Capacity Analysis of Wireless Powered Cooperative NOMA Networks over Generalized Fading

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Abstract—This paper provides a performance evaluation of two-hop non-orthogonal multiple access (NOMA) architecture with energy harvesting (EH) cooperative agent and channel gains following the k-m fading, through analysis of ergodic capacity with respect to hardware impairments and channel conditions. The performance results of the distant user over two EH protocols, namely power splitting and time-switching relaying, are obtained and compared with simulation outcomes. The developed framework allows to evaluate the network under a range of external conditions and infers the importance of considering the hardware impairments.

Index Terms—Cooperative communications, ergodic capacity, simultaneous wireless information and power transfer (SWIPT), non-orthogonal multiple access (NOMA).

I. INTRODUCTION

The trends for high data rates, low latency and reliable communication have been raised over the last couple of years. According to the Cisco’s recent report [1], the amount of traffic generated by wirelessly connected devices will increase to unprecedented rates, while the issue of bandwidth shortage is still present. A promising technique to enable such an architecture and provide spectral efficiency is considered to be non-orthogonal multiple access (NOMA) [2]. In contrast to its counterparts, NOMA allows us to use the same resource of wireless medium, i.e., time, frequency or code [3]. The research conducted on NOMA highlighted its advantages to enhance the overall network performance at the physical layer [4]. The authors in [5]–[7] studied the architecture where a cooperative NOMA network is considered.

In cooperative networks, the resources of cooperative agents are shared to create multiple replicas of the transmitted signal at the destination node [8]. Such a network architecture is widely used in applications, where the destination node’s signal is subject to significant fading and the communicating nodes are equipped with single-antennas due to mobility, size and other constraints. Often, such applications ranging from Internet-of-Things to biomedical sensors experience energy limitations as well. This issue can be resolved by implementing wireless energy transfer methods.

One of the candidates to solve the energy limitation is simultaneous wireless information and power transfer (SWIPT). Although, it has not been implemented practically, the core idea of SWIPT was introduced in [9]. Two practical approaches for the energy harvesting (EH), namely, time-switching relaying (TSR) and power-splitting relaying (PSR) protocols, were introduced in [10], which became widely used approaches among the researchers. The PSR and TSR protocols operate through partitioning of the resources in the power and time domains, respectively. In particular, the authors in [11] provided some results on the performance estimation of a two-hop cooperative NOMA network architecture over Weibull channels. However, more generalized fading such as an \( \alpha - \mu \) statistical model was studied in [12]–[14] while omitting the impact of hardware imperfections. Due to the work in [15], the complementary metal–oxide–semiconductor (CMOS) technology enabled hardware operating with high-frequency signals, is subject to the imperfections. Thus, the detrimental effect of the hardware impairments is a significant issue to consider in the performance evaluation of wireless networks [16].

Hence, this paper analyzes the performance of a two-hop NOMA network, where the cooperative agent harvests the energy and relays the message in an amplify-and-forward (AF) mode. Moreover, the considered network incorporates the effect of hardware imperfections at the transceivers’ radio frequency (RF) front-ends. The expressions of the ergodic capacity are derived for the abovementioned network. Using this performance metric, the effects of EH protocols and hardware impairments are evaluated over a wide range of channel conditions represented via a \( \kappa - \mu \) general fading model.

II. SYSTEM AND CHANNEL MODEL

We consider the cooperative NOMA network comprising a source (S) and two receiver nodes, i.e., \( U_1 \) and \( U_2 \), (see Fig. 1), where \( U_1 \) operating in the AF relaying mode works as a cooperative agent. The network is a constellation of downlink half-duplex operated nodes with only one antenna for all purposes. Following the NOMA technique, S broadcasts a composite message \( x_s = \sum_{i=1}^{2} \sqrt{a_i P_S x_i} \), where \( i, x_i \) and \( a_i \) refer to the respective user, its message and power allocation factor, accordingly. For the analysis, S transmits with a constant power of \( P_S \) and the wireless channel conditions are assumed to be as \( h_1 > h_2 \) due to their relative proximity and other factors. Therefore, following the NOMA approach, the relative power allocation factors are chosen as \( a_1 < a_2 \).
TABLE I. The variables in Eq. (7) for the PSR and TSR protocols.

