Outage and capacity analysis of NOMA systems over dual-hop mixed powerline-wireless channels

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Abstract

Non-orthogonal multiple access (NOMA) has emerged as a promising technology for 5G networks and beyond. In order to fully reap the benefits of NOMA, it is essential to characterize its performance under different channel fading. In this paper, we carry out a performance analysis of the downlink NOMA-based dual-hop mixed powerline/wireless communication (PLC/WLC) system. Specifically, we investigate the system reliability and spectrum efficiency in terms of the outage probability and the ergodic capacity, respectively, where the PLC link undergoes lognormal fading while the WLC link undergoes block Rayleigh fading. The accuracy of the analytical results has been validated, the effect of different system controlling parameters is tested; and the optimal values of power allocation factors are obtained. Also, the system performance is compared to a benchmark system through representative extensive simulations.

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Keywords: Outage probability; Ergodic capacity; Non-orthogonal multiple access; Dual-hop relaying; Hybrid powerline-wireless

1. Introduction

Recently, powerline communication (PLC) technology has been considered for enabling data transfer over already deployed electrical power lines. One promising application of such technology is for smart grid networks, where communication can be established between consumer-end home area networks and transmission and distribution networks. Additionally, PLC can be efficiently used for enabling local area networks, home automation, and the Internet of Things (IoT) using the indoor wiring infrastructure [1]. Additionally, PLC could solve the penetration loss problem of wireless communication links that occurs when wireless signals go through walls and metallic structures. The main advantage of PLC lies in the significant reduction of deployment costs and the wide-spread access in buildings everywhere. Technically speaking, PLC works by superimposing high-frequency information signals, in the range of a few hundred Hz to hundreds of mega-hertz, over the low-frequency electrical power carriers to reuse the existing power lines for communication purposes [2–4]. However, to accomplish reliability for beyond fifth generation (B5G) and 6G networks, flexible, heterogeneous, and hybrid architectures are necessary to harness the benefits of different communication technologies to cover all possible scenarios.

Several studies investigated the coexistence of PLC links in conjunction with other communication links such as visible-light communication (VLC) and wireless communication links [5–10]. The authors in [5] investigated the performance of a dual-hop powerline/wireless communication (PLC/WLC) system through deriving closed-form expressions of the outage probability (OP), bit error rate (BER), and the ergodic capacity (EC). The PLC and wireless channels were modeled by lognormal and Rayleigh fading distributions, respectively. In [6], the authors presented an effective solution to protect the hybrid PLC/WLC system from malicious attacks by utilizing the physical characteristics of PLC channels. In [7], a
power allocation problem for a hybrid system composed of a cascaded PLC/VLC link in parallel to an RF wireless link is investigated. In [8], the authors derived an approximate closed-form expression of the average BER and an exact closed-form expression of the OP under a Nakagami-m distributed WLC link, while the PLC link is characterized by the lognormal distribution and Bernoulli-Gaussian noise. Similar to [8], the authors in [9] derived the average BER for the same system model and channel assumptions. However, they used the Gamma approximation to avoid the complication of using the exact lognormal distribution. Additionally, the authors in [10] considered an amplify and forward (AF) relaying over hybrid PLC/WLC architecture, where a Nakagami-m wireless fading and lognormal power line fading are used.

On the other hand, non-orthogonal multiple access (NOMA) technology is capable of boosting the communication system capacity [11] by allowing many users to non-orthogonally share the available resources at the expense of more complicated receivers. There are many categories of NOMA according to the multiplexing method, while the most widely used one is power domain-NOMA (PD-NOMA) [12,13], in which the multiplexing is carried out by superimposing users’ information with different power levels. The PD-NOMA receiver employs successive interference cancellation (SIC) for decoding messages of higher power users before detecting its own message.

