

# On the Performance of Downlink NOMA Systems over Hyper-Rayleigh Fading Channels

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**Abstract**— This paper presents the bit error rate (BER) evaluation of downlink power-domain nonorthogonal multiple access (NOMA) systems in hyper-Rayleigh fading channels. Hyper-Rayleigh models perfectly captures the fading scenarios in which few number of strong components dominate the received signal strength. In addition, the order statistics model is applied to capture the difference in distance between the basestation and users. The transmitted signals of all users are modulated using phase shift keying (PSK), and then multiplexed at the basestation using power-domain NOMA. The well-known successive interference cancellation (SIC) detection is applied at the receivers' side to recover the information signals. Because hyper-Rayleigh models do not generally have closed form expressions, and due to the randomly located discontinuity in the amplitude probability density function (PDF), Monte Carlo simulation is used to evaluate the BER where  $10^7$  symbols are generated per simulation run. The obtained results show that hyper-Rayleigh fading may severely degrade the system BER at high signal-to-noise ratios (SNRs) as compared to Rayleigh channels. However, the Rayleigh fading seems to cause higher BERs at low SNRs.

## I. INTRODUCTION

There is no doubt that the introduction of the fifth generation (5G) communication networks has managed to increase the network capacity, reduce the delay, and provide higher data rates as compared to fourth generation (4G) networks. Nevertheless, the demand for extra capacity and higher data rates is continuously increasing, which maintains the pressure for the design of more efficient networks. For example, while the mobile data traffic in North America is about 9 GB per month in 2019, it is expected to hike up to about 45 GB per month, as anticipated by Ericsson [1]. Therefore, extensive research will remain focused on capacity enhancement technologies such as massive multiple-input-multiple-output (mMIMO) [2], network densification [3], network virtualization [4]–[6], millimeter wave communications [7], free-space optical (FSO) and visible light communication (VLC) [8], [9], and nonorthogonal multiple access (NOMA), which has been considered extensively in the recent literature [10]–[21].

In conventional orthogonal multiple access (OMA), a dedicated resource element in the form of time, frequency, space,

or code is assigned to each user. Although OMA can effectively eliminate the inter-user interference (IUI), it does not provide sufficient spectral efficiency, which is highly desired for most applications. Consequently, NOMA has been introduced as a solution for increasing the capacity of future cellular networks by allowing multiple users to share the same resource element [21]. However, the increase in network capacity is obtained at the expenses of more complex receivers to mitigate the IUI, and higher bit error rate (BER) values [17]–[20].

The IUI caused by nonorthogonal transmissions can be controlled in the power or code domains. In power-domain NOMA, multiple users share the same resource element by assigning different transmission power for each user to enable demultiplexing the signal of each user. In code-domain NOMA, spreading sequences are applied at the basestation (BS), where these sequences have typical properties such as sparsity, low-density and low correlation. At the receiver side, i.e., users' equipment in the downlink scenario, successive interference cancellation (SIC) can be applied to demultiplex and detect the signal for each user. Accordingly, the  $n$ th user detects the signals of users  $1, 2, \dots, n-1$ , and subtracts them from the received signal before extracting its own symbols [17]–[20].

It is worth noting that most of the work reported in the literature about NOMA considers the analysis of the capacity [10]–[13] and outage probability (OP) [14]–[16], whereas a few researchers have considered the bit error rate (BER) analysis [17]–[20]. For example, the authors in [12] introduced a comparison between the capacity of MIMO-NOMA and MIMO-OMA. Power allocation algorithms are proposed in [10] to maximize the ergodic capacity of MIMO-NOMA systems. The capacity of NOMA based cooperative relaying systems is analyzed in [11], and a capacity scaling scheme is proposed in [13] for device-to-device cooperative relaying. In [14], a dynamic power allocation algorithm is proposed for hybrid NOMA systems, and the OP is derived. Moreover, the OP with statistical channel state information (CSI) is derived in [15], whereas imperfect CSI is considered in [16]. The BER analysis of NOMA-based systems has received increased

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attention in the recent literature [17]–[20]. The pairwise error probability (PEP) of NOMA based systems over Nakagami-m fading channels is analyzed in [17], whereas closed form expressions for BER are provided in [20]. Derivations for the error probabilities of space shift keying (SSK) based NOMA, and NOMA-based relay networks with simultaneous wireless information and power transfer (SWIPT) systems are provided in [18], and [19], respectively.