<table>
<thead>
<tr>
<th>Variable</th>
<th>$n_p$</th>
<th>$H_p$</th>
<th>$R_p$</th>
<th>$F_p^R$</th>
<th>$G_p$</th>
</tr>
</thead>
<tbody>
<tr>
<td>PSR</td>
<td>$\sqrt{1 - \rho n_1 + n_c}$</td>
<td>1/2</td>
<td>0.5$\varpi \rho T$</td>
<td>$\varpi \rho P_S d_1^{-\tau} (1 + \Omega_{s,u_1}^2)$</td>
<td>$\sqrt{\varpi \rho}$</td>
</tr>
<tr>
<td>TSR</td>
<td>$n_1$</td>
<td>$(1 - \eta)/2$</td>
<td>$\varpi \eta T$</td>
<td>$\frac{2\varpi \rho P_S (1 + \Omega_{s,u_1}^2)}{(1 - \eta) d_1^{-\tau}}</td>
<td>h_1</td>
</tr>
</tbody>
</table>

Due to the assumption that $U_1$ has better channel conditions, it acts as a cooperative agent to further amplify and forward the received message from $S$ to $U_2$. Thus, creating spatial diversity at the receiver side of $U_2$, which is considered as the main destination. Another assumption is that $U_1$ needs to harvest energy from the incoming electromagnetic signal to be able to relay the message.

The distances related to the source-to-relay (S-to-$U_1$), relay-to-destination ($U_1$-to-$U_2$) and source-to-destination ($S$-to-$U_2$) links are denoted by $d_1$, $d_2$, and $d_3$, respectively. The corresponding channel fluctuations, denoted by $h_1$, $h_2$ and $h_3$, follow the $\kappa$-$\mu$ fading which is a generalized channel model able to emulate the Rayleigh ($\kappa \rightarrow 0$ and $\mu = 1$), Rice ($\kappa = k$ and $\mu = 1$), etc. [17], [18]. The probability density function (PDF) of the $\kappa$-$\mu$ distribution is represented as follows

$$f_h(r) = \frac{2\mu_i (1 + \kappa_i)}{e^{\mu_i \kappa_i} \kappa_i^{\frac{\mu_i + 1}{\kappa_i}} r^{\mu_i} e^{-\frac{\mu_i}{\kappa_i}} I_{\nu}(2\lambda_i r),$$

where $\kappa > 0$ is the ratio of line-of-sight and scattered signal wave powers, $\phi_i = \mu_i (1 + \kappa_i)$ and $\lambda_i = \mu_i \sqrt{\kappa_i} (1 + \kappa_i)$. The $I_{\nu}(\cdot)$ is the modified Bessel function of first kind and order $\nu$.

III. ENERGY HARVESTING PROTOCOLS

Due to the broadcasting nature of communication, both $U_1$ and $U_2$ receive the superimposed NOMA signal directly as

$$y_i = \sqrt{\frac{P_S}{d_{s,u_1}^2}} h_{s,u_1} \left( \sum_{j=1}^{\nu} a_{j} x_j + \eta_{h_{s,u_1}} \right) + n_i,$$

where $\gamma$ is the path-loss coefficient, $n_i$ is the additive white Gaussian noise (AWGN) seen at the receiver node with the mean and variance being equal to zero and $\sigma_i^2$, respectively. Moreover, the combined hardware impairment level, denoted by $\eta_{h_{s,u_1}}$, follows the Complex Gaussian distribution with zero mean and variance of $\Omega_{s,u_1}^2$. According to [15], it is defined as $\Omega_{s,u_1} \triangleq \sqrt{\Omega_{s,t}^2 + \Omega_{h_{s,u_1}}^2}$, considering the RF front-end imperfections of both transmitter and receiver. Moreover, $U_2$ receives the NOMA signal through direct path, and decodes message $x_2$ using the signal-to-interference-plus-distortion-plus-noise ratio (SINDR) of

$$\gamma_{U_2}^{[x_2]} = \frac{e_1 Z}{e_2 Z + e_3},$$

where $Z = |h_3|^2$, $e_1 = a_2$, $e_2 = a_1 + \Omega_{s-u_2}^2$, and $e_3 = \sigma_{n_c}^2 d_1^2$. The whole transmission time block necessary for the message to reach its destination is denoted by $T$. In the PSR, $T$ is divided into 2 equal portions. The first time slot is dedicated to the $S$-to-$U_1$ transmission, where $U_1$ harvests the $\rho P_S$ portion of power for further use and $(1 - \rho) P_S$ portion of power is used for the information detection purpose. During the second time slot, $U_1$ uses the harvested energy to amplify and then forward the second user’s signal. On the other hand, for the TSR case, $\eta T$ portion of the time ($0 \leq \eta \leq 1$) is dedicated for the energy harvesting purposes and the rest of the time is split equally between the $S$-to-$U_1$ and $U_1$-to-$U_2$ transmissions.