Hybrid PLC/WLC system that utilizes orthogonal multiple access (OMA) as a multiplexing technique has been studied in many research, but to the best of the authors’ knowledge, the coexistence of NOMA and hybrid PLC/WLC has not been investigated before. In this work, we provide an analytical study for investigating the performance of NOMA-based dual-hop hybrid PLC/WLC communication systems with a decode-and-forward (DF) relay. The contributions of this paper can be listed as follows: (1) Derive closed-form expressions for the outage probabilities (OPs) besides analyzing the diversity order assuming that the wireless channels are characterized by block Rayleigh fading with additive white Gaussian noise (AWGN), while the PLC channel is characterized by Log-normal distribution with Bernoulli Gaussian noise. (2) Derive approximate expressions for the ergodic capacities (ECs). (3) Validate the analysis via extensive representative Monte-Carlo simulations. (4) Study the impact of system parameters on the performance and obtain the optimal power allocation factor. (5) Compare the performance of the proposed system with OMA as a benchmark.

The rest of the paper is organized as follows. The system model is introduced in Section 2. The analytical outage probabilities and ergodic capacities are derived in Sections 3 and 4 respectively. The proposed optimal power allocation algorithm is proposed in Section 5. Analytical and simulation results are discussed in Section 6 and our conclusions are provided in Section 7.

2. System model

In this work, we consider a NOMA-based mixed dual-hop relaying system, at which a source (S) exploits NOMA to transmit information to two wireless users (far user $D_1$ and near user $D_2$) via an intermediate hybrid DF multi-modal relay node ($R$). The relay is capable of receiving over a PLC link ($h_p$) from the source, then broadcast the reconstructed message over wireless links ($h_1$ and $h_2$) as shown in Fig. 1. We assume that a perfect channel state information (CSI) is available at $R$, $D_1$ and $D_2$. The source-to-relay PLC link undergoes a log-normal fading [9] and Bernoulli Gaussian noise (BGN), while the wireless links undergo a block Rayleigh fading with AWGN noise, where the channel coefficients are constant for each transmission block but vary independently between different blocks. Following the two-phases DF relaying protocol, at the first phase, S transmits the symbol $X_s$ through the PLC channel to $R$, given that $x_s = \sqrt{\rho} h_p s_1 + \sqrt{(1-\rho)} h_p s_2$, without loss of generality the source assign the higher power allocation factor ($0.5 < \alpha < 1$) to the far user $D_1$, $S_1$ to represent the information-bearing symbol transmitted from the BS to $D_1$, with expectation $E[|s_i|^2] = 1$ for $i \in \{1, 2\}$, and $P_s$ is the source total transmit power. The received signal at $R$ can be expressed as

$$y_R = h_p e^{-\theta} d_p x_s + n_R,$$

where $\theta = a_0 + a_1 f^k$ represents the attenuation over the PLC link, $a_0$ and $a_1$ are measurement-based constants, $f$ is the operating frequency, $k$ is the attenuation factor exponent [14], and $d_p$ is the distance between $S$ and $R$. $n_R$ is a BGN with $n_R \sim CN(0, \sigma^2)$ where $CN(0, \sigma^2)$ stands for the complex Gaussian distribution having zero mean and $\sigma^2$ as variance. Following NOMA principles, $R$ decodes the strong message $S_1$ firstly while treating $S_2$ as interference, then applies SIC to decode $S_2$. Consequently, the signal-to-interference-plus-noise ratios (SINRs) for detecting $S_1$ and $S_2$ are given as follows

$$y_R^1 = \frac{\rho_s \alpha e^{-2\theta d_p |h_p|^2}}{\rho_s^2 (1-a_0) e^{-2\theta d_p |h_p|^2} + 1},$$

$$y_R^2 = \rho_s (1 - \alpha) e^{-2\theta d_p |h_p|^2},$$

where $\rho_s = \frac{P_s}{\sqrt{P}}$. During the second phase, $R$ combines the two detected symbols using a second power allocation factor, $\beta^1$, such that $0.5 < \beta < 1$, where $x_R = \sqrt{\rho} h_p S_1 + \sqrt{(1-\beta)} h_p S_2$ and $P_R$ is the relay transmit power. Then $R$ broadcasts $x_R$ over the wireless channels ($h_1$ and $h_2$) to $D_1$ and $D_2$, where the signals received at both destinations are given as follows

