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User-Antenna Selection for Physical-Layer Network Coding based on Euclidean Distance

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Abstract—In this paper, we present the error performance analysis of a multiple-input multiple-output (MIMO) physical-layer network coding (PNC) system with two different user-antenna selection (AS) schemes in asymmetric channel conditions. For the first antenna selection scheme (AS1), where the user-antenna is selected in order to maximize the overall channel gain between the user and the relay, we give an explicit analytical proof that for binary modulations, the system achieves full diversity order of \( \min(N_A, N_R) \times N_R \) in the multiple-access (MA) phase, where \( N_A \) and \( N_R \) denote the number of antennas at user \( A \), user \( B \) and relay \( R \) respectively. We present a detailed investigation of the diversity order for the MIMO-PNC system with AS1 in the MA phase for any modulation order. A tight closed-form upper bound on the average symbol error rate (SER) is also derived for the special case when \( N_R = 1 \), which is valid for any modulation order. We show that in this case the system fails to achieve transmit diversity in the MA phase, as the system diversity order drops to 1 irrespective of the number of transmit antennas at the user nodes. Additionally, we propose a Euclidean distance (ED) based user-antenna selection scheme (AS2) which outperforms the first scheme in terms of error performance. Moreover, by deriving upper and lower bounds on the diversity order for the MIMO-PNC system with AS2, we show that this system enjoys both transmit and receive diversity, achieving full diversity order of \( \min(N_A, N_R) \times N_R \) in the MA phase for any modulation order. Monte Carlo simulations are provided which confirm the correctness of the derived analytical results.

I. INTRODUCTION

Wireless PNC has received a lot of attention among researchers in recent years due to its inherent desirable properties of delay reduction, throughput enhancement and better spectral efficiency. The advantage of PNC can easily be seen in a two-way relay channel (TWRC), where bidirectional information exchange takes place in the half-duplex mode between two users \( A \) and \( B \) with the help of a relay \( R \). In a TWRC, PNC requires only two time slots to exchange the information between the users compared to three time slots required by traditional network coding [1]. In the first time slot, also termed the multiple access (MA) phase, both users \( A \) and \( B \) simultaneously transmit their data to the relay \( R \). Based on its received signal, the relay forms the maximum-likelihood (ML) estimate of the pair of transmitted user constellation symbols. This estimate of the pair of user symbols is then mapped to a network-coded constellation symbol using the denoise-and-forward (DNF) protocol [2] and the relay broadcasts this to both users in the next time slot, called the broadcast (BC) phase. User constellation symbol pairs which are mapped to the same complex number in the network-coded constellation are said to form a cluster. Using its own message transmitted in the previous MA phase, \( A \) can decode the message transmitted from \( B \) and vice versa.

Different aspects of PNC relating to communication theory, information theory, wireless networking, finite-field and infinite-field PNC, as well as synchronization issues and the use of PNC for passive optical networks were discussed in [1], [3], [4]. The first software radio based implementation of PNC was reported in [5], together with a discussion on related problems and solutions.

A performance comparison among four time slot transmission scheme (non network-coded scheme), three time slot transmission scheme (network coding scheme) and two time slot transmission scheme (PNC scheme) for TWRCs in terms of bit-error rate (BER) and maximum sum-rate was presented in [6]. Closed-form expressions for a tight upper and lower bound on the average SER at the relay and tight bounds on the average end-to-end BER for a PNC system in a Rayleigh fading channel were presented in [7]. An exact BER expression for a PNC system operating over a TWRC exhibiting fading was presented in [8] using Craig’s polar coordinate form. A general framework for the symbol-error-rate (SER) performance analysis of PNC systems operating over AWGN channels was presented in [9]. In [10], a high signal-to-noise ratio (SNR) analysis for the error performance at the relay in a PNC system with binary or higher-order real/complex modulation as well as for real and complex channel coefficients was presented.

In [11], a linear vector PNC scheme was proposed for an open-loop spatial MIMO TWRC, where no CSI was available at the users’ end. An explicit solution for the network coding (NC) generator matrix to minimize the error probability at high SNR was proposed, and a novel closed-form expression for the average SER in Rayleigh fading was also presented. In [12], a multiuser communication scenario was considered, where \( K \) users simultaneously communicate with a receiver using space-time coded MIMO with linear PNC. All the user messages were encoded by the same linear dispersion space-time code and a novel iterative search algorithm was used.
to optimize the space-time coded linear PNC mapping. It was shown that the system achieves full rate and full diversity while achieving the maximum coding gain. However, despite the advantages of the linear MIMO-PNC systems proposed in [11] and [12], the implementation cost of these systems are high – all of the antennas from all users are utilized simultaneously to transmit the data, which require a large number of radio-frequency (RF) chains. One of the key differences between the MIMO-PNC systems proposed in [11], [12] and the MIMO-PNC system proposed in this paper is that we take advantage of switched diversity (by virtue of user-antenna selection) to reduce the required number of RF chains - only one antenna per user is active during transmission. This reduces the overall cost of the system, while maintaining the full diversity order.

In the case of a fixed network coding (FNC) system, the network code applied at the relay is always fixed and does not depend on channel conditions. One of the bottlenecks in the FNC system limiting the error performance is the existence of singular fade states [13] that result in the phenomenon of distance shortening, which will be explained later in this paper. To solve this problem, a number of adaptive physical-layer network coding (ANC) schemes have been proposed [13]–[15], where the relay adaptively selects the network mapping that offers the best performance based on the channel conditions. A similar scheme for multiple-input multiple-output (MIMO) two-way relaying was applied in [16], and it was shown that the minimum distance between the network-coded constellation points at the relay becomes zero when all of the rows of the channel matrix belong to a finite number of subspaces referred to as singular fade subspaces. A computationally efficient analytical framework to choose the appropriate adaptive network codes at the relay for heterogeneous symmetric PNC was presented in [17]. A detailed introduction to wireless multi-way relaying using ANC was presented in [18]. It was shown in [19] that every valid network mapping can be represented by a Latin square and that this relationship can be used to obtain network maps with optimized intercluster distance profiles. It has been shown in [13], [14] that for a 4-ary modulation scheme in the MA phase, ANC may result in a 5-ary network map, and therefore a non-standard 5-ary modulation scheme will be required for the BC phase under certain channel conditions. Although ANC alleviates the problem of distance shortening in an efficient way, the related system complexity increases significantly due to the required clustering algorithm, and the increased cardinality of the relay’s transmit constellation may incur a sacrifice in the reliability in the BC phase.

In [20], tight upper and lower bounds on the average BER of a multiple-antenna PNC system were presented, where the AS scheme based on the maximization of the overall channel gain between the user and the relay (which we will refer to as AS1 in the rest of this paper) is applied at both user nodes, and the users employ BPSK modulation. It was stated that for BPSK modulation, the MIMO-PNC system achieves a diversity order of min\( (N_A, N_R) \times N_t \) in the MA phase, and an explicit proof of the diversity order was provided for the special cases of the MISO-PNC system \( (N_A, N_R > 1, N_t = 1) \) and the SIMO-PNC system \( (N_A, N_R = 1, N_t > 1) \).