For the PSR, the portion of the signal dedicated to the energy harvesting is written as

$$\sqrt{\rho y_1} = \sqrt{\frac{\rho P_S}{d_1^2}} h_1 \left( \sum_{j=1}^{\nu} a_{j} x_j + \eta_{h_{s,u_1}} \right) + \sqrt{\rho n_1}.$$

From this, the energy that can be harvested is further specified as

$$E_p^H = R_p(P_S|h_1|^2 (1 + \Omega_{s,u_1}^2) d_1^{-\tau} + \sigma_{n_c}^2)$$

where $0 \leq \varpi \leq 1$ is the harvesting efficiency and $p$ denotes the association of the variable either to the PSR or TSR protocol (See Table I), respectively. Furthermore, disregarding the little amount of energy harvested from the noise term, Eq. (5) can be simplified and the expression of power for the next time slots is deduced as $F_p^R = \frac{E_p^H}{R_p T}$.

In the PSR protocol, the information detection is performed using a portion of the incoming signal, $y_1^T = \sqrt{1 - \rho y_1 + n_c}$, where $n_c$ is the AWGN at the information receiver.

Following the NOMA principle, $U_1$ is capable of performing successive interference cancellation (SIC) by first decoding the stronger message $x_2$, and then its own message $x_1$ using
signal-to-distortion-plus-noise ratio (SDNR) given by
\[ \gamma_{U_1} = \frac{b_1 X}{b_2 X + b_3}, \]
where \( X = |h_1|^2 \) and the rest parameters for both protocols are presented in Table II.

During the final transmission phase, \( U_1 \) uses the harvested energy and spends it on the amplification and forwarding the message towards \( U_2 \). The signal seen at \( U_2 \) is written as
\[ y_{1\rightarrow2} = \frac{G_p h_2}{\sqrt{d_2}} (D_p (y_1 - n_1) + n_p + \eta_{u_1, u_2}) + n_2, \]
where \( D_p \) is equal to \( \sqrt{1 - \rho} \) for the PSR and to 1 for the TSR, respectively. \( n_2 \) is the AWGN introduced at \( U_2 \) and \( G_p \) is the amplification factor. Assuming the high signal-to-noise (SNR) values, the amplification factor can be simplified as \( G_p = \sqrt{\frac{P_{\text{tx}}}{W_{\text{RX}}}} \), where \( W_{\text{px PSR}} = (1 - \rho) \left( P_s (1 + \Omega_{u_1}) d_1^{-\gamma} |h|^{2} + \sigma_1^2 \right) + \sigma_2^2 \) and \( W_{\text{px TSR}} = \frac{d_1^{-\gamma} |h_1|^2}{\mu} + \sigma_1^2 \). Further, substituting the obtained gain expressions into Eq. (7), the SINDR for both protocols to decode \( x_2 \) is derived as
\[ \gamma_{U_1\rightarrow2} = \frac{c_1 X Y}{c_2 X Y + c_3 Y + c_4}, \]
where \( Y = |h_2|^2 \) and \( c_1, c_2, c_3 \) and \( c_4 \) are shown in Table II.

### IV. Ergodic Capacity

**Proposition 1:** For the S-to-\( U_2 \) direct link, the ergodic capacity expression is defined as in Eq. (9) for both protocols, at the top of the next page, where \( E = c_1 + c_2 \) and \( \alpha_3 = \mu_3 + g \).

**Proof:** See Appendix A.

Due to its complexity, the ergodic capacity evaluation for the exact SINDR expression of the \( U_1 \)-to-\( U_2 \) link is not feasible. Thus, in the high-SNR regime (i.e., \( c_3 \to 0 \)), Eq. (8) can be approximated as
\[ \gamma_{U_1\rightarrow2} \approx \frac{c_1 X Y}{c_2 X Y + c_3 Y + c_4}. \]

**Proposition 2:** For the \( U_1 \)-to-\( U_2 \) link, the ergodic capacity expression is defined as in Eq. (10) for both protocols, where \( c_5 = c_1 + c_2 \), \( \beta_1 = \frac{\mu \sigma_2^2}{\mu_1 + \mu_2 + \mu_3} \) and \( \beta_2 = \frac{\mu_2 \sigma_2^2}{\mu_1 + \mu_2 + \mu_3} \).