$$y_i^p = h_i d_i^{\alpha/2} x_R + n_i, \quad \text{for } i \in \{1, 2\}$$

where $d_i$ is the normalized distance from $R$ to $D_i$ with respect to the cell radius, $v$ is wireless channel path-loss exponent, $n_i$ is the AWGN at $D_i$ with $n_i \sim CN(0, \sigma^2)$. Both destinations follow downlink NOMA detection, where $D_1$ directly decodes its intended symbol ($S_1$) directly since it is allocated higher power. The SINR for detecting $S_1$ at $D_1$ is given as follows

$$y_{D_1}^1 = \frac{\rho_R h_i d_i^{-\alpha} |h_1|^2}{\rho_R (1-\beta) d_i^{\alpha/2} |h_1|^2 + 1},$$

\[1\] It is noteworthy that for simplicity, the same power allocation factor, $\alpha$, can be used. However, using another factor increases the degree of freedoms to improve the system performance.
Fig. 1. A dual-hop mixed powerline-wireless system.

while $D_2$ decodes $S_1$ first which is followed by using SIC to detect $S_2$. The SINRs for detecting $S_1$ and $S_2$ at $D_2$ are given as follows:

$$
\gamma_{D_2}^1 = \frac{\rho_R \beta d_2^{-\epsilon} |h_2|^2}{\rho_R (1-\beta) d_2^{-\epsilon} |h_2|^2 + 1},
$$

$$
\gamma_{D_2}^2 = \rho_R (1-\beta) d_2^{-\epsilon} |h_2|^2,
$$

where $\rho_R = \frac{P_g}{\sigma^2}$.

3. Outage probability analysis

In this section, we derive the outage probabilities at both destinations and the system OP using (2), (4), and (5), which are given as follows:

$$
O_{P_1} = 1 - Pr(\gamma_R^1 > \pi_1, \gamma_{D_1}^1 > \pi_1) \equiv 1 - P_1(\pi_1) P_2(\pi_1),
$$

$$
O_{P_2} = 1 - Pr(\gamma_R^2 > \pi_1, \gamma_{D_1}^2 > \pi_1, \gamma_{D_2}^2 > \pi_2),
$$

$$
O_{P_{sys}} = 1 - Pr(\gamma_R^1 > \pi_1, \gamma_{D_1}^1 > \pi_1, \gamma_{D_2}^2 > \pi_2),
$$

where $\pi_i = 2^{R_i} - 1$ is the threshold SNR, $R_i$ is the normalized threshold rate for $D_i$, $(a),(b)$, and $(c)$ stem from the decorrelation of the random variables, and

$$
P_1(\pi_1) = Pr(\gamma_R^1 > \pi_1),
$$

$$
P_2(\pi_1, \pi_2) = Pr(\gamma_{D_1}^1 > \pi_1, \gamma_{D_2}^2 > \pi_2),
$$

$$
P_3(\pi_1, \pi_2) = Pr(\gamma_{D_2}^2 > \pi_2),
$$

$$
P_4(\pi_1, \pi_2) = Pr(\gamma_R^1 > \pi_1, \gamma_{D_2}^2 > \pi_2).
$$

In the following, we introduce the channel statistics of the involved channels and derive the closed-form expressions of $P_1$, $P_2$, $P_3$, and $P_4$, respectively.

