoder are initialized to these LLR values. To exhibit good performance with BP decoding, this channel LLR should be obtained with high accuracy, but it becomes complicated for some channel models such as wireless fading channel [9].

In this study, we focus on a calculation of channel LLR over the *uncorrelated flat Rayleigh fading channels* where the discrete-time component transmitted signal is input to a band-limited channel; that is, $y_t = r_t w_t + z_t$ [10]. Here y_t , r_t ,

 w_t , and z_t denote a channel output, a fading gain, a channel input, and an additive white Gaussian noise (AWGN) $z_t \sim \mathcal{N}(0,\sigma^2)$ with variance σ^2 at time t, respectively. Hereinafter we drop the subscript t. If r at each received bit position is known to the receiver, we call this case *known channel state information* (CSI). If r at each received bit position is unknown to the receiver, we call this case *unknown CSI*. For a known CSI case, channel LLR ℓ can be easily calculated using the channel outputs y, σ^2 , and r. However, for an unknown CSI case which is more practical than known CSI and is our main interest, a calculation of the channel LLR ℓ is rather complex due to an integration of r.

The studies of wireless fading channels with the LDPC codes or turbo codes were presented in [11–20] with several modulation schemes [9, 10, 21] such as binary modulation (binary phase shift keying (BPSK)) or nonbinary modulations. In [14] for BPSK, Hou et al. have studied designing irregular LDPC codes [6] using density evolution [7, 8] and have shown that these codes can approach the Shannon limit.

But they have used the following simple linear approximation $\hat{\ell}_{\rm ex}$ for a calculation of the channel LLR:

$$\widehat{\ell}_{\rm ex} = \frac{2}{\sigma^2} \cdot E(r) \cdot y,\tag{1}$$

where E(r) denotes the expectation of the channel gain r. Although the previous approximation is simple and is easy to implement, it is inaccurate that degradation in the decoding performance compared with the true LLR [19] can be seen. Yazdani and Ardakani [19, 20] have also proposed a linear LLR approximation whose performance is almost identical to the true LLR:

$$\widehat{\ell}_{\rm opt} = \widehat{\alpha} y, \tag{2}$$

where $\hat{\alpha}$ is obtained by maximizing

$$\widehat{\alpha} = \arg\max_{\alpha} \left(1 - \int_{-\infty}^{\infty} \log_2 \left(1 + \exp\left(-\ell \right) \right) f_{\alpha} \left(\ell \right) d\ell \right). \quad (3)$$

Here $f_{\alpha}(\ell)$ is given by

 $f_{\alpha}(\ell)$

$$= \frac{\sqrt{2}\Delta^{2}}{\sqrt{\pi}\alpha\sigma} \exp\left(-\left(\frac{\Delta\ell}{\alpha\sigma}\right)^{2}\right) \times \left[\exp\left(-\left(\frac{\Delta\ell}{\sqrt{2}\alpha\sigma^{2}}\right)^{2}\right) + \frac{\Delta\sqrt{\pi}\ell}{\sqrt{2}\alpha\sigma^{2}} \cdot \operatorname{erfc}\left(-\frac{\Delta\ell}{\sqrt{2}\alpha\sigma^{2}}\right)\right], \tag{4}$$

where $\Delta = \sqrt{\sigma^2/(2\sigma^2 + 1)}$ and erfc(·) denotes the complementary error function [9]. However, an optimization of $\widehat{\alpha}$ using (3) and (4) requires for each channel parameter σ , so that it needs large complexity to implement.

Recently Asvadi et al. [11] have applied Taylor approximation of order *n* to the true LLR function such that

$$\widehat{\ell}_{\mathrm{T}n} = \sum_{i=0}^{n} c_i y^i,\tag{5}$$

where c_i denotes the coefficient of Taylor series of order i. They have derived both linear and nonlinear approximations with small orders. For a linear approximation (Taylor series of order n = 1), it is given by

$$\widehat{\ell}_{\mathrm{T}_1} = c_1 \, y. \tag{6}$$

For a nonlinear approximation (Taylor series of order n = 3), it is given by

$$\widehat{\ell}_{\mathrm{T3}} = c_1 y + c_3 y^3. \tag{7}$$

From the previous approximations, one can obtain accurate LLR without optimizing complicated functions such as in [19, 20].

To move our attentions to nonbinary modulations, which is more practical case, the LLR calculations are performed bitwise [15, 21]. The previous work by Yazdani and Ardakani [20], which is an extension of [19], has devised the LLR

approximation method, but it becomes complex to evaluate LLR due to the increment of the number of parameters for the optimization. To fit the true LLR functions, the authors in [11] have modified the approximation functions of Taylor series of order 3. This modification is not easy to replicate and is required for each parameter of the channels. Moreover it is well known that the Taylor approximation quickly becomes inaccurate as the variable *y* leaves the derivative point.