**Proof:** See Appendix B.

Additionally, the asymptotic characteristic of the \( U_1 \)-to-\( U_2 \) link capacity is evaluated to be explicit. As a result, the SINDR at \( U_2 \) is changed to
\[ \gamma_{U_1\rightarrow2} \approx \frac{c_1 X}{c_2 X + c_3}, \]
accounting for \( c_4 \to 0 \). Due to the similarity of the general form of asymptotic SINDR with the previously mentioned case of direct transmission, the explanations are omitted here. The asymptotic expression defining the average capacity for the \( U_1 \)-to-\( U_2 \) link is shown in Eq. (11), where \( \alpha_1 = \mu_1 + q \). Then, the overall ergodic capacity at \( U_2 \) over the whole transmission period is expressed as
\[ C_{\text{erg}} = C_{\text{erg}}^{\gamma_1\rightarrow2} + C_{\text{erg}}^{\gamma_2}. \]

### V. Numerical Results

This section describes the obtained results for the ergodic capacity expressions and their respective comparison with the Monte-Carlo simulations. For this analysis, the distances are set as \( d_1 = 2 \text{ m} \), \( d_2 = 1.3 \text{ m} \), and \( d_3 = 3 \text{ m} \). In addition, we set \( \sigma = 0.6, \tau = 3, \alpha_1 = 0.2 \) and \( \alpha_2 = 0.8, [19], [20] \).

Fig. 2 illustrates the change of ergodic capacity due to different channel conditions for both protocols, while assuming the perfect hardware impairments scenario. For the whole range of SNR values and respective conditions, the
Fig. 3. The ergodic capacity for the TSR ($\eta = 0.2$) and PSR ($\rho = 0.7$) protocols using NOMA and OMA techniques, when $\Omega = \{0, 0.3\}$, $\kappa = 3$ and $\mu = 1$.

PSR protocol shows a better performance compared to the TSR. Increasing the number of multipath clusters ($\mu$) improves the overall performance of both protocols. The saturation for both protocols can be explained by the NOMA’s constraints. Overall, the approximation results follow closely the exact values for both TSR and PSR protocols, while a slightly better correlation is observed for the TSR case.

Similarly, Fig. 3 shows the effect of different hardware impairment values on the average capacity over Rician channels. In all of the cases below, the Rician $K$ parameter signifying the ratio of in-phase and quadrature-phase signals is defined to be 3. For both protocols, a significant degradation of the capacity performance is observed due to the increase in hardware impairments. For example, in the PSR, the average capacity at 50 dB is around 2.3 bits/s/Hz for the case with no hardware impairments, while it approaches 1.87 bits/s/Hz, when impairments are introduced. Additionally, for comparison with the OMA benchmark, the capacity plots are shown, where the saturation starts after 40 dB due to the hardware impairments. In essence, all NOMA cases start saturating at 30 dB under any hardware impairments level.

Fig. 4 presents the simulation results for the performance change under a set of energy harvesting factors $\rho$ and $\eta$ at 50 dB. As in the previous cases, the metric shows better results under the PSR rather than the TSR. For both PSR and TSR cases, the effect of hardware impairments is seen as more drastic, when compared with the change of $\mu$ values. When the portion of the power allocated for the energy harvesting is small or too much, the performance degrades. The optimal energy harvesting factors for the TSR and PSR are defined as 0.2 and 0.9, respectively. These values hold true when the external factors affecting the network are changed.
VI. CONCLUSION

This paper analyses the ergodic capacity metric of the distant user in a cooperative two-hop NOMA network, while considering the energy-constrained cooperative agent with hardware impairments and channel gains following the $\kappa$-$\mu$ fading model. The approximate and asymptotic expressions were derived and examined over the PSR and TSR protocols. The obtained results demonstrated the importance of the hardware impairments, as they tend to significantly affect the overall performance. The accomplished work serves as a framework to foster further research in the area, as it contributes by delivering a general expression for the performance evaluation under a set of channel statistics and transceiver hardware quality.

VII. ACKNOWLEDGMENT

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APPENDIX A
DERIVATION OF PROPOSITION 1

For the $S$-to-$U_2$ direct link, the ergodic capacity expression is defined the following way

$$ C_{\text{erg}}^{\gamma_{U_2}} = \int_0^\infty H_p \log_2 (1 + \gamma_{U_2}) f_{\gamma_{U_2}} d\gamma_{U_2} $$

$$ = \int_0^\infty H_p \log_2 (e_3 + (e_1 + e_2) Z) f_Q(q) dq $$

$$ - \int_0^\infty H_p \log_2 (e_3 + e_2 Z) f_Q(q) dq $$

$$ = \frac{H_p}{\ln(2)} \left[ \int_0^\infty \ln \left( 1 + \frac{g}{e_3} \right) f_G(g) dq \right] $$

$$ - \int_0^\infty \ln \left( 1 + \frac{g}{e_3} \right) f_Q(q) dq \right], \quad (A.1) $$

where $G = EZ$ and $Q = c_2Z$ and their probability density functions are defined as $f_G(g) = \frac{1}{g} f_Z \left( \frac{g}{e_3} \right)$ and $f_Q(q) = \frac{1}{c_2} f_Z \left( \frac{q}{c_3} \right)$, respectively. Using the “change of variable” method [23] for $Z = |h_3|^2$, and defining $\psi = \frac{\ln(1+\mu) - a \mu}{e^{\mu} \sqrt{\pi}}$, the PDF of $Z$ can be written as

$$ f_Z(z) = \psi z^{\frac{\mu-1}{2}} e^{-\phi z} I_{\frac{\mu-1}{2}}(2\lambda \sqrt{z}). \quad (A.2) $$