**Channels’ Statistics:** The amplitude of the PLC channel ($h_P$) is distributed according to log-normal distribution, $h_P \sim LN(\mu_{h_P}, \sigma_{h_P}^2)$, where $\mu_{h_P}$ is the mean and $\sigma_{h_P}^2$ is the variance. Consequently, the channel gain, $|h_P|^2$, follows a log-normal distribution, $|h_P|^2 \sim LN(\mu', \sigma'^2)$, where $\mu' = 2\mu_{h_P}$ and $\sigma'^2 = 4\sigma_{h_P}^2$ [9]. The cumulative distribution function of the channel gain, $F_{|h_P|^2}(x)$, is given as follows:

$$
F_{|h_P|^2}(x) = Q\left(\frac{\mu'(\ln(x))}{\sigma'}\right) = 1 - Q\left(\frac{\ln(x) - \mu'}{\sigma'}\right),
$$

where $Q(.)$ is the Gaussian Q-function [15]. On the other hand, the two wireless links, $h_i, i \in \{1, 2\}$, undergo block Rayleigh fading distribution, which means that the channel gains, $|h_i|^2$, follow exponential distributions, $|h_i|^2 \sim \text{exp}(1)$, whose CDF is given as follows:

$$
F_{|h_i|^2}(x) = 1 - e^{-x}.
$$

**Derivation of $P_1$:** Using (2), (7), and (8), $P_1$ can be expressed as

$$
P_1(\pi_1) \equiv \left\{ \begin{array}{ll}
Q\left(\frac{\ln(\pi_1^{2\rho/d_2}) - \mu'}{\sigma'}\right), & \alpha > L_1 \\
0, & \alpha < L_1,
\end{array} \right.
$$

where $L_1 = \pi_1/(1+\pi_1)$, $\epsilon = \alpha(1+\pi_1) - \pi_1$ and $(d)$ stems from using the CDF of $|h_P|^2$ in (8).

**Derivation of $P_2$:** Using (4), (7), and (9), $P_2$ can be expressed as

$$
P_2(\pi_1, \pi_2) \equiv \left\{ \begin{array}{ll}
P_{31} = e^{-\frac{\pi_1^{2\rho/d_2}}{\rho_R}}, & \beta > L_1 \\
P_{32} = e^{-\frac{\pi_1^{2\rho/d_2}}{\rho_R}}, & \beta < L_1 \\
0, & \text{Otherwise},
\end{array} \right.
$$

where $\kappa = \beta(1+\pi_1) - \pi_1$ and $(e)$ stems from using the CDF of $|h_P|^2$ in (9).

**Derivation of $P_3$:** Using (5), (7), and (9), $P_3$ can be expressed as

$$
P_3(\pi_1, \pi_2) \equiv \left\{ \begin{array}{ll}
P_{41} = Q\left(\frac{\ln(\pi_1^{2\rho/d_2}) - \mu'}{\sigma'}\right), & \alpha > L_2 \\
P_{42} = Q\left(\frac{\ln(\pi_1^{2\rho/d_2}) - \mu'}{\sigma'}\right), & \alpha < L_2 \\
0, & \text{Otherwise},
\end{array} \right.
$$

where $(g)$ stems from using the CDF of $|h_P|^2$ in (8). By substituting (10), (11), (12), and (13) into (6), we obtain closed form expressions for $O_{P_1}$, $O_{P_2}$, and $O_{P_{sys}}$.

**Diversity Order:** To obtain deep insights regarding our system, we derive the diversity order of the outage probabilities in (6), where the diversity order is defined as the slope of those curves. Using [16,17], we can calculate diversity orders as $d_{\alpha}^m = -\lim_{\mu \to -\infty} (\log(O_{P_{\infty}})/\log(\mu))$, where $m \in \{1, 2\}$ and $O_{P_{\infty}} = O_{P_{sys}}$ when $\mu$ tends to $\infty$, which can be calculated by finding $P_{\infty}$ for $i \in \{1, 2, 3, 4\}$ as follows:

$$
P_{1,\infty} = \begin{cases} 
\frac{\sigma^2 \epsilon}{\sqrt{2\pi^3(\ln(\pi_1^{2\rho/d_2}) - \mu')^2}}, & \alpha > L_1 \\
0, & \text{Otherwise}
\end{cases}
$$
be formulated as

\[ P_2 = \begin{cases} 1 - \frac{\pi d^2}{\rho k}, & \beta > L_1 \\ 0, & \text{Otherwise} \end{cases} \]  