To overcome these problems, we devise a new LLR approximation using Padé approximation [22] on the uncorrelated flat Rayleigh fading channels with unknown CSI for BPSK and 8-PAM. Padé approximation expresses the original function by rational form of two polynomials with the same total number of coefficients of the Taylor series, and it can accelerate the Taylor approximation. Generally Padé approximation is accurate not only at the derivative point but also at the wide range of intervals of variables. We show by simulation results and analysis based on the density evolution that our method can approximate LLR function with high accuracy and can yield almost the same decoding performance as the true LLR. The Padé approximation is a generalization of the Taylor approximation, and the proposed method exhibits slightly better performance than the method using Taylor approximation [11]. Moreover we design irregular LDPC codes based on our LLR approximation function.

This paper is organized as follows: Section 2 gives the channel model, LLR calculation method, and LDPC codes. In Section 3, we briefly review Taylor and Padé approximations, and then we present the proposed LLR calculation by Padé approximation. Numerical results are shown in Section 4, and Section 5 concludes the paper.

2. Preliminaries

2.1. Channel Model and LLR Calculation. We here consider the following discrete-time channel model:

$$y = rw + z, (8)$$

where $w \in \chi$ and $y, z \in \mathcal{R}$ represent the channel input, output, and noise, respectively, and χ , \mathcal{R} denote a set of transmitted symbols and that of real numbers, respectively. (As mentioned in Section I, we drop a time subscript for y, r, w, and z.) Moreover $r \ge 0$ is the channel gain with an uncorrelated flat Rayleigh distribution by its probability density function (pdf) $P(r) = 2re^{-r^2}$, and z is the white Gaussian noise with mean 0 and variance σ^2 .

Using the bit-interleaved coded modulation (BICM) scheme [10, 15, 21] for a transmission, an information bit sequence is mapped to the codeword (bit sequence) of length N by error-correcting codes, and then it is partitioned into N/q blocks denoted by $\mathbf{b}_k = \{b_k^{(1)}, b_k^{(2)}, \dots, b_k^{(q)}\}, k = 1, 2, \dots, N/q$, of length q. Hereinafter we drop the subscript k of both \mathbf{b}_k and $b_k^{(i)}$ to simplify discussions. This block \mathbf{b} is mapped to transmit a signal $w \in \chi$ in a Gray-labeled Q-ary signal constellation χ of size $Q = 2^q$. The signal constellation for BSPK and 8-PAM is depicted in Figure 1.

For the fading channel, two cases can be considered which depends on the knowledge of *r* at the receiver.

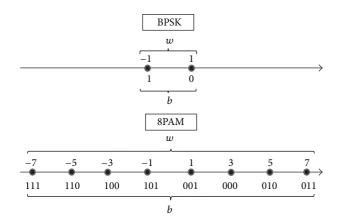


FIGURE 1: Signal constellations for BPSK and 8-PAM with Gray labeling.

2.1.1. Known CSI. For a known CSI case, we can use channel fading gain r for each bit position i = 1, 2, ..., q. The channel LLR is given by

$$\ell_{\text{CSI}}^{(i)} = \log \frac{P(y \mid b^{(i)}(w) = 0, r)}{P(y \mid b^{(i)}(w) = 1, r)}$$

$$= \log \frac{\sum_{w \in \chi_0^{(i)}} P(y \mid w, r)}{\sum_{w \in \chi_0^{(i)}} P(y \mid w, r)},$$
(9)

where $b^{(i)}(w)$ denotes the ith bit of block \mathbf{b} which is mapped to w and $\chi_j^{(i)}$ is a set of blocks \mathbf{b} which satisfies $b^{(i)}(w) = j$ for j = 0, 1. Moreover a base of logarithm takes a natural number e, and $P(y \mid w, r)$ is given by

$$P(y \mid w, r) = \frac{1}{\sqrt{2\pi\sigma^2}} \cdot \exp\left(-\frac{(y - rw)^2}{2\sigma^2}\right).$$
(10)

For BPSK, (9) is reduced to

$$\ell_{\text{CSI}}^{(1)} = \frac{2ry}{\sigma^2}.\tag{11}$$

The calculation of the previous equation is not a difficult task.

2.1.2. Unknown CSI. For an unknown CSI case, we cannot use channel fading gain r for each received bit position. The channel LLR is given by

$$\ell_{\text{NSI}}^{(i)} = \log \frac{P(y \mid b^{(i)}(w) = 0)}{P(y \mid b^{(i)}(w) = 1)} \\
= \log \frac{\sum_{x \in \chi_0^{(i)}} P(y \mid w)}{\sum_{w \in \chi_1^{(i)}} P(y \mid w)}, \tag{12}$$

where $P(y \mid w)$ is given by

$$P(y \mid w)$$

$$= \int_{0}^{\infty} P(y \mid w, r) P(r) dr$$

$$= \int_{0}^{\infty} \frac{1}{\sqrt{2\pi\sigma^{2}}} \cdot \exp\left(-\frac{(y - rw)^{2}}{2\sigma^{2}}\right) \times 2r \exp\left(-r^{2}\right) dr.$$
(13)

For BPSK, (12) is given by

$$\ell_{\text{NSI}}^{(1)} = \log \frac{\Psi\left(y/\sqrt{2\widetilde{\sigma}}\right)}{\Psi\left(-y/\sqrt{2\widetilde{\sigma}}\right)},$$
 (14)

where $\tilde{\sigma} = \sigma^2(1 + 2\sigma^2)$ and $\Psi(x) = 1 + \sqrt{\pi}x \exp(x^2) \operatorname{erfc}(-x)$. The calculation of (12) is so complicated that several works have tried to reduce the computational complexity by approximations [11, 13, 14, 19, 20].