A popular paradigm for antenna selection in the literature is that based on the ED criterion, where the antenna at the transmitter node is selected such that the minimum ED between different symbols in the received constellation is maximized. Such an antenna selection scheme was discussed in [21] for spatial multiplexing (SMx) systems, in [22] for opportunistic PNC scheduling and in [23] for spatial modulation (SM). In [21], a SMx system with \( \mathcal{N}_t \) transmit antennas, \( \mathcal{N}_r \) receive antennas, and a \( 1 : \mathcal{N}_t > \mathcal{N}_r > \mathcal{N}_s > \mathcal{N}_i \) multiplexer was considered, where a low-bandwidth, zero-delay, error-free feedback path indicated the optimal \( \mathcal{N}_i \) of \( \mathcal{N}_r \) antennas for transmission, computed using current channel state information at the receiver. For the ML receiver, the authors proposed to choose the subset of transmit antennas which resulted in a constellation with largest minimum ED. It was shown that this ED based AS resulted in minimum error rate for the SMx system based on the ML receiver. In [22], a three-way wireless communication system was considered, where each user desires to transmit independent data to other users via a relay. Since the overall throughput of such a system is limited by the worst channel, a scheduling system employing PNC was considered to optimize the overall system throughput. It was shown that the selection of a user-pair based on the largest minimum ED between the superposed constellation at the relay resulted into better overall throughput compared to that of the channel-norm based user-pair selection and round-robin based scheduling. In [23], a comprehensive analysis of the transmit diversity order for the ED based antenna selection scheme in an SM system (consisting of \( \mathcal{N}_t \) transmit antennas and \( \mathcal{N}_r \) receive antennas) was presented. For each transmission, \( \mathcal{N}_s \) out of \( \mathcal{N}_r \) transmit antennas were selected in order to achieve spatial switching gain (SSG). It was proved explicitly that such an SM system enjoys both transmit and receive diversity, achieving a diversity order of \( \mathcal{N}_s (\mathcal{N}_t - \mathcal{N}_s + 1) \).

It is important to note that the error performance analysis in the BC phase of the PNC system in a TWRC is similar to a traditional (non network-coded) point-to-point communication system and the end-to-end error performance of the PNC system will be dominated by the error performance in the MA phase. Therefore, in this paper we analyze the error performance and the diversity order of the MIMO-PNC system in the MA phase only. The main contributions of this paper are summarized as follows:

- We give an explicit analytical proof that the diversity order of the MIMO-PNC system with binary modulation and AS1 is equal to \( \min(N_A, N_R) \times N_t \).
- We provide a detailed investigation of the error rate performance and diversity order of the MIMO-PNC system with AS1 for any modulation order \( M \). A closed-form expression for a tight upper bound on the average SER is derived for the special case when \( N_R = 1 \). The presented diversity analysis confirms that the performance of the MIMO-PNC system with AS1 degrades severely for non-binary modulations due to the distance shortening phenomenon at the relay, and the system fails to achieve transmit diversity. We give an analytical proof that the diversity order of such system drops to 1 for the case
when \( N_R = 1 \).

- We propose an ED based AS scheme (which we will refer to as AS2 in the rest of this paper) for the MIMO-PNC system to mitigate the deleterious effects of distance shortening at the relay. Furthermore, we derive upper and lower bounds on the diversity order and prove that the system with AS2 achieves a full diversity order of

\[
\min(N_A, N_B) \times N_R
\]

for any modulation order.

The rest of this paper is organized as follows: In Section II, we present the system model for the MIMO-PNC system. In Section III, we introduce the AS1 and AS2 schemes and give an illustration of the performance superiority of AS2 over AS1. In Section IV, we present a comprehensive error performance analysis of the MIMO-PNC system with AS1 and also provide a diversity analysis. Section V deals with the derivation of upper and lower bounds on the diversity order of the MIMO-PNC system with AS2. In Section VI, we present extensive simulation and analytical results along with discussion. Finally, the conclusion is presented in Section VII.

### II. SYSTEM MODEL

The system model for the MIMO-PNC system is shown in Fig. 1 where two users \( A \) and \( B \) are equipped with \( N_A > 1 \) and \( N_B > 1 \) antennas, respectively, while relay \( R \) is equipped with \( N_R \geq 1 \) antennas. During the MA phase, only one of the antennas from each user is used for signal transmission, and the choice of antennas is based on feedback received from the relay. The channel between user \( m \in \{A, B\} \) and the relay \( R \) is modeled as slow Rayleigh fading with perfect CSI available at \( R \) only. We assume that the channel remains constant during a frame transmission and changes independently from one frame to another. Hence the channel coefficient between the \( i \)-th antenna of user \( m \) and the relay is distributed according to \( CN(0,1) \). Both users employ the same unit-energy \( M \)-ary constellation \( \mathcal{X} \), and \( \Delta \mathcal{X} \) denotes the difference constellation set of \( \mathcal{X} \), defined as

\[
\Delta \mathcal{X} = \{\Delta x = x - x' | x, x' \in \mathcal{X}\}.
\]

Let \( s_m \in \mathbb{Z}_M = \{0,1, \ldots, M-1\} \) denote the message symbol at user \( m \), and \( x_m = \mathcal{F}(s_m) \in \mathcal{X} \) denote the corresponding transmitted constellation symbol, where \( \mathcal{F} \) denotes the constellation mapping function. The signal vector received at the relay during the MA phase is

\[
y = \sqrt{E_A} h_A x_A + \sqrt{E_B} h_B x_B + n,
\]

where \( n \in \mathbb{C}^{N_R \times 1} \) denotes the noise vector at the relay whose elements are assumed to be distributed according to \( CN(0,1) \), \( E_m \) denotes the energy of the transmitted signal from user \( m \) and \( h_m = [h_{m,1} \ h_{m,2} \ \cdots \ h_{m,N_R}]^T \in \mathbb{C}^{N_R \times 1} \) is the channel coefficient vector of the link between the selected antenna of user \( m \) and the relay antennas (based on the AS scheme). The relay’s goal is to determine the network-coded symbol \( s_R = \mathcal{M}_c(s_A, s_B) \), where \( \mathcal{M}_c : \mathbb{Z}_M^2 \rightarrow \mathbb{Z}_M \) is the PNC mapping, or equivalently to determine the corresponding constellation symbol \( x_R = \mathcal{F}(s_R) = \mathcal{M}_c(s_A, s_B) \), where \( \mathcal{M}_c : \mathcal{X}^2 \rightarrow \mathcal{X} \) represents the “constellation-domain” version of the PNC mapping. Table I shows an example PNC mapping for QPSK modulation, where the PNC mapping \( \mathcal{M}_c : \mathbb{Z}_M^2 \rightarrow \mathbb{Z}_M \) represents bitwise addition (XOR) in \( \mathbb{Z}_M \).