(15)

\[ P_3 = \begin{cases} 1 - \frac{\pi d^2}{\rho k}, & L_1 < \beta < L_2 \\ 0, & \text{Otherwise} \end{cases} \]  

(16)

\[ P_4 = \begin{cases} \frac{\alpha}{\sqrt{2}} \left( \frac{\mu - \pi d^2}{\rho k} \right)^2, & \beta > L_2 \\ \frac{\alpha}{\sqrt{2}} \left( \frac{\mu - \pi d^2}{\rho k} \right)^2, & \alpha > L_2 \\ 0, & \text{Otherwise} \end{cases} \]  

(17)

Taking into consideration (14), (15), (16), and (17), and some simple mathematical expressions, we can express the plots in Fig. 3 at Section 6.

4. Ergodic capacity analysis

In this section, we investigate the ergodic capacity of the considered system. The instantaneous capacities for both users are given by [18] as

\[ C_1 = \frac{1}{2} \log(1 + \min\{\gamma^1_R, \gamma^1_{D1}\}) \]

\[ C_2 = \frac{1}{2} \log(1 + \min\{\gamma^2_R, \gamma^2_{D2}\}) \]  

(18)

Consequently, the ergodic sum capacity of the system can be formulated as \( EC = EC_{S1} + EC_{S2} \), where \( EC_i \) for \( i \in \{1, 2\} \) is the expectation of the channel capacity as follows [18]

\[ EC_{S1} = \frac{1}{2 \ln 2} \int_{\pi = 0}^{\pi = \pi} \frac{1}{1 + \pi} \left[ 1 - F_{\gamma_i}(\pi) \right] d\pi. \]  

(19)

**Ergodic Capacity of D1:** The CDF \( F_{\gamma_i}(\pi) \) can be written as

\[ F_{\gamma_i}(\pi) = 1 - pr(\gamma^i_R > \pi, \gamma^i_{D1} > \pi, \gamma^i_{D2} > \pi) \]

(20)

where \( \gamma \) stems from the decorrelation of the random variables, and \( P_3(\pi) = pr(\gamma_{D2}^2 > \pi) \), which is reformulated as

\[ P_3(\pi) = \begin{cases} e^{-\frac{\gamma^2_{D2}}{\rho k}}, & \beta > L_3 \\ 0, & \text{otherwise} \end{cases} \]  

(21)

By substituting from (10), (11), and (21) into (19), \( EC_{S1} \) can be formulated as

\[ EC_{S1} = \frac{1}{2 \ln 2} \int_{\pi = 0}^{\pi = \pi} e^{-\frac{\gamma^2_{D2} + \gamma^2_{D2}}{\rho k}} \left( \frac{\ln(\frac{\gamma^2_{D2} + \gamma^2_{D2}}{\rho k}) - \mu}{\sigma} \right) d\pi. \]  

(22)

However, to the best of our knowledge, there is no closed-form expression for such intractable integration. In the following, we provide an approximation using Gauss–Laguerre quadrature [19] by introducing the following variable transformation, where \( \tau = \pi (d_1^2 + d_2^2)/\rho k \). Consequently, we can express \( EC_{S1} \) as

\[ EC_{S1} = \frac{1}{2 \ln 2} \int_{\tau = 0}^{\infty} e^{-\tau} f_1(\tau) d\tau, \]  

(23)

where,

\[ f_1(\tau) = \alpha e^\tau Q(\frac{\ln(\tau + \frac{\tau}{\rho k}) - \mu}{\sigma} - \mu). \]  

(24)

Using [[19], eq. (25.4.45)], we can express the approximated value as

\[ EC_{S1} = \frac{1}{2 \ln 2} \sum_{i=1}^{n} w_i f_1(\tau_i) \]  

(25)

where \( \tau_i \) is the \( i \)th zero of the Laguerre polynomial, \( L_m(\tau) \), while \( w_i \) is the weights that can be calculated as follows:

\[ w_i = \frac{(n!)^2}{(n + 1)^2 [L_{n+1}(\tau_i)]^2} \]  