For the Q-ary PAM, $P(y \mid w)$ in (13) becomes

$$P(y \mid w) = \frac{\exp(-y^2/\tilde{\sigma}_1^2)}{\sqrt{\pi}\tilde{\sigma}_1^3} \times \left(\operatorname{erfc}\left(-\frac{wy}{\sqrt{2}\sigma\tilde{\sigma}_1}\right) \times \sqrt{\pi}wy + \sqrt{2}\sigma\tilde{\sigma}_1 \exp\left(-\frac{w^2y^2}{2\sigma^2\tilde{\sigma}_1^2}\right)\right),$$
(15)

where $\tilde{\sigma}_1 = \sqrt{w^2 + 2\sigma^2}$. Using (15), the LLR in (12) can be evaluated. The previous equations are so complicated to implement that several works have tried to reduce the computational complexity by approximations.

Notice that for the fading channels with some modulations, the log-sum approximation was used for bitwise linear approximation. This approximation is only efficient for a known CSI case, since an integration of fading factor r for a calculation of LLR is not needed. But for an unknown CSI case, an integration of fading factor r is needed. Moreover it is effective only for a high signal-to-noise (SNR) region, where the sum in (9) is dominated by a single large term.

2.2. LDPC Codes. We here consider binary LDPC codes. An LDPC code is represented by the Tanner graph which consists of the variable nodes and the check nodes. These nodes are incident with the edges. Let d_v and d_c denote the maximum number of edges incident to the variable nodes and check nodes, respectively. Let $\lambda(x) = \sum_{i=2}^{d_v} \lambda_i x^{i-1}$ and $\rho(x) = \sum_{i=2}^{d_c} \rho_i x^{i-1}$ be variable node degree distribution and check node degree distribution where λ_i and ρ_i denote fractions of the number of edges incident to the variable node and check node of degrees i in the Tanner graph of the code, respectively. An LDPC code is specified by N, $\lambda(x)$, and $\rho(x)$. The rate of the codes is given by R = 1 - M/N where $M = N \int_0^1 \rho(x) dx / \int_0^1 \lambda(x) dx$.

An ensemble of LDPC codes [2] is denoted by $\mathscr{C}^N(\lambda(x))$, $\rho(x)$). Combined with the BP decoding algorithm, LDPC codes with optimized $(\lambda(x), \rho(x))$ by density evolution [5, 7, 8] can attain high performance which is close to the theoretical limit (Shannon limit). The iterative threshold σ^* of an ensemble of LDPC codes $\mathscr{C}^{\infty}(\lambda(x), \rho(x))$ is defined as the maximum standard deviation σ on the channel in (8) such that $\lim_{t\to\infty} P_e(t) = 0$ where $P_e(t)$ denotes the message error probability in iteration *t* of the BP decoding algorithm. $P_{e}(t)$ is calculated recursively by the density evolution [5, 7, 8] which keeps track the message error probability of the BP decoding algorithm from the pdf of channel LLR. Iterative threshold is sometimes measured by E_b/N_0 or signal-to-noise ratio (SNR) such that $(E_b/N_0)^* = 1/2R(\sigma^*)^2$ (for BPSK) or $21/2(\sigma^*)^2$ (for 8-PAM), respectively, where E_b and N_0 denote the average energy per information bit and one-sided power spectral density of the additive white Gaussian noise (AWGN).

3. LLR Approximation Based on Padé Approximation

Before approximating the true LLR function in (12), we briefly explain the Taylor approximation and then describe the Padé approximation.

3.1. Brief Review of Taylor and Padé Approximations. Let f(y) be the original function, and let $f^{(n)}(y)$ be nth derivative of f(y). Let $[s_1, s_2]$ and (s_1, s_2) be closed and open intervals between s_1 and $s_2, s_2 \ge s_1$, respectively.

Definition 1. Suppose that f(y) has n derivatives at point $y_0 \in [s_1, s_2]$ and $f^{(n+1)}(\xi)$ has a derivative at (s_1, s_2) where $\xi = s_1 + \theta(s_2 - s_1)$, $0 < \theta < 1$. Assume that $f^{(1)}(y_0)$, $f^{(2)}(y_0), \ldots, f^{(n)}(y_0)$ are required to be continuous on $[s_1, s_2]$ and $f^{(n+1)}(\xi)$ is required to exist on (s_1, s_2) . The Taylor polynomial of order n for f(y) at point y_0 is then defined by

$$P_{n}(y) = f(y_{0}) + f^{(1)}(y_{0})(y - y_{0}) + \cdots$$

$$+ \frac{f^{(n)}(y_{0})}{n!}(y - y_{0})^{n}$$

$$= \sum_{k=0}^{n} \frac{f^{(k)}(y_{0})}{k!}(y - y_{0})^{k}.$$
(16)

The remainder term, which is a difference between true value of the function and its Taylor series of polynomial, is given by

$$\frac{f^{(n+1)}(\xi)}{(n+1)!} (y - y_0)^{n+1}.$$
 (17)

For some n and y_0 , one can approximate the original function by $P_n(y)$ in (16). However, this function quickly becomes inaccurate as y leaves y_0 , even though n is large.