The relay will form an estimate of \( x_R \), denoted by \( \hat{x}_R \), as follows. First, the relay computes the ML estimate of the transmitted symbol pair \( (x_A, x_B) \in \mathcal{X}^2 \) given by

\[
(\hat{x}_A, \hat{x}_B) = \arg\min_{(x_A, x_B) \in \mathcal{X}^2} \left\| y - \sqrt{E_A} h_A x_A - \sqrt{E_B} h_B x_B \right\|.
\]

Having this joint estimate \( (\hat{x}_A, \hat{x}_B) \in \mathcal{X}^2 \), the relay calculates \( \hat{x}_R = \mathcal{M}_c(\hat{x}_A, \hat{x}_B) \). The error performance at the relay will depend on the minimum distance between the signal points in different clusters, defined as

\[
d_{\min}(h_A, h_B) = \min_{(x_A, x_B) \in \mathcal{X}^2} \left\| \sqrt{E_A} h_A (x_A - x_A') + \sqrt{E_B} h_B (x_B - x_B') \right\|.
\]

It is clear from (3) that the value of \( d_{\min} \) depends on the channel between the users and the relay. In general, when the values of channel coefficients are such that the distance between the clusters is significantly reduced, the phenomenon is called distance shortening.

In the next section, we introduce two different AS schemes and explain the performance superiority of one over the other with the help of an example.

### III. USER-ANTENNA SELECTION FOR PNC

This section presents two different AS schemes for the MIMO-PNC system. In the first scheme (AS1), the user-
antenna is selected in order to maximize the overall channel gain between the user and the relay. Hence we define

$$3_m = \max_{1 \leq i \leq N_m} \sum_{j=1}^{N_R} |h_{m,i,j}|^2,$$  (4)

where $h_{m,i,j} \sim CN(0, 1)$ is the channel coefficient between the $i$th antenna of user $m$ and $j$th antenna of relay $R$. Since $h_{m,j}$ is the channel coefficient between the selected antenna of user $m$ and $j$th antenna of $R$, we may write

$$3_m = \sum_{j=1}^{N_R} |h_{m,j}|^2.$$  (5)

We present the probability density function (PDF) of $3_m$ in the following proposition.

**Proposition 1:** The PDF of $3_m$ is given by

$$f(3_m) = \frac{N_m}{(N_R - 1)!} \sum_{k_0=0}^{N_m-1} \binom{N_m - 1}{k_0, \ldots, k_{N_R}} (-1)^{N_R-1-k_0} \times \prod_{j=1}^{N_R-1} \left( \frac{1}{j!} \right)^{3_m} e^{-N_m - k_0} 3_m^{s-1} \exp[-(N_m - k_0)3_m].$$  (6)

**Proof:** See Appendix A.

In contrast to this, in the second scheme (AS2) the user-antenna of each user is selected such that the minimum ED between the clusters at the relay is maximized. Let $I = \{(i,j) : 1 \leq i \leq N_A, 1 \leq j \leq N_B\}$ be the set which enumerates all of the possible $N = N_A \times N_B$ combinations of selecting one antenna from each user. Among these $n$ combinations, the set of user-antennas that maximizes the minimum ED between the clusters is obtained as

$$I_{ED} = \arg \min_{I \in I} \left\{ \min_{x,x',x''} \mathbb{E} \left[ \left| H_I \left( \frac{\sqrt{E_A} x_A}{\sqrt{E_B} x_B} \right) - \frac{\sqrt{E_A} x_A'}{\sqrt{E_B} x_B'} \right|^2 \right] \right\}$$  (7)

where $H_I = [h_{A,i}^{T}, h_{B,j}^{T}] \in \mathbb{C}^{N_R \times 2}$, $h_{A,i}^{T} = [h_{A,i,1} \ldots h_{A,i,N_B}]^{T}$, $h_{B,j}^{T} = [h_{B,j,1} \ldots h_{B,j,N_B}]^{T}$, and $H_{I_{ED}} = [h_{A}^{T}, h_{B}^{T}] \in \mathbb{C}^{N_R \times 2}$ is the optimal channel matrix.

To understand the performance superiority of AS2 over AS1, we first consider a simple example of one transmission slot where the users transmit their messages using QPSK modulation.

Suppose that $N_A = N_B = 2$, $N_R = 1$ and $h_{A,1,1} = (1 + i)/\sqrt{2}$, $h_{A,2,1} = (1 - 0.5i)/\sqrt{2}$, $h_{B,1,1} = (1 - 0.8i)/\sqrt{2}$ and $h_{B,2,1} = (1 + 0.7i)/\sqrt{2}$. In this case, since $|h_{A,1,1}| > |h_{A,2,1}|$ and $|h_{B,1,1}| > |h_{B,2,1}|$, AS1 will select the antenna combination $I = (1, 1)$. With this combination the minimum distance between the clusters at the relay becomes very small, which can lead to an incorrect ML estimate at the relay. Fig. 2 shows a plot, for AS1, of the noise-free received signal at the relay, i.e., $h_{x,A} + h_{x,B}$ (here we assume $E_A = E_B = 1$), together with the corresponding network-coded symbols, where each 2-tuple in the figure represents $(s_A, s_B)$.

In contrast to this, the proposed antenna selection scheme (AS2) chooses $I = I_{ED} = (1, 2)$ as the optimal combination and the resulting network-coded symbols are shown in Fig. 3. It is clear that AS2 overcomes the distance shortening phenomenon. The following section presents the error performance analysis and the diversity analysis of the MIMO-PNC system with AS1.

IV. AS1: ANTENNA SELECTION BASED ON THE MAXIMUM OVERALL CHANNEL GAIN

For the error performance analysis of AS1, we use the union-bound approach given in [13] rather than that the approach given in [20] which applies only for binary modulations. Using [13, eq. (6)], the average SER for FNC is given by (8), shown at the beginning of the next page, where $\mathbb{E}[-]$ is the expectation operator (here the expectation is performed with respect to $h_A$ and $h_B$) and $\mathbb{P}_n(x_1, x_2) \rightarrow (x_1', x_2')$ denotes the pairwise error probability, i.e., the probability that the signal pair $(x_1', x_2')$ is more likely than $(x_1, x_2)$ from the receiver's
perspective. An upper bound on the average SER can be given by

\[
P_e \leq \frac{1}{M^2} \sum_{(x_A,x_B) \in X^2} \sum_{(x'_A,x'_B) \in X^2 \setminus M_c(x_A,x_B) \cup M_c(x'_A,x'_B)} E \left[ Q \left( \frac{1}{\sqrt{2N_0}} \right) \right.
\]

\[
\times \left. \left( \frac{-E_A h_A x_m - E_c h_c x_n}{\sqrt{E_m h_m x_m - E_n h_n x_n}} \right) \right],
\]

where \( Q(\cdot) \) is the Gaussian Q-function. The Chernoff bound on the Q-function used in [13] results in a loose upper bound for the present case and hence we use the Chiani approximation [24, eq. (14)] instead, yielding (9), shown above.