Although the Nakagami-m and Rayleigh are widely used to model wireless channels in urban and suburban scenarios, such channel models are typically not accurate in rural scenarios, where the number of reflected multipath components is very small [22], [23], which corresponds to hyper-Rayleigh fading channels. Therefore, this paper aims at evaluating the BER of NOMA systems in hyper-Rayleigh fading channels, which is to the best of the authors’ knowledge, has never been considered in the literature. The channel is modeled as an ordered hyper-Rayleigh to capture both the small and large fading effects of the wireless channels. The obtained simulation results show that the BER of NOMA systems is drastically different from the case of Rayleigh fading, and from the case of single-user scenarios. More specifically, the results show that the BERs of the Rayleigh and hyper-Rayleigh intersect at certain SNR values, which implies that each fading model will be more severe than the other in a certain SNR range.

The rest of the article is organized as follows. Sections II and III introduce the system and fading channel models, respectively. The SIC based detector is discussed in Section IV. Simulation results are presented in Section V. Finally, the conclusion is provided in Section VI.

## II. SYSTEM AND CHANNEL MODELS

### A. Transmitter Model

In this work, power-domain downlink NOMA system is considered with a single antenna BS and single antenna  $N$  users,  $U_1, \dots, U_N$ . The BS multiplexes the  $N$  signals in power domain, to generate a multiuser signal, by assigning each user a different transmission power. Then, composite signal is transmitted to all users over the same resource element. Therefore, the multiuser signal transmitted from the BS can be expressed as

$$x = \sum_{n=1}^N \sqrt{\beta_n P_T} s_n \quad (1)$$

where  $s_n$  is the  $n$ th user’s data symbol picked from a Gray coded PSK constellation,  $P_T$  is the total transmission power, and  $\beta_n$  is the allocated power coefficient for the  $n$ th user, i.e.,  $\sum_{n=1}^N \beta_n = 1$ . Fig. 1 shows an example for the resultant composite symbol constellation for  $N = 2$ , where both users apply quadrature PSK (QPSK) and the bits of the first user, i.e., far user, are the most left two bits. In this case, the inphase and quadrature components of the multiplexed signal  $x$ , i.e.,  $x_I \triangleq \Re(x)$  and  $x_Q \triangleq \Im(x)$ , can be written as [20],

$$A_{u_1 u_2} = u_1 \sqrt{\beta_1} + u_2 \sqrt{\beta_2}, \quad u_i \in \{1, -1\}. \quad (2)$$

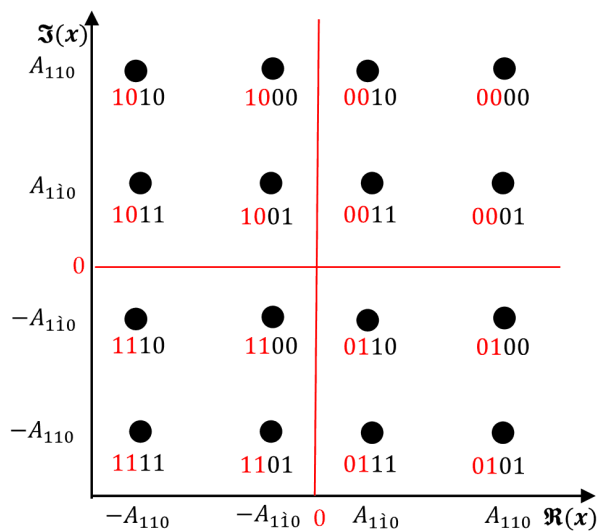


Fig. 1. The constellation diagram of the superimposed symbol  $x$  for  $N = 2$ , [20].

As can be observed from the figure, the superimposed NOMA symbol  $x$  is similar to a 16 quadrature amplitude modulation (16 QAM) constellation with modulation order  $M = 16$ .

### B. Channel Model

The transmission medium in wireless communication systems is typically subject to multipath fading, where multiple copies of the transmitted signal are received at the destination with different time delays, frequency offsets, and amplitudes. The overall effect of the multipath phenomenon depends on the environment where the transmitter and receiver are located, which is widely modeled as Rician or Rayleigh fading. In such models, it is generally assumed that there is a large number of reflected signals, with or without a line-of-sight component, which implies that the signal distribution becomes Gaussian, based on the central limit theorem (CLT). For example, if the number of reflected signals is large and there is a line-of-sight component, then the received signal becomes Gaussian with a non-zero mean, and hence, its envelop is Rician. If there is no line-of-sight, the signal mean becomes zero and its envelop is Rayleigh distributed. Nevertheless, in several wireless applications, only a few number of reflected signals are received, and thus, using the Gaussian assumption based on the central limit theorem (CLT) becomes inaccurate, and the widely adopted Rayleigh fading model will not accurately capture the fading effect. In such scenarios, the hyper-Rayleigh fading model would be more accurate and will be more suitable to use [3], [22], [23].