The approximation in (16) can often be accelerated by rearranging it into a ratio of two series using Padé approximation. It generalized the Taylor approximation with the same

total number of coefficients of two series. Before describing the Padé approximation, we rewrite (16) as follows:

$$P_{n}(y) = \sum_{k=0}^{n} \frac{f^{(k)}(y_{0})}{k!} (y - y_{0})^{k}$$

$$= \sum_{k=0}^{n} c_{k} y^{k}.$$
(18)

Definition 2. Suppose that f(y) is approximated by the Taylor series in (18). The Padé approximation of order (m', n'), n = m' + n', m', $n' \ge 0$, is given by

$$P_{m'}^{n'}(y) = \frac{\sum_{k=0}^{n'} a_k y^k}{\sum_{k=0}^{m'} b_k y^k},$$
(19)

where $a_0, a_1, \ldots, a_{n'}$ and $b_0, b_1, \ldots, b_{m'}$ are determined, so that the coefficients of the terms y^0, y^1, \ldots, y^n of $P_{m'}^{n'}(y)$ in (19) are equal to those of $P_n(y)$ in (18).

From Definition 2, the polynomials $\sum_{k=0}^{n'} a_k y^k$, $\sum_{k=0}^{m'} b_k y^k$, and $\sum_{k=0}^{n} c_k y^k$ satisfy the equation

$$\sum_{k=0}^{n'} a_k y^k - \left(\sum_{k=0}^{m'} b_k y^k\right) \cdot \left(\sum_{k=0}^{n} c_k y^k\right) = O\left(y^{n+1}\right). \tag{20}$$

Equation (20) tells that all the coefficients of $y^0, y^1, ..., y^n$ of $\sum_{k=0}^{n'} a_k y^k$ and $(\sum_{k=0}^{m'} b_k y^k)$ $(\sum_{k=0}^{n} c_k y^k)$ in left-hand side of (20) are equal. We can express these relations by the following simultaneous equation for i = 0, 1, ..., n:

$$\sum_{i=\max(i-m',0)}^{i} c_j b_{i-j} = \begin{cases} a_i, & i = 0, 1, \dots, n', \\ 0, & i = n' + 1, n' + 2, \dots, n. \end{cases}$$
(21)

Notice that the terms $y^{n+1}, y^{n+2}, \ldots, y^{n+m'}$ in left-hand side of (20) are included in $O(y^{n+1})$ in right-hand side of the equation. These terms are not necessary for the evaluation of $a_0, a_1, \ldots, a_{n'}$ and $b_0, b_1, \ldots, b_{m'}$.

We have already evaluated the coefficients c_0, c_1, \ldots, c_n in (18) by Taylor series. Moreover we assume that $a_0, a_1, \ldots, a_{n'}$ and $b_0, b_1, \ldots, b_{m'}$ are normalized, so we set $b_0 = 1$. This normalization is valid since the Padé approximation in (19) is of a rational form. Substituting c_0, c_1, \ldots, c_n and $b_0 = 1$ into (20), we can obtain coefficients $a_0, a_1, \ldots, a_{n'}$ and $b_0, b_1, \ldots, b_{m'}$. Note that the Taylor approximation and the Padé approximation are equivalent to each other if m' = 0. Therefore we can see that the Padé approximation is a generalization of the Taylor series approximation.

3.2. Applying Padé Approximation to LLR Function

3.2.1. LLR Calculation for BPSK. By applying the Padé approximation to the true LLR function for BPSK in (14), we can obtain the approximated function. To fit the true LLR function, we have searched the approximated function for several pairs of (m', n') and found that Padé approximation of

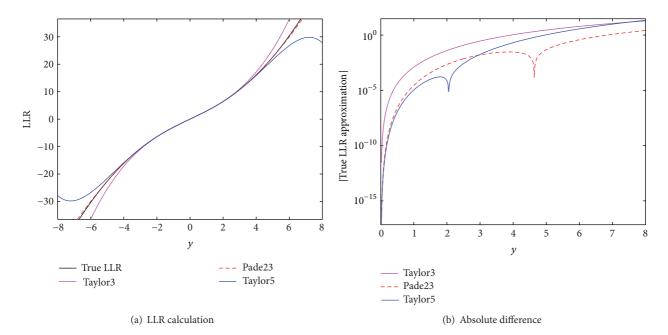


FIGURE 2: Comparison of true LLR and approximated LLRs (Taylor3, Taylor5, and Pade23) for the uncorrelated flat Rayleigh fading channel with unknown CSI and $\sigma = 0.6449$. From (a), Pade23 gives nearly the same values as the true function for various channel output y, and Taylor5 shows inaccurate as |y| becomes large. From (b), accuracy of Pade23 and Taylor5 is good especially for small |y|. But accuracy of them is different for large |y|; that is, Pade23 shows the best accuracy in three methods, but Taylor5 shows the worst accuracy.