Now we analyze the three different terms on the right-hand side of (8) separately as follows:

**Case I** – When \( x_A \neq x'_A \) and \( x_B = x'_B \): In this case \( \Delta x_B = 0 \) and hence

\[
\gamma_1 = \exp \left( \frac{-E_A}{4N_0} \| h_A \| \Delta x_A \|^2 \right) \exp \left( \frac{-E_A \| \Delta x_A \|^2}{4N_0} \sum_{j=1}^{N_R} |h_{A,j}|^2 \right),
\]

and

\[
\gamma_2 = \exp \left( \frac{-E_A}{3N_0} \| h_A \| \Delta x_A \|^2 \right) \exp \left( \frac{-E_A \| \Delta x_A \|^2}{3N_0} \sum_{j=1}^{N_R} |h_{A,j}|^2 \right).
\]

Defining \( \Theta_{A,1} = \mathbb{E}[\gamma_1]/12 \) and \( \Theta_{A,2} = \mathbb{E}[\gamma_2]/4 \), the average SER arising from the case when \( x_A \neq x'_A \) and \( x_B = x'_B \) can be written as

\[
\mathcal{P}\left( (x_A,x_B) \rightarrow (x'_A,x'_B) \right) \leq \frac{1}{M^2} \sum_{(x_A,x_B) \in X^2 \setminus M_c(x_A,x_B)} \left( \Theta_{A,1} + \Theta_{A,2} \right).
\]

The closed-form expressions for \( \Theta_{A,1} \) and \( \Theta_{A,2} \) is given by (12) and (13) respectively, shown on the next page. The derivation of the closed-form expression for \( \Theta_{A,1} \) is presented in Appendix B and the closed-form expression for \( \Theta_{A,2} \) can be derived in the same fashion.

**Case II** – When \( x_A = x'_A \) and \( x_B \neq x'_B \): In this case \( \Delta x_A = 0 \) and hence the average SER arising from the case when \( x_A = x'_A \) and \( x_B \neq x'_B \) can be written as

\[
\mathcal{P}\left( (x_A,x_B) \rightarrow (x'_A,x'_B) \right) \leq \frac{1}{M^2} \sum_{(x_A,x_B) \in X^2 \setminus M_c(x_A,x_B)} \left( \Theta_{B,1} + \Theta_{B,2} \right),
\]

where \( \Theta_{B,1} = \mathbb{E}[\gamma_1]/12 \) and \( \Theta_{B,2} = \mathbb{E}[\gamma_2]/4 \). The closed-form expression for \( \Theta_{B,1} \) can be obtained by replacing \( N_A, E_A \) and \( \Delta x_A \) by \( N_B, E_B \) and \( \Delta x_B \), respectively, in (12). The closed-form expression for \( \Theta_{B,2} \) can be obtained in a similar fashion using (13).

**Case III** – When \( x_A \neq x'_A \) and \( x_B \neq x'_B \) and \( M_c(x_A,x_B) \neq M_c(x'_A,x'_B) \): This case is possible only for \( M > 2 \), because for the case of binary modulation (e.g., BPSK), if \( x_A \neq x'_A \) and \( x_B \neq x'_B \), both \( (x_A, x_B) \) and \( (x'_A, x'_B) \) will lie in the same cluster for fixed network coding, i.e., \( M_c(x_A,x_B) = M_c(x'_A,x'_B) \) and hence a confusion among these pairs will not cause a symbol error event. Using (9), \( \mathbb{E}[\gamma_1] \) for the given case can be expressed as (15), shown on the next page.

Since it is difficult in general to determine the PDF of \( \sum_{j=1}^{N_R} |h_{A,j}|^2 \), we consider here the analytically tractable case where \( N_R = 1 \). In this case, \( \mathbb{E}[\gamma_1] \) is given by (16), shown on the next page, where \( \theta = \angle h_A - \angle h_B \) is a random variable uniformly distributed over \([\pi, \pi)\). (a) holds due to the fact that \( \exp(\cos \theta) \) is an even function of \( \theta \) and the integration w.r.t \( \theta \) is solved using [25, p. 376], where \( I_0(\cdot) \) is the modified Bessel function of the first kind; (b) is obtained from (a) using the fact that the PDF of \( f(|h_m|), m \in \{A,B\} \) can be found by putting \( N_R = 1 \) in (6). Then the inner integral in (b) is solved using [26, p. 306], yielding (c). Furthermore, \( \Psi_{A,B} \), \( \Psi_{B,A} \) and \( \Omega_{A,B} \) in (16) are defined as \( \Psi_{A,B} = 1 + \frac{E_{B} |\Delta x_A|^2}{4N_0} \), \( \Psi_{B,A} = 1 + \frac{E_{A} |\Delta x_B|^2}{4N_0} \) and \( \Omega_{A,B} = \left( \frac{\sqrt{E_{A} E_{B} |\Delta x_A \Delta x_B|}}{2N_0} \right)^2 \).
\[ \Theta_{A,1} = \frac{N_A}{12(N_R - 1)!} \sum_{k_0 + k_1 + \ldots + k_{N_R} = N_A - 1} (N_A - 1)_{N_A - 1 - k_0} \left( \prod_{j=0}^{N_R-1} \left( \frac{1}{j!} \right)^{k_j} \right) \left( N_R + s - 1! \right) \left( \frac{E_A |\Delta x_A |^2}{4N_0} + N_A - k_0 \right)^{(N_R + s + s)} \] 

\[ \Theta_{A,2} = \frac{N_A}{4(N_R - 1)!} \sum_{k_0 + k_1 + \ldots + k_{N_R} = N_A - 1} (N_A - 1)_{N_A - 1 - k_0} \left( \prod_{j=0}^{N_R-1} \left( \frac{1}{j!} \right)^{k_j} \right) \left( N_R + s - 1 \right) \left( \frac{E_A |\Delta x_A |^2}{3N_0} + N_A - k_0 \right)^{(N_R + s + s)} \].