Noticing the similarity in a structure of $f_G(g)$ and $f_Q(q)$, and using the infinite sum representation of the Bessel function, we write

$$ f_G(g) = \sum_{q=0}^{\infty} \frac{\lambda^2_{q} \mu^{2-1+2q}}{\Gamma(\alpha_3) q!} \psi^3 \left( \frac{1}{E} \right)^{\alpha_3} g^{\alpha_3-1} e^{\psi z}. \quad (A.3) $$

As part of the capacity derivation process, using the following replacements

$$ \ln(1 + a) = G_{2,2}^{1,2} \left( a \left| 1, 1, 1 \right. \right), \quad (A.4a) $$

$$ e^{-a} = G_{0,1}^{1,0} \left( a \left| 0 \right. \right), \quad (A.4b) $$

and the results presented in [22, Eq. (21)], the following is true

$$ \int_0^\infty \ln(1 + \frac{g}{e_3}) f_G(g) dq = \sum_{q=0}^{\infty} \frac{\lambda^2_{q} \mu^{2-1+2q}}{\Gamma(\alpha_3) q!} \psi^3 \left( \frac{1}{E} \right)^{\alpha_3} $$

$$ \times G_{2,2}^{1,2} \left( \frac{\phi_3 c_3}{E} \right)^{-\alpha_3,1-\alpha_3,0}. \quad (A.5) $$

Extrapolating the obtained results to the case of variable $q$ and further substituting them gives the general ergodic capacity expression shown as in Eq. (9).

APPENDIX B
DERIVATION OF PROPOSITION 2

The capacity expression for the $U_1$-to-$U_2$ link is given by

$$ C_{\text{erg}} = \int_0^\infty H_p \log_2 \left( 1 + \gamma_{U_1 \to U_2} \right) f_{\gamma_{U_1 \to U_2}}(\gamma_{U_1 \to U_2}) d\gamma_{U_1 \to U_2} $$

Due to their similarity, the thorough derivation of the PDF is shown only for variable $t$. Re-expressing the Bessel function in terms of infinite sum and using [21, Eq. 3.478.4], the new PDF is written as

$$ f_T(t) = \sum_{q=0}^{\infty} \frac{2 \psi_1 \psi_2 \lambda_{q+1+2q}}{\Gamma(\alpha_1) q!} \sum_{i=0}^{\infty} \frac{\lambda^2_{q+1+2i} t^{i-1}}{\Gamma(\mu_2 + i)!} $$

$$ \times \left( \frac{t}{c_5} \right)^{\alpha_1+\alpha_2} \left( \frac{\phi_1}{\phi_2} \right)^{\beta_1} K_{2\beta_2} \left( 2 \sqrt{\frac{\phi_1}{\phi_2}} \right). \quad (B.3) $$

In the same manner, as part of the capacity derivation process, using the $K_{\nu}(a) = \frac{1}{\sqrt{\pi a^2}} \Gamma \left( \nu \left| \frac{1}{2}, -\frac{1}{2} \right. \right)$ as a replacement, we can further write

$$ \int_0^\infty \ln \left( 1 + \frac{1}{c_4} \right) f_T(t) dt = \sum_{q=0}^{\infty} \sum_{i=0}^{\infty} \frac{\psi_1 \psi_2 \lambda_{q+1+2q}}{\Gamma(\alpha_1) q! c_5^{\alpha_1+\beta_2} (\phi_1)^{\beta_1}} $$

$$ \times \frac{\lambda^2_{q+1+2i} \beta_1 c_4}{\Gamma(\mu_2 + i)!} G_{3,2}^{2,4} \left( \begin{array}{c} -\beta_1, \beta_1 \beta_1, -\beta_1, -\beta_1 \end{array} \right). \quad (B.4) $$

Moreover, applying the results presented in [22, Eq. (21)], the integral in Eq. (B.4) is solved. Extrapolating the obtained results to the case of $v$ and further substituting them, gives the general ergodic capacity expression, as shown in Eq. (10).
REFERENCES