order (m', n') = (2, 3) is more accurate than Taylor series of order n = 3 around 0 in (7) which is previously known as the best approximated one [11]. The proposed LLR approximation is given as follows:

$$\widehat{\ell}_{P23} = \frac{a_0 + a_1 y + a_2 y^2 + a_3 y^3}{b_0 + b_1 y + b_2 y^2},$$
(22)

where $a_0 = 0$, $a_2 = 0$, $b_0 = 1$, and $b_1 = 0$,

$$a_{1} = \sqrt{\frac{2\pi}{\tilde{\sigma}}}, \qquad b_{2} = \frac{-35 + 30\pi - 6\pi^{2}}{20(-3 + \pi)\tilde{\sigma}},$$

$$a_{3} = -\sqrt{\frac{\pi}{2}} \cdot \frac{15 - 30\pi + 8\pi^{2}}{30(-3 + \pi)\tilde{\sigma}\sqrt{\tilde{\sigma}}}.$$
(23)

The previous approximation is obtained by the Taylor series of order n = 5:

$$\widehat{\ell}_{T5} = c_1 y + c_3 y^3 + c_5 y^5, \tag{24}$$

where

$$c_{1} = \sqrt{\frac{2\pi}{\tilde{\sigma}}}, \qquad c_{3} = \frac{\sqrt{\pi}(-3+\pi)}{3\tilde{\sigma}\sqrt{2\tilde{\sigma}}},$$

$$c_{5} = \frac{\sqrt{2\pi}\left(35 - 30\pi + 6\pi^{2}\right)}{120\tilde{\sigma}^{2}\sqrt{\tilde{\sigma}}}.$$
(25)

Then (21) becomes

$$c_0b_0 = a_0,$$

$$c_0b_1 + c_1b_0 = a_1,$$

$$c_0b_2 + c_1b_1 + c_2b_0 = a_2,$$

$$c_1b_2 + c_2b_1 + c_3b_0 = a_3,$$

$$c_2b_2 + c_3b_1 + c_4b_0 = 0,$$

$$c_3b_2 + c_4b_1 + c_5b_0 = 0.$$
(26)

Substituting $c_0, c_1, ..., c_5$ in (24) and $b_0 = 1$ into (26), we get a_0, a_1, a_2, a_3 , and b_1, b_2 .

We then compare the accuracy of the approximated LLR functions. Figures 2(a) and 3(a) show LLR values for the uncorrelated flat Rayleigh fading channel with unknown CSI for $\sigma = 0.6449$ and 1.0264, respectively. LLR values in these figures are evaluated by the true LLR in (14), Taylor series of orders n = 3 and n = 5 ("Taylor3" and "Taylor5") in (7) and (24), respectively, the Padé approximation of order (m', n') = (2, 3) ("Pade23") in (22), and linear approximation ("Ex") in (1). From these figures, Padé approximation is almost identical to the true LLR for various channel output y. This may be contributory to the fact that the order n = 1m' + n' = 5) of Pade23 is larger than that of Taylor3 (n = 3). However, we can see that Taylor5 (n = 5) is inaccurate as |y| becomes large. Therefore large n is not an answer for an accurate LLR approximation, especially for using the Taylor approximation.

To compare the LLR approximations in detail, Figures 2(b) and 3(b) show the absolute differences between true LLR

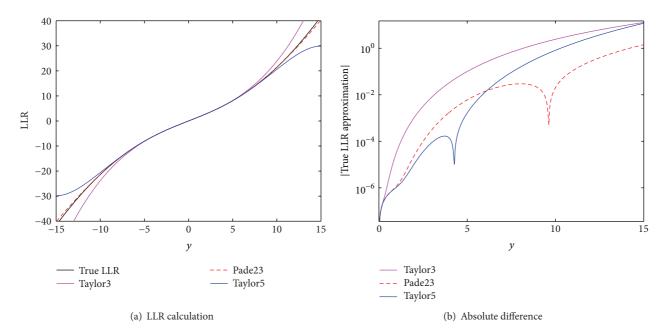


FIGURE 3: Comparison of true LLR and approximated LLRs (Taylor3, Taylor5, and Pade23) for the uncorrelated flat Rayleigh fading channel with unknown CSI and $\sigma = 1.0264$. Pade23 also gives almost the same values as the true function.

and approximated LLRs (Taylor3, Taylor5, and Pade23). We only show the case of $y \ge 0$, since the true LLR function is odd symmetric; that is, f(y) = -f(-y). From Figures 2 and 3, accuracy of Pade23 and Taylor5 are good especially for small |y|. But accuracy of them are quite different for large |y|, that is, Pade23 shows the best accuracy, but Taylor5 shows the worst accuracy.