\[
E[T_1] = E \left[ \exp \left( -\frac{1}{4N_0} \left| \sqrt{E_A} h_A \Delta x_A + \sqrt{E_B} h_B \Delta x_B \right|^2 \right) \right] = E \left[ \exp \left( -\frac{1}{4N_0} \sum_{j=1}^{N_R} \left| \sqrt{E_A} h_{A,j} \Delta x_A + \sqrt{E_B} h_{B,j} \Delta x_B \right|^2 \right) \right] 
\leq E \left[ \exp \left( -\frac{1}{4N_0} \left| E_A |\Delta x_A |^2 \sum_{j=1}^{N_R} |h_{A,j}|^2 + E_B |\Delta x_B |^2 \sum_{j=1}^{N_R} |h_{B,j}|^2 + 2\sqrt{E_A E_B} |\Delta x_A \Delta x_B | \sum_{j=1}^{N_R} |h_{A,j} h_{B,j}|^2 \right) \right].
\]

\[
E[T_1] = E \left[ \exp \left( -\frac{1}{4N_0} \sqrt{E_A} h_A \Delta x_A + \sqrt{E_B} h_B \Delta x_B \right)^2 \right] = E \left[ \exp \left( -\frac{1}{4N_0} \sum_{j=1}^{N_R} \sqrt{E_A} h_{A,j} \Delta x_A + \sqrt{E_B} h_{B,j} \Delta x_B \right)^2 \right] 
\leq E \left[ \exp \left( -\frac{1}{4N_0} \left( E_A |\Delta x_A |^2 \sum_{j=1}^{N_R} |h_{A,j}|^2 + E_B |\Delta x_B |^2 \sum_{j=1}^{N_R} |h_{B,j}|^2 + 2\sqrt{E_A E_B} |\Delta x_A \Delta x_B | \sum_{j=1}^{N_R} |h_{A,j} h_{B,j}|^2 \right) \right].
\]

\[
\Xi_{A,k} = 1 + \frac{E_A |\Delta x_A |^2}{3N_0}, \quad \Xi_{B,l} = 1 + \frac{E_B |\Delta x_B |^2}{3N_0} \quad \text{and} \quad \Phi_{A,B} = \left( \frac{2\sqrt{E_A E_B} |\Delta x_A \Delta x_B |}{3N_0} \right)^2.
\]

Solving in the same fashion for \( E(T_2) \), we obtain

\[ E(T_2) = \sum_{k=1}^{N_A} \sum_{l=1}^{N_B} \binom{N_A}{k} \binom{N_B}{l} \left( -1 \right)^{(k+1-2)} (\Xi_{A,k} \Xi_{B,l} - 1) \Phi_{A,B}. \]

where \( \Xi_{A,k} = 1 + \frac{E_A |\Delta x_A |^2}{3N_0}, \quad \Xi_{B,l} = 1 + \frac{E_B |\Delta x_B |^2}{3N_0} \) and \( \Phi_{A,B} = \left( \frac{2\sqrt{E_A E_B} |\Delta x_A \Delta x_B |}{3N_0} \right)^2 \). Hence, the average SER arising from the case when \( x_A \neq x_A' \), \( x_B \neq x_B' \) and \( \mathcal{M}_c(x_A, x_B) \neq \mathcal{M}_c(x_A', x_B') \) becomes equal to (17), shown at the beginning of the next page, where

\[ \xi_1 = \sum_{k=1}^{N_A} \sum_{l=1}^{N_B} \binom{N_A}{k} \binom{N_B}{l} \frac{(-1)^{k+1-2}}{12 (\Psi_{A,k} \Psi_{B,l} - \frac{\Omega_{A,B}}{4k})}. \]

An upper bound on the average SER for the MIMO-PNC system with AS1 for the special case of \( N_R = 1 \) can be obtained by putting \( N_R = 1 \) in (11) and (14), and then adding (11), (14) and (17).

1) Diversity Analysis: To determine the diversity order of the MIMO-PNC system with AS1, we analyze the asymptotic decay rates of all three terms (i.e., corresponding to the three cases discussed above) on the right-hand side of (8) separately. Considering first Case I \((x_A \neq x_A' \) and \( x_B = x_B' \)), using (26) and (28), the PDF of \( f(\zeta) \) can be rewritten as

\[ f(\zeta) = \frac{N_A N_R^{-1} \exp(-3\zeta)}{(N_R - 1)!} \left[ 1 - \exp(-3) \sum_{j=0}^{N_R-1} \frac{3^j}{j!} \right]^{N_A^{-1}}. \]
Note that for the case when \( r_{12} \) into (18), the diversity order of the term arising from \( z \) can be assumed to be absent (as the transmission from user 1 does not cause any error at the relay) and the PNC system reduces to simple single transmitter – single receiver system. Therefore, using [27, Proposition 1], we can conclude that \( \Theta_{A,1} \) and \( \Theta_{A,2} \) decay as \( (E_{A}/N_0)^{-N_{RA,1}} \) for large values of \( E_{A}/N_0 \), and hence the diversity order of the term arising from Case I is \( N_{RA,2} \). Following the same argument, \( \Theta_{B,1} \) and \( \Theta_{B,2} \) decay as \( (E_{B}/N_0)^{-N_{BA,2}} \) for large values of \( E_{B}/N_0 \), the diversity order of the term arising from Case II is \( N_{BA,2} \).

For Case III with \( N_R = 1 \), substituting the values of \( \Psi_{A,k} \), \( \Psi_{B,l} \), and \( \Omega_{AB} \) into (18), \( \xi_1 \) becomes (where we define \( E_{\text{min}} \equiv \min(E_{A}, E_{B}) \))

\[
\xi_1 = \frac{1}{12} \sum_{k=1}^{N_A} \sum_{l=1}^{N_B} \left( N_A \right) \left( N_B \right) \left( -1 \right)^{k+l-2} \frac{E_A |\Delta x_A|^2}{4kN_0} \frac{E_B |\Delta x_B|^2}{4lN_0}
\]

where \( g(k, l) = \frac{|\Delta x_A|^2}{4k} + \frac{|\Delta x_B|^2}{4l} \).

Using the binomial expansion, \( \xi_1 \) can be rewritten as

\[
\xi_1 = \frac{1}{12} \sum_{k=1}^{N_A} \sum_{l=1}^{N_B} \left( N_A \right) \left( N_B \right) \left( -1 \right)^{k+l-2} \frac{E_A |\Delta x_A|^2}{4kN_0} \frac{E_B |\Delta x_B|^2}{4lN_0}
\]

Similarly, for \( \xi_2 \) it can be shown that

\[
\xi_2 < B_1 \left( \frac{E_{\text{min}}}{N_0} \right)^{-1} + O \left( \frac{E_{\text{min}}}{N_0} \right)^{-1}
\]

where

\[
B_1 = 4 \sum_{k=1}^{N_A} \sum_{l=1}^{N_B} \left( N_A \right) \left( N_B \right) \left( -1 \right)^{k+l-2} \frac{E_{\text{min}} |\Delta x_A|^2}{4kN_0} \frac{E_{\text{min}} |\Delta x_B|^2}{4lN_0}
\]

From (17), (19) and (20), it is clear that the average symbol error probability due to the case when \( N_R = 1 \), \( x_A \neq x_A' \) and \( x_B \neq x_B' \) for higher values of \( E_A \) and \( E_B \) while the second term decays as \( (E_{A}/N_0)^{-N_{BA,2}} \) for higher values of \( E_{A}/N_0 \). It is important to recall that for binary modulations \( (M = 2) \) (as in [20]), the third term in the right-hand side of (8) (which is analyzed in Case III) do not contribute to the symbol error probability at the relay and hence the overall system diversity order becomes \( \min(N_{A,2}, N_{B,2}) \times N_R \). This completes the proof for the diversity order of the MIMO-PNC system with BPSK modulation.