(26)

**Ergodic Capacity of D2:** Similar to \( EC_{S1} \), the CDF of \( \gamma_2 \) can be expressed as

\[ F_{\gamma_2}(\pi) = 1 - pr(\gamma^2_R > \pi, \gamma^2_{D2} > \pi) = 1 - P_3(\pi) P_3(\pi) \]  

(27)

where

\[ P_3 = pr(\gamma_{D2}^2 > \pi) = e^{-\frac{\tau^2}{\rho k}} \]  

By Substituting from (27) into (19) then

\[ EC_{S2} = \frac{1}{2 \ln 2} \int_{\pi = 0}^{\pi = \pi} \frac{e^{-\frac{2 \tau^2}{\rho k}}}{1 + \pi} Q \left( \frac{\ln(\frac{2 \tau^2}{\rho k} - \mu)}{\sigma} \right) d\pi. \]  

(29)

Eq. (29) can be approximated similar to (23) as [[19], eq. (25.4.45)]

\[ EC_{S2} = \frac{1}{2 \ln 2} \sum_{i=1}^{n} w_i f_2(\tau_i) \]  

(30)

where,

\[ f_2(\tau) = \frac{\tau^2}{1 + \pi} Q \left( \frac{\ln(\frac{2 \tau^2}{\rho k} - \mu)}{\sigma} \right) \]  

(31)

5. Proposed power allocation algorithm

In this section, we discuss a proposed power allocation algorithm for optimizing the outage probability of the system. The proposed optimization problem is given as follows:

\[ \min_{\alpha, \beta} \quad OP_{\text{sys}} \quad \text{(32a)} \]

s.t. \( 0.5 < \alpha < 1 \) \quad \text{(32b)}

\( 0.5 < \beta < 1 \) \quad \text{(32c)}

In the following, we provide Theorem 1 to prove the convexity of the problem.

**Theorem 1.** The problem in (32) is a convex optimization problem.
Table 1

<table>
<thead>
<tr>
<th>Parameter</th>
<th>(\sigma') (dB)</th>
<th>(\mu')</th>
<th>(\alpha) or (\beta)</th>
<th>(\pi_i)</th>
<th>(\rho) (dB)</th>
<th>(f) (kHz)</th>
<th>(a_0)</th>
<th>(a_1)</th>
<th>(k)</th>
<th>(d_1)</th>
<th>(d_2)</th>
<th>(d_p)</th>
<th>(v)</th>
</tr>
</thead>
<tbody>
<tr>
<td>value used</td>
<td>2</td>
<td>1</td>
<td>0.5 : 0.99</td>
<td>1 : 2</td>
<td>0 : 30</td>
<td>500</td>
<td>2.03e - 3</td>
<td>3.75e - 7</td>
<td>0.7</td>
<td>2</td>
<td>1</td>
<td>10</td>
<td>3</td>
</tr>
</tbody>
</table>

Fig. 2. The system OP versus power allocation factors (\(\alpha\) and \(\beta\)) with \(\rho = 30\) dB, \(\pi_1 = 1\), and \(\pi_2 = 2\).

**Proof.** Please, refer to Appendix.

Since (32) is a convex optimization problem, it can be solved using any commercial solver such as Matlab or Mathematica. With the system parameters setting defined in Table 1, we show the variations of the system OP as a function of both \(\alpha\) and \(\beta\) in Fig. 2, to show the convexity graphically. The solution of such problem gives an optimal value for power allocation factors as \(\alpha = 0.6\) and \(\beta = 0.83\). This solution is used as an optimal power allocation scheme in all figures of outage in Section 6.1.

6. Discussions and results

In this section, we present representative numerical results to demonstrate the behavior of the system and to validate the derived expressions. Unless mentioned otherwise, the simulation parameters used in the plots are given in Table 1, which depend on [9,20,21]. The notation “Ana” is used for the exact analytical results of the OP and the approximated EC, and “Sim” denotes Monte-Carlo simulation results in all legends. We have assumed that \(\rho_s = \rho_R = \rho\).