Next we consider analysis based on the density evolution for different LLR calculation methods. We derive the pdf of LLR assuming that w = +1 is transmitted. From (10), we have

$$P(y \mid w = +1, r) = \frac{1}{\sqrt{2\pi\sigma^2}} \cdot \exp\left(-\frac{(y-r)^2}{2\sigma^2}\right). \tag{27}$$

Averaging (27) over r by an integration, we obtain

$$P(y \mid w = +1)$$

$$= \frac{\sqrt{2}}{\sqrt{\pi} (1 + 2\sigma^{2})} \cdot \exp\left(-\frac{y^{2}}{1 + 2\sigma^{2}}\right)$$

$$\times \left(\sigma \cdot \exp\left(-\frac{y^{2}}{2\widetilde{\sigma}}\right) + \frac{y}{2} \sqrt{\frac{2\pi}{1 + 2\sigma^{2}}} \cdot \operatorname{erfc}\left(-\frac{y}{\sqrt{2\widetilde{\sigma}}}\right)\right).$$
(28)

Then the density of the LLR function can be expressed in a parametric form [4]. For $\hat{\ell}_{P23}$ in (22), this is given by

$$\left(\widehat{\ell}_{P23}, \frac{P(y \mid w = +1)}{\widehat{\ell}'_{P23}}\right), \tag{29}$$

where $\hat{\ell}'_{P23}$ denotes a derivative of the function $\hat{\ell}_{P23}$ with y. Equation (29) is a case of the proposed approximation function, but one can obtain the pdf of the other LLR

functions. For example, replacing $\hat{\ell}_{P23}$ with $\ell_{NSI}^{(1)}$ in (14), we can obtain the pdf of the true LLR function $\ell_{NSI}^{(1)}$.

3.2.2. LLR Calculation for 8-PAM. We demonstrate Padé approximation for bitwise LLR of 8-PAM constellation with Gray labeling on the Rayleigh fading channel without CSI (SNR = 7.91 (dB) (σ = 1.699)) in Figure 4. The coefficients of LLR approximation functions of Taylor3 and Padé approximation are listed in Table 1. The derivative points y_0 for each bit LLR are chosen where these functions take $\ell_{\rm NSI}^{(i)}$ = 0 for i = 1, 2, 3. Notice that we can omit the coefficients for the case y_0 < 0 (bits 2 and 3), since these LLR functions are even functions; that is, we can derive from f(y) = f(-y) for y < 0. The orders of Padé approximation for each bit are different since each bit LLR function is distinct.

For bit 1, this point is $y_0 = 0$. For bit 2, these points are $y_0 = \pm 3.3449$ (2 points). So we have two LLR approximation functions for bit 2. We chose the order pairs of Padé approximation (4, 7) and (2, 4) (denoted by "Pade47" and "Pade24") for i = 1 and 2, respectively. The true LLR function for bit 2 is an even function, so we switch two LLR approximation functions on y = 0. Since these approximation functions intersect on y = 0, the resulting function becomes continuous.

For bit 3, these points are $y_0 = \pm 1.8848$, ± 6.9832 (4 points). For $y_0 = \pm 1.8848$ and $y_0 = \pm 6.9832$, we use Padé approximation of orders (4, 1) and (3, 4) (denoted by "Pade41" and "Pade34"), respectively. The true LLR function for bit 3 is also an even function, so we switch two LLR approximation functions, whose derivative points are $y_0 = \pm 1.8848$ (bit 3 (a) in Table 1), on the intersection point y = 0. But two LLR approximation functions (bit 3 (a) and (b) in Table 1) do not intersect on any y. Therefore we searched two

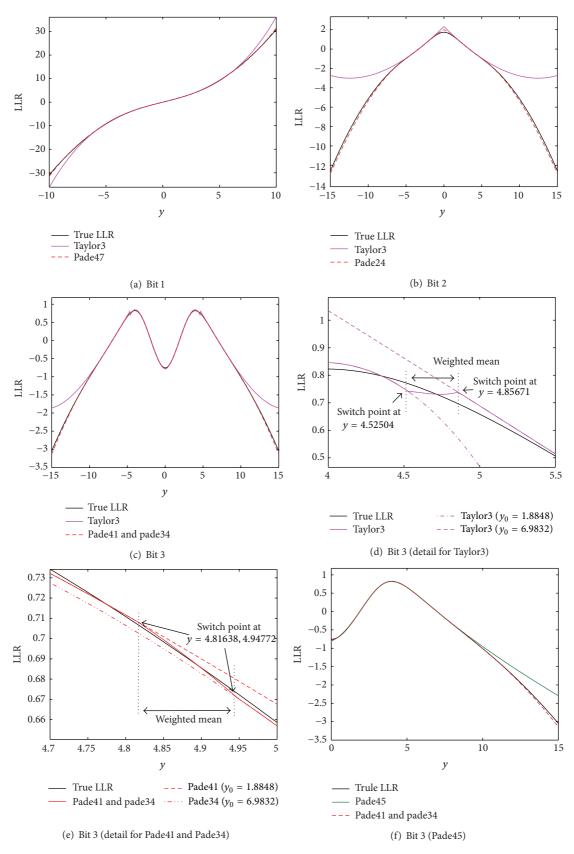


FIGURE 4: Comparison of bitwise True LLR and approximated LLRs for the 8-PAM with Gray labeling (SNR = 7.91 (σ = 1.699)).