For the MISO-PNC system \( N_R = 1 \) with AS1 and non-binary modulations \( (M > 2) \), since the first two terms on the right-hand side of (8) will decay as \( (E_A/N_0)^{-N_{A,2}} \) and \( (E_A/N_0)^{-N_{B,2}} \) respectively, for higher values of \( E_A/N_0 \) and \( E_B/N_0 \), respectively, while the third term will decay as \( (E_{\text{min}}/N_0)^{-1} \) for higher values of \( E_{\text{min}}/N_0 \), the system
diversity order becomes \( \min(N_A, N_B, 1) = 1 \), irrespective of the number of antennas at the users.

In case of the MIMO-PNC system with AS1 and non-binary modulations, since it is difficult in general (as concluded from (15)) to represent/bound the average SER for Case III in a tractable form, we present extensive simulation results (detailed in Section VI), which confirms that the system fails to achieve transmit diversity and the system diversity order drops to \( N_R \).

The following section deals with the derivation of upper and lower bounds on the average SER of the MIMO-PNC system with AS2.

V. AS2: ANTENNA SELECTION BASED ON ED METRIC

To analyze the performance of AS2, we follow the approach adopted in [23] and [28]. In [23] the authors analyzed the error performance of an ED based antenna selection scheme for SM. The analysis for PNC differs due to that fact that for an SM system, only a single antenna is active during transmission, while for a TWRC with PNC, two antennas (one from each user) transmit simultaneously. In [28] the authors used a similar technique to analyze the error performance of an SMx system with ED based transmit antenna selection.

Given a set of user-antenna indices \( I = (i, j) \in J \), the set of possible transmit vectors for the PNC system can be defined as \( \mathcal{G}_I = \left\{ \left[ \sqrt{E_A} \Delta x_A^{(1)} \right]^T, \left[ \sqrt{E_B} \Delta x_B^{(1)} \right]^T \right\} \mid x_A, x_B \in \mathcal{S} \} \), where \( e_i \) and \( e_j \) are row vectors of length \( N_A \) and \( N_B \) respectively with all zero elements apart from a 1 at the \( i \)th and \( j \)th position, respectively. Let \( \xi_I(x_A, x_B) = \left[ \sqrt{E_A} \Delta x_A^{(1)} \right]^T \).

1) Lower bound on diversity order: Defining \( \Delta \mathcal{C}_I = \{ z_I(x_A^{(1)}), x_B^{(1)}), z_I(x_A^{(2)}), x_B^{(2)} \} | x_A^{(1)}, x_A^{(2)}, x_B^{(1)}, x_B^{(2)} \in \mathcal{S}, M_c \left( x_A^{(1)}, x_B^{(1)} \right) \neq M_c \left( x_A^{(2)}, x_B^{(2)} \right) \} \) as the set of difference vectors corresponding to the codebook \( \mathcal{G}_I \), the set of matrices \( \Delta \mathcal{D} \) can be defined as

\[
\Delta \mathcal{D} = \{ [x_1, x_2, \ldots, x_n] | x_k \in \Delta \mathcal{C}_I \forall k \in \{1, 2, \ldots, n\} \}.
\]

Each element in \( \Delta \mathcal{D} \) will be of size \( (N_A + N_B) \times n \) where \( n = N_A \times N_B \). Let \( r_{min} \) be defined as

\[
r_{min} = \min \{ \text{rank}(X) | X \in \Delta \mathcal{D} \}.
\]

The minimum number of linearly independent columns in \( X \), i.e., \( r_{min} \) will be \( \min(N_A, N_B) \). To understand this, consider an example when \( N_A = 3 \) and \( N_B = 2 \). The structure of each element in \( \Delta \mathcal{D} \) is given by (21) at the beginning of this page.

Thanks to definition of \( \Delta \mathcal{C}_I \), in which the condition \( M_c \left( x_A^{(1)}, x_B^{(1)} \right) \neq M_c \left( x_A^{(2)}, x_B^{(2)} \right) \) ensures that \( z_I \left( x_A^{(1)}, x_B^{(2)} \right) \), implies that in any of the columns of \( X \), both \( \Delta x_A^{(1)} \) and \( \Delta x_B^{(2)} \) cannot be zero simultaneously. If \( \Delta x_A^{(1)} \) and \( \Delta x_B^{(2)} \) are non-zero, they form a non-zero minor (a diagonal matrix) and the minimum possible rank of \( X \) becomes 2. Now if \( \Delta x_B^{(1)} \) is zero and any one from \( \Delta x_A^{(3)} \) or \( \Delta x_B^{(5)} \) is non-zero then also a 2 \( \times \) 2 non-zero minor can be formed using \( \Delta x_A^{(2)} \). A similar argument applies when \( \Delta x_A^{(1)} \) is non-zero and \( \Delta x_B^{(2)} \) is zero and a 2 \( \times \) 2 minor can be formed using \( \Delta x_A^{(4)} \) and \( \Delta x_B^{(4)} \) or \( \Delta x_B^{(6)} \) with the help of a column swap. In the case where \( \Delta x_A^{(1)}, \Delta x_B^{(3)} \) and \( \Delta x_B^{(5)} \) are zero, the \( \Delta x_A^{(4)} \) values in the corresponding columns will be non-zero and they will form three linearly independent columns and the rank of matrix \( X \) will be at least 3. A similar argument is valid for the case when \( \Delta x_A^{(2)}, \Delta x_B^{(4)} \) and \( \Delta x_B^{(6)} \) are zero. On the other hand, if all the \( \Delta x_A^{(4)} \) values are zero then the first three rows will be linearly independent and the rank of \( X \) will be 3. Hence the minimum possible rank of \( X \) is 2 i.e., \( \min(N_A, N_B) \). It is straightforward to generalize this argument to the case of an arbitrary number of antennas at each user.