6.1. Outage probability results

Fig. 3 shows the variations of the OP at \(D_1\), \(D_2\), and \(OP_{sys}\) versus SNR \(\rho\) assuming \(R_1 = 0.5\) bits/s/Hz and \(R_2 = 0.75\) bits/s/Hz. The optimal values of the power allocation factors obtained from Section 5; \(\alpha = 0.6\), and \(\beta = 0.83\); have been used. The result shows that the exact and the simulation result coincide for the entire span of \(\rho\) which validates the analysis. Additionally, we note an improvement in OPs performance as \(\rho\) increases. To clarify the advantages of the proposed system in this work; the NOMA-based mixed dual-hop relaying system; against a well-known reference system, we compared our system with an OMA-based mixed dual-hop relaying system. Through the comparison held, it becomes clear to us the superiority of the proposed system due to the spectrum efficiency of NOMA. Since \(OP_{sys} = 0.034\) at \(\rho = 30\) dB and \(OP_{sys} = 0.0034\) at \(\rho = 40\) dB, so the diversity order can be calculated as \(\log(0.034) - \log(0.0034) \approx 1\).

Under the same settings, Fig. 4 shows the effect of changing distance between end-users and relay on the OPs at both users and the system OP. The results show that the OPs degrade with the increase in distance due to the corresponding decrease in the received SINR.

Fig. 5 shows the variations of the OPs as functions of the power allocation factor in the range from 0.5 to 0.99 assuming \(\alpha = \beta, \rho = 30\) dB, \(\pi_1 = 1\), and \(\pi_2 = 2\). The results show the improvement in OP at \(D_1\) with increasing \(\alpha\) as this leads to an increase in the power of its own message and in the same time reducing interference power from \(D_2\). The situation of OP at \(D_2\) witnesses a great reduction at the beginning until \(\alpha\) reaches a certain value at which the curve inflected and its OP starts to
Fig. 5. OP versus power allocation factor ($\alpha$ or $\beta$) with $\rho = 30$ dB, $\pi_1 = 1$, and $\pi_2 = 2$.

6.2. Ergodic capacity results

Fig. 6 shows the EC for both users and the ESC against the transmit SNR. It is noteworthy that $EC_S_1$ increases gradually at low-SNR region and saturates at the high-SNR due to the interference of the second user. On the other hand, $EC_S_2$ increases with SNR with no saturation since we assume a perfect SIC scenario. Although the performance of EC for each message is different, the performance of ESC follows the trend of $EC_S_2$. Additionally, we compare the performance of the proposed NOMA-based system with an OMA-based system under the same parameter settings. The results show that the ESC of the NOMA-based system is better compared with the OMA-based one.

To present the effect of changing distance between end-users and relay on the ECs, Fig. 7 shows the ECs under different settings of the distances $d_1$ and $d_2$. From the figure, it is clear that the capacities degrade while increasing the distance, also this is due to the decrease in the received SINR with the increase in distance.

7. Conclusion

Through this paper, we proposed a novel two-user NOMA-based dual-hop hybrid PLC/WLC system that uses a DF relay. The closed-form formulas for the $OPs$ of both users and the system OP, the $ECs$ for the two messages, the ESC, and the diversity order for the OPs were derived. Monte-Carlo simulations were used after that to confirm the accuracy of driven forms. Also, we investigated the effect of varying the distance between the end-users and the relay on the system performance. Also, we studied the impact of the power allocation factors on the system’s OP performance and the existence of an optimal value under constant SNR. Also, we held a comparison between the proposed system in contrast with a well-known benchmark system to illustrate that our system can add a good contribution to the communication practical systems. As future research directions, we may find the optimal values of power allocation factors analytically.