Table 1: The coefficients of LLR approximation functions for 8-PAM at SNR = 7.91 (σ = 1.699).

(a) Taylor series of order n = 3

	\mathcal{Y}_0	c_0	c_1	c_2	c_3
bit 1	0.0	0.0	1.2135	0.0	0.0241
bit 2	3.3449	2.3273	-0.7706	0.0205	0.0006
bit 3 (a)	1.8848	-0.7855	0.0544	0.2848	-0.0491
bit 3 (b)	6.9832	2.2528	-0.2497	-0.0181	0.0011

(b) Padé approximation of numerator

	y_0	a_0	a_1	a_2	a_3	a_4	a_5	a_6	a_7
bit 1	0.0	0.0	1.2135	0.0	0.0334	0.0	0.0025	0.0	2.61E - 05
bit 2	3.3449	3.3382	-1.5496	0.1836	0.0039	-0.0028	_	_	_
bit 3 (a)	1.8848	-1.1396	0.6046	_	_	_	_	_	_
bit 3 (b)	6.9832	5.6135	-3.1331	0.5963	-0.0481	0.0017	_	_	_

(c) Padé approximation of denominator

	y_0	b_0	b_1	b_2	b_3	b_4
bit 1	0.0000	1.0000	0.0000	0.0076	0.0000	0.0016
bit 2	3.3449	1.6585	-0.3831	0.0557	_	_
bit 3 (a)	1.8848	1.4533	-0.6204	0.3220	-0.0852	0.0113
bit 3 (b)	6.9832	1.3191	-0.5502	0.0496	-0.0026	_

switch points between the interval of $y \in [1.8848, 6.9832]$ to minimize the loss of accuracy, and we set these points for Taylor approximation on 4.52504, 4.85671. We take the function (bit 3 (a)) for y < 4.52504 and take the function (bit 3 (b)) for y > 4.85671. Between $y \in [4.52504, 4.85671]$, we take weighted mean of two approximation functions (bit 3 (a) and (b)) which is shown in Figure 4(d). Likewise, we switch LLR functions for Padé approximation on 4.81638, 4.94772 as shown in Figure 4(e). From Figure 4(c), we can easily see switch points for Taylor3, but for Padé approximation we cannot see these points explicitly. This is because accuracy of Padé approximation is higher than that of Taylor3 for wide range of variables.

Notice that for bit 3, it is not necessary to consider weighted mean of two approximation functions if we use Padé approximation of higher order pairs. In this case, we found that Padé approximation of orders (4,5) at point $y_0 = \pm \arg\max_{y \ge 0} \ell_{\text{NSI}}^{(3)}$ provides almost the same accuracy as the combination of (3,4) and (4,1) which is shown in Figure 4(f) (denoted by "Pade45"). But the accuracy of Padé approximation of (4,5) for large |y| is not so good compared with the combination of Padé approximations (3,4) and (4,1). Also note that the previous work in [11] have modified Taylor3 functions to fit the true LLR for i=2 and 3. But this is not easy to replicate, so we only show the original form of Taylor3.

4. Numerical Results and Discussion

In order to compare the proposed LLR approximations, we use LDPC codes with BP decoding and show results by simulations and analysis based on the density evolution. We only show the case where CSI is unknown to the receiver.

TABLE 2: Comparison between the iterative thresholds for LDPC code ensembles using different LLR calculation methods.

(a) $\mathscr{C}^{\infty}(x^2, x^5)$ LDPC code ensemble

(a) 6 (x,x) LDFC code ensemble					
	True LLR	Taylor3	Pade23		
σ^{\star}	0.644755	0.644754	0.644755		
$(E_b/N_0)^*$ [dB]	3.810759	3.810772	3.810759		
(b) $\mathscr{C}^{\infty}(x^3, x^{15})$ LDPC code ensemble					
	True LLR	Taylor3	Pade23		
σ^{\star}	0.36770	0.36766	0.36770		
$(E_b/N_0)^*$ [dB]	6.929215	6.930160	6.929215		
(c) $\mathscr{C}^{\infty}(x^2, x^3)$ LDPC code ensemble					
	True LLR	Taylor3	Pade23		

	True LLR	Taylor3	Pade23
σ^{\star}	1.0263757	1.0263752	1.0263757
$(E_b/N_0)^*$ [dB]	2.784088	2.784092	2.784088

4.1. Results for BPSK. By using the density evolution [5, 7, 8], Table 2 shows iterative thresholds of $\mathscr{C}^{\infty}(x^2, x^5)$, $\mathscr{C}^{\infty}(x^3, x^{15})$, and $\mathscr{C}^{\infty}(x^2, x^3)$ LDPC code ensembles whose rates R are, respectively, R=0.5, 0.75, and 0.25 for different LLR calculation methods. Evaluating preciously, the thresholds of three methods are almost identical especially for true LLR and Pade23, but there exists a little gap between Taylor3 and the other methods.