Let the transmit vectors in the each codebook be denoted as \( C_k = \{ x_i(k) | k \in \{1, 2, \ldots, M^2\} \} \) and the optimal set of user-antennas for any particular channel realization \( H \) be \( I^* \). For \( E_{\min}/N_0 \gg 1 \), the average pairwise error probability between any two different transmit vectors indexed by \( I_1 \) and \( I_2 \) in the codebook \( C_{I}^* \) can be expressed, using the Chernoff bound, as \( [23, \text{eq. } (4)-(10)] \)

\[
\mathbb{P}(\xi_I(x_i) \rightarrow \xi_I(x_j)) \leq \frac{1}{2} \left( \frac{E_{\min} \lambda^*}{4nN_0} \right)^{-N_R r_{min}}.
\]

where \( \lambda^* = \min_{X \in \Delta \mathcal{D}} \lambda_X(X^TX) \) and \( \lambda_X(Y) \) denotes the smallest nonzero eigenvalue of matrix \( Y \). An upper bound on the average SER for AS2 at \( E_{\min}/N_0 \gg 1 \) can therefore be given as

\[
P_e \leq \frac{1}{2M^2} \sum_{x_i \in C_{I}^*} \sum_{x_i \in C_{I}^*} \left( \frac{E_{\min} \lambda^*}{4nN_0} \right)^{-N_R r_{min}}.
\]

It is clear from (22) that the PNC system with AS2 achieves a diversity order lower bounded by \( \min(N_A, N_B) \times N_R \) for any modulation order \( M \).

2) Upper bound on diversity order: The average pairwise error probability i.e., the probability that the signal pair \( \hat{x} = (\sqrt{E_A} \Delta x_A, \sqrt{E_B} \Delta x_B) \) is more likely than \( x = \ldots
Using (24) and (25), the upper bound on the diversity order for MIMO-PNC with AS2 can be given by \( \min(N_A, N_B) \times N_R \) for any modulation order \( M \).

Since both upper and lower bounds on the diversity order for MIMO-PNC with AS2 are equal, we may conclude that the exact diversity order for the MIMO-PNC system is equal to \( \min(N_A, N_B) \times N_R \) for any modulation order \( M \).

VI. Results and Discussion

In this section, we present a performance comparison of the two AS schemes discussed in the previous sections.

A. Simulation setup

For all Monte Carlo simulations, our setup is as follows. We generate random QPSK symbols \( x_A, x_B \in \mathcal{X} \), and then compute \( x_R = M_c(x_A, x_B) \) using the PNC mapping shown in Table I. Next, we generate i.i.d. random samples of \( b_{m,i,j} \) with \( 1 \leq i \leq N_m, 1 \leq j \leq N_R \). For AS1, the index of the optimal user-antenna is generated using (7). The noise vector \( \mathbf{n} \in \mathbb{C}^{N_R \times 1} \) is then generated whose elements are independent and complex Gaussian with zero mean and variance \( N_0 \), and given \( E_A \) and \( E_B \) we obtain the signal vector received at the relay as shown in (1). Finally, we compute the relay’s ML estimate of the transmitted symbol pair \((\hat{x}_A, \hat{x}_B)\) using (2), and this is used to obtain the estimated network-coded symbol, denoted by \( \hat{x}_R = M_c(\hat{x}_A, \hat{x}_B) \). The average SER is measured by counting the number of error events, i.e., \( \hat{x}_R \neq x_R \), and dividing by the number of symbols transmitted.

B. Discussion

In Fig. 4, the SER performance for the two AS schemes is shown for the case when \( N_R = 1 \) with different number of antennas at the users in symmetric channels (i.e., \( E_A / N_0 = E_B / N_0 = E_{\text{min}} / N_0 \)). In the figure legend, ‘UB’
Average SER

10

-10

10

-8

10

-6

10

-4

10

-2

10

0

Average SER

15

16

17

18

Fig. 6. Average SER performance comparison for the two AS schemes in MIMO case \((N_R > 1)\) with symmetric channels, where \(E_A/N_0 = E_B/N_0 = E_{\text{min}}/N_0\).

denotes the upper bound on the average SER for AS1 (as derived in Section IV) and the numbers in parentheses denote \((N_A, N_R, N_B)\). The plots marked ‘UB’ have been drawn by substituting \(N_R = 1\) in (11) and (14), and then adding (11), (14) and (17), whereas the plots for AS1 and AS2 have been drawn using the Monte Carlo simulations (as described in the previous subsection). It is clear from the figure that for AS1 (where the user-antenna is selected based on the maximization of the overall channel gain between the user and the relay), the system diversity order becomes equal to 1 irrespective of the number of antennas at the users’ end, as for higher values of \(E_{\text{min}}/N_0\) the SER curve becomes parallel to \((E_{\text{min}}/N_0)^{-1}\) in each case as was proved in Section IV. It is also worth noting that the derived closed-form expression for the upper bound on the SER is very tight for AS1.

In contrast to this, the PNC system with AS2 (where the user antenna is selected based on the ED metric) outperforms the one with AS1 while achieving a higher diversity order. For the case when \(N_A = N_B = 2\) and \(N_A = 3, N_B = 2\), the average SER in the PNC system with AS2 decays more rapidly as compared to AS1 and becomes parallel to \((E_{\text{min}}/N_0)^{-1}\) for higher values of \(E_{\text{min}}/N_0\) as was proved in Section V. Similarly, for the case when \(N_A = N_B = 3\) the average SER for AS1 decays as \((E_{\text{min}}/N_0)^{-1}\) while for AS2 the average SER decays as \((E_{\text{min}}/N_0)^{-3}\) at higher values of \(E_{\text{min}}/N_0\).

Fig. 5 shows the average SER performance comparison of the two AS schemes in symmetric channels with different number of user antennas and \(N_R = 1\). In this case, \(E_A/N_0 = (E_{\text{min}}/N_0) + 5\) dB (and thus \(E_B/N_0 = E_{\text{min}}/N_0\)). Similar to the previous results, the diversity order achieved by the MISO-PNC system for asymmetric channels with AS1 is 1, while the system with AS2 achieves the full diversity order of \(\min(N_A, N_B)\) (recall that here \(N_R = 1\)).

Fig. 6 shows the average SER performance for the two AS schemes in the MIMO-PNC setting \((N_R > 1)\) for symmetric channels. It is clear from the figure that the average SER for the MIMO-PNC system with AS1 decays as \((E_{\text{min}}/N_0)^{-2}\) for higher values of \(E_{\text{min}}/N_0\) when \((N_A, N_R, N_B)\) is \((2, 2, 2), (2, 2, 3)\) or \((3, 2, 3)\), and it is clear that the system fails to achieve transmit diversity – the diversity order of the system depends only on the number of antennas at the relay. On the other hand, for the MIMO-PNC system with AS2, the average SER decays as \((E_{\text{min}}/N_0)^{-4}\) for higher values of \(E_{\text{min}}/N_0\) when \((N_A, N_R, N_B)\) is \((2, 2, 2)\) or \((2, 2, 3)\). Similarly, the average SER decays as \((E_{\text{min}}/N_0)^{-6}\) for higher values of \(E_{\text{min}}/N_0\) when \((N_A, N_R, N_B)\) is \((3, 2, 3)\). Therefore, it is clear from Fig. 6, that in case of MIMO-PNC, the diversity order is equal to \(N_R\) with AS1 and \(\min(N_A, N_B) \times N_R\) with AS2 resulting in the performance superiority of AS2.