CRediT authorship contribution statement

Basem M. El Halawany: Conception and design of study, Drafting the manuscript, Revising the manuscript critically for important intellectual content, Analysis and/or interpretation of data. Ahmed Samir: Conception and design of study, Drafting the manuscript, Analysis and/or interpretation of data. Mohamed Elsayed: Conception and design of study, Analysis and/or interpretation of data, Drafting the manuscript. Wali Ullah Khan: Conception and design of study, Acquisition of data, Revising the manuscript critically for important intellectual content. Kaishun Wu: Conception and design of study, Revising the manuscript critically for important intellectual content. Ehab Mahmoud Mohamed: Conception and design of study, Acquisition of data, Revising the manuscript critically for important intellectual content.
Declarations of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

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All authors approved the final version of the manuscript.

Appendix

In this appendix, a detailed derivation that proves the convexity of the problem is described in (32). The first step is to re-write the system OP as $\mathbf{O}_\mathbf{P}_{sys} = 1 - I_{sys}$ for different intervals of $\alpha$ and $\beta$, where

$$I_{sys} = \begin{cases} l_1 = P_3 P_4, & L_1 \triangleleft \alpha \triangleleft L_2, L_1 < \beta < L_2 \\ l_2 = P_3 P_4, & \alpha > L_2, L_1 < \beta < L_2 \\ l_3 = P_3 P_4, & L_1 < \alpha < L_2, \beta > L_2 \\ l_4 = P_3 P_2 P_3, & \alpha > L_2, \beta > L_2 \\ 0, & \text{otherwise}. \end{cases}$$

(A.33)

To show the convexity, we need to show that the Hessian matrix ($H$) is positive semi-definite, where $H_1$ under the first region $L_1 < \alpha < L_2, L_1 < \beta < L_2$ is given as

$$H = \begin{bmatrix} \Delta_1 & \Delta_2 \\ \Delta_3 & \Delta_4 \end{bmatrix} = \begin{bmatrix} -\frac{\partial^2 I_2}{\partial \alpha^2} & -\frac{\partial^2 I_2}{\partial \beta^2} \\ -\frac{\partial^2 I_2}{\partial \alpha \partial \beta} & -\frac{\partial^2 I_2}{\partial \alpha^2} \end{bmatrix}.$$  

(A.34)

Finding those derivatives we reach

$$\Delta_1 = \frac{1 + \pi_1}{e_2^2\sigma^2}\left(\frac{c_3 + \mu}{\sigma^2} - \ln \frac{c_3 + \mu}{\sigma^2}\right)^2, \quad \Delta_2 = -\frac{c_1 + c_2 + (1 + \pi_1)}{e_2\sigma^2}\left(\frac{c_3 + \mu}{\sigma^2} - \ln \frac{c_3 + \mu}{\sigma^2}\right)^2, \quad \Delta_3 = -\frac{\sigma^2 (1 + \pi_1)^2 (c_1 + c_2) (1 - \frac{2\sigma}{c_1 + c_2}) e^{-\frac{c_3 + \mu}{\sigma^2}}}{\kappa^4 (\ln \frac{c_1 + c_2}{\sigma^2} - \mu)^2}, \quad \Delta_4 = \frac{\kappa^4 (\ln \frac{c_1 + c_2}{\sigma^2} - \mu)^2}{\sigma^2 (1 + \pi_1)^2 (c_1 + c_2) (1 - \frac{2\sigma}{c_1 + c_2}) e^{-\frac{c_3 + \mu}{\sigma^2}}}.$$  

(A.35a)  

where $c_1 = \pi_1 d_1^2 / \rho_R$, $c_2 = \pi_1 d_2^2 / \rho_R$, and $c_3 = \pi_1 e^{2d_3 d_4} \rho_S$. As we can see that $\Delta_1$: which is the first leading minor; is positive, and after some mathematical manipulation we can prove that $\Delta_1 - \Delta_2 - \Delta_3$: which is the second leading minor; is positive as well, which demonstrates the convexity of the proposed optimization problem. We can follow the same steps to prove the convexity over the other regions as well.

References