Table 3 shows degree distribution profiles and their iterative thresholds for (a) threshold optimized and (b) rate optimized irregular LDPC codes. These profiles and their corresponding thresholds are evaluated based on the true LLR, Taylor3, and Pade23, respectively. From Table 3(a) for

Table 3: Degree distribution profiles and their iterative thresholds for the uncorrelated flat Rayleigh fading channel with unknown CSI by the density evolution. These profiles are (a) threshold optimized and (b) rate optimized codes. The Shannon limit for rate half code is $\sigma = 0.7436$ ($E_b/N_0 = 2.572$ dB).

(a) Threshold opting	nized
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	True LLR	Taylor3	Pade23		
λ_2	0.200284000	0.200283000	0.200284000		
λ_3	0.228588000	0.228591000	0.228588000		
λ_7	0.067795000	0.067777920	0.067794487		
λ_8	0.210231000	0.210246680	0.210231613		
λ_{30}	0.292602000	0.292601400	0.292601850		
ρ_9	1.000000000	1.000000000	1.000000000		
Rate	0.500000	0.500000	0.500000		
σ^{\star}	0.7232423	0.7232385	0.7232422		
$(E_b/N_0)^*$ [dB]	2.7068537	2.7069429	2.7068549		
(b) Rate optimized					
	True LLR	Taylor3	Pade23		
λ_2	0.202860	0.202758	0.202860		
λ_3	0.214366	0.214369	0.214366		
λ_7	0.157257	0.157244	0.157258		
λ_8	0.112911	0.112922	0.112859		
λ_{30}	0.312606	0.312607	0.312607		
ρ_9	1.000000	1.000000	1.000000		
Rate	0.4946847	0.4946836	0.4946845		
σ^{\star}	0.740167	0.740167	0.740167		
$(E_b/N_0)^*$ [dB]	2.613406	2.613406	2.613406		

fixed rate, the threshold of Pade23 is almost the same as that of true LLR and is slightly better than that of Taylor3. From Table 3(b) for fixed threshold, the rate of the code by Pade23 is almost the same as that by true LLR and is slightly higher than that by Taylor3. From these thresholds, they are close to the Shannon limit; that is, for rate half code, it is $\sigma = 0.7436$ ($Eb/N_0 = 2.572$ [dB]).

4.2. Results for 8-PAM. Figure 5 shows bit error rate (BER) of $(\lambda(x), \rho(x)) = (x^2, x^3)$ LDPC codes with N=15000 for the uncorrelated flat Rayleigh fading channel with 8-PAM. The number of transmitted codewords is 5×10^5 . BERs of true LLR and two Padé approximations (combination of orders (4, 1) and (3, 4) and orders (4, 5)) are almost the same, and that of Taylor3 is slightly higher than those of three methods.

Table 4 shows iterative thresholds of $\mathcal{C}^{\infty}(x^2, x^3)$ LDPC code ensemble for different LLR calculation methods with 8-PAM. The threshold of Padé approximation is almost identical to that of true LLR (inferior to 0.02 dB), and it is better than that of Taylor3. Notice that threshold of Taylor3 in [11] was SNR = 7.86 (dB) and it is better than that of Padé approximation. But the method in [11] performed several modifications to fit the true functions for the LLR functions of bits 2 and 3, it is not easy to replicate the approximations. Padé approximation is better than linear approximation in [20] (SNR = 7.88 (dB)). The threshold of Padé approximation

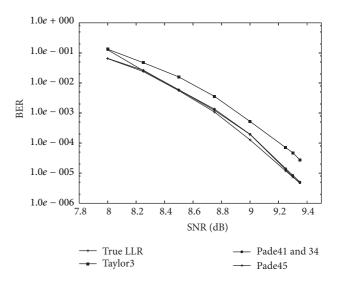


FIGURE 5: Comparison of BER of (x^2, x^3) LDPC codes with N = 15000 for the uncorrelated flat Rayleigh fading channel with 8-PAM.

Table 4: Comparison between the iterative threshold of $\mathscr{C}^{\infty}(x^2,x^3)$ LDPC code ensembles using different LLR calculation methods on 8-PAM.

	True LLR	Taylor3	Pade41 & Pade34
SNR*	7.85	8.05	7.87

of orders (4, 5) for bit 3 LLR function is SNR = 7.88 (dB) which is the same as the linear approximation.

5. Conclusion

In this paper, we apply the Padé approximation, which is a generalization of the Taylor approximation, to the LLR function on the uncorrelated flat Rayleigh fading channels with unknown CSI. Using Padé approximation, we can accelerate the accuracy of the LLR function that it is accurate not only around the derivative point but also at the other intervals of the variable. From the simulation results and analysis based on the density evolution, our method can yield almost the same decoding performance as the true LLR function and is slightly better than the conventional approximation method (Taylor approximation of order n=3). Moreover we derive some irregular LDPC code profiles whose iterative thresholds are close to the Shannon limit.

To apply Padé approximation, other modulations (e.g., 16-QAM) or other channels (Rician fading) are remained for further works.

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