According to hyper-Rayleigh model, the received signal  $r_n$

at the  $n$ th user can be written as

$$\begin{aligned} r_n &= \left( \sum_{i=1}^L a_{i,n} \cos \phi_{i,n} \right) x + w_n \\ &= h_n x + w_n \\ &= |h_n| e^{j\theta_n} x + w_n \end{aligned} \quad (3)$$

where  $h_n$  is a complex fading coefficient between the BS and the  $n$ th user,  $w_n \sim \mathcal{CN}(0, \sigma_n^2)$  is the additive white Gaussian noise (AWGN),  $L$  is the number of specular received components,  $\phi_{i,n}$  is the phase of the  $i$ th received component and follows uniform distribution, i.e.,  $\phi_{i,n} \sim \mathcal{U}[-\pi, \pi]$ , and  $a_{i,n}$  is the amplitude. Moreover,  $\phi_{i,n}$  and  $a_{i,n}$  are mutually independent  $\forall \{i, n\}$ . For a special case where  $L \rightarrow \infty$ , the CLT can be applied, and thus  $h_n \sim \mathcal{CN}(0, \sigma_{h,n}^2)$  follows the complex Gaussian distribution and  $\alpha_n = |h_n|$  is Rayleigh distributed random variable. However,  $h_n$  generally follows hyper-Rayleigh distribution. For the sake of completeness, the hyper-Rayleigh PDFs for  $L = 2, 3$ , and  $4$  are given by [3], [22], [23]:

Two-ray model,  $L = 2$ :

$$f_\alpha(\alpha) = \frac{2\alpha}{\pi \sqrt{4a_1^2 a_2^2 - (\alpha^2 - a_1^2 - a_2^2)^2}}, \quad 0 \leq \alpha \leq a_1 + a_2 \quad (4)$$

Three-ray model,  $L = 3$ :

$$f_\alpha(\alpha) = \int_0^\pi \frac{2\alpha \Phi(1 - |\bar{\psi}_3|) d\phi_1}{\pi^2 \sqrt{4a_3^2 b_2^2 - (b_3^2 - b_2^2 - a_3^2)^2}}, \quad \check{b}_3 \leq \alpha \leq \hat{b}_3, \quad (5)$$

where  $\Phi(\cdot)$  is the Heaviside step function with  $\Phi(0) = 0$ ,

$$\bar{b}_2^2 \triangleq \|\mathbf{a}_2\|^2 + 2a_1 a_2 \cos(\phi_1),$$

$$\bar{\psi}_3 \triangleq \frac{(\alpha^2 - \bar{b}_2^2 - a_3^2)}{2a_3 \bar{b}_2},$$

$$\check{b}_L = \max \left\{ \left[ 2 \max(\mathbf{a}_L) - \sum_{i=1}^L a_i \right], 0 \right\},$$

$$\|\mathbf{a}_L\|^2 = \sum_{i=1}^L a_i^2, \text{ and } \hat{b}_L = \sum_{i=1}^L a_i.$$

Four-ray model,  $L = 4$ :

$$f_\alpha(\alpha) = \int_{-\pi}^\pi \int_{-\pi}^\pi \frac{\alpha \Phi(1 - |\bar{\psi}_4|) d\phi_1 d\phi_2}{2\pi^3 \sqrt{4a_4^2 b_3^2 - (\alpha^2 - b_3^2 - a_4^2)^2}}, \quad \check{b}_4 \leq \alpha \leq \hat{b}_4, \quad (6)$$

where

$$\bar{\psi}_4 \triangleq \frac{(b_4^2 - \bar{b}_3^2 - a_4^2)}{2A_4 \bar{b}_3},$$

$$\begin{aligned} \bar{b}_3^2 &= \|\mathbf{a}_3\|^2 + 2a_1 a_2 \cos(\phi_1) + a_2 a_3 \cos(\phi_2) \\ &\quad + a_1 a_3 \cos(\phi_2 - \phi_1). \end{aligned}$$

Fig. 2 shown the PDF of hyper-Rayleigh random variable for different values of  $L$ , where the average power is normalized