Fig. 7 shows the average SER performance comparison of the two AS schemes in asymmetric channels with different number of user antennas in the MIMO-PNC setting (i.e., \(N_R > 1\)). In this case, \(E_A/N_0 = (E_{\text{min}}/N_0) + 5\) dB (and thus \(E_B/N_0 = E_{\text{min}}/N_0\)). Similar to the previous results for the MIMO-PNC system in symmetric channels, the diversity order achieved by the MIMO-PNC system with AS1 for asymmetric channels is 2 (= \(N_R\)) for the case when \((N_A, N_R, N_B)\) is \((2, 2, 2), (3, 2, 2)\) or \((3, 2, 3)\). On the other hand, the diversity order achieved by the MIMO-PNC system with AS2 is 4 for the case when \((N_A, N_R, N_B)\) is \((2, 2, 2)\) or \((3, 2, 2)\), while the system achieves a diversity order of 6 when \((N_A, N_R, N_B)\) is \((3, 2, 3)\). Hence, it is clear that the MIMO-PNC system with AS2 achieves full diversity order of \(\min(N_A, N_B) \times N_R\) and therefore outperforms the MIMO-PNC system with AS1 in asymmetric channels also.

VII. CONCLUSION

In this paper, we analyzed the error performance of a MIMO-PNC system with fixed network coding under two different user-antenna selection schemes in an asymmetric scenario, where the users may have different number of antennas and different average SNR to the relay, this analysis being valid for any modulation order \(M\). A detailed investigation of the error performance and the diversity order
was presented. It was shown analytically that for the first antenna selection scheme (AS1), where each user-antenna is selected to maximize the overall channel gain between the user and the relay, the MIMO-PNC system achieves full diversity order of \( \min(N_A, N_R) \times N_R \) for binary modulations. For non-binary modulations, a closed-form expression for a tight upper bound on the average SER was derived for the special case \( N_R = 1 \), and diversity analysis confirmed that the system fails to achieve transmit diversity. Simulation results show that the derived upper bound is tight and that for \( N_R > 1 \) the system fails to achieve transmit diversity, as the diversity order of the system remains equal to \( N_R \). To overcome the aforementioned problems, a Euclidean distance based user-antenna selection scheme (AS2) was proposed which outperforms the first scheme in terms of error performance. Upper and lower bounds on the resulting average SER were derived, and it was shown that the MIMO-PNC system with AS2 achieves both transmit and receive diversity, resulting in a full diversity order of \( \min(N_A, N_B) \times N_R \). This new user-antenna selection scheme allows the MIMO-PNC system to avoid the harmful effect of singular fade states without any need for adaptive network codes or nonstandard constellation design, reducing the overall system complexity.

**APPENDIX A**

**PROOF OF PROPOSITION 1**

Since \( h_{m,i,j} \sim CN(0, 1) \), the magnitude of \( h_{m,i,j} \) is Rayleigh distributed. Defining \( \omega_{m,i} \triangleq \sum_{j=1}^{N_R} |h_{m,i,j}|^2 \), then \( \omega_{m,i} \sim Ga(N_R, 1) \). Therefore the PDF\(^{1}\) of \( \omega_{m,i} \) can be written as,

\[
 f_{\omega_{m,i}}(\tau) = \frac{1}{\Gamma(N_R)} \tau^{N_R-1} \exp(-\tau).
\]

For \( N_R > 0 \), the cumulative distribution function (CDF) of \( \omega_{m,i} \) is a special case of that of an Erlang distribution, i.e.,

\[
 F_{\omega_{m,i}}(\tau) = 1 - \exp(-\tau) \sum_{j=0}^{N_R-1} \frac{\tau^j}{j!}.
\]

Using (5), \( 3m = \sum_{j=1}^{N_R} |h_{m,j}|^2 \), we have

\[
 f_{3m}(\tau) = \frac{d}{d\tau} \prod_{i=1}^{N_R} F_{\omega_{m,i}}(\tau) = \frac{d}{d\tau} \left[ 1 - \exp(-\tau) \sum_{j=0}^{N_R-1} \frac{\tau^j}{j!} \right]^{N_R} \left[ 1 - \exp(-\tau) \sum_{j=0}^{N_R-1} \frac{\tau^j}{j!} \right]^{N_R-1} \left[ 1 - \exp(-\tau) \sum_{j=0}^{N_R-1} \frac{\tau^j}{j!} \right]^{N_R-1}.
\]

\[
x_1 = \sum_{k_0+k_1+\cdots+k_{N_R} = N_R-1} (N_R-1)! \sum_{j=0}^{N_R-1} \frac{\tau^j}{j!} \left[ 1 - \exp(-\tau) \sum_{j=0}^{N_R-1} \frac{\tau^j}{j!} \right]^{N_R-1-k_0} \times \prod_{j=0}^{N_R-1} \frac{1}{j!}^{k_j} \tau^j \exp[-(N_R - 1 - k_0)\tau]. \tag{27}
\]

where \( s = \sum_{j=0}^{N_R} k_{j+1} \).

Moreover,

\[
x_2 = \frac{1}{(N_R - 1)!} \tau^{N_R-1} \exp(-\tau). \tag{28}
\]

Using (26), (27) and (28), the closed-form expression for \( f(3m) \) becomes equal to (6).

**APPENDIX B**

**DERIVATION OF THE CLOSED-FORM EXPRESSION FOR \( \Theta_{A1} \)**

For \( 3_A = \sum_{j=1}^{N_R} |h_{A,j}|^2 \), (10) can be rewritten as

\[
 \mathcal{Y}_1 = \exp \left( - \frac{E_A[\Delta \tau x_A]^2}{4N_0} \right) \implies \mathbb{E}[\mathcal{Y}_1] = \int_0^\infty \exp \left( - \frac{E_A[\Delta \tau x_A]^2}{4N_0} \right) f(3_A) d3_A. \tag{29}
\]

Substituting the expression (6) for \( f(3_A) \) into (29) yields

\[
 \mathbb{E}[\mathcal{Y}_1] = \frac{N_A}{(N_R - 1)!} \sum_{k_0+k_1+\cdots+k_{N_R} = N_R-1} \left( \frac{N_A-1}{k_0, \ldots, k_{N_R}} \right)(-1)^{N_R-1-k_0} \times \prod_{j=0}^{N_R-1} \frac{1}{j!}^{k_j} \left[ \frac{E_A[\Delta \tau x_A]^2}{4N_0} \right]^{N_R-s-1} \exp(-\left( \frac{E_A[\Delta \tau x_A]^2}{4N_0} \right) + N_A - k_0) \right] d3_A.
\]

The integration above is solved using [29, p. 322] to obtain (12).

**REFERENCES**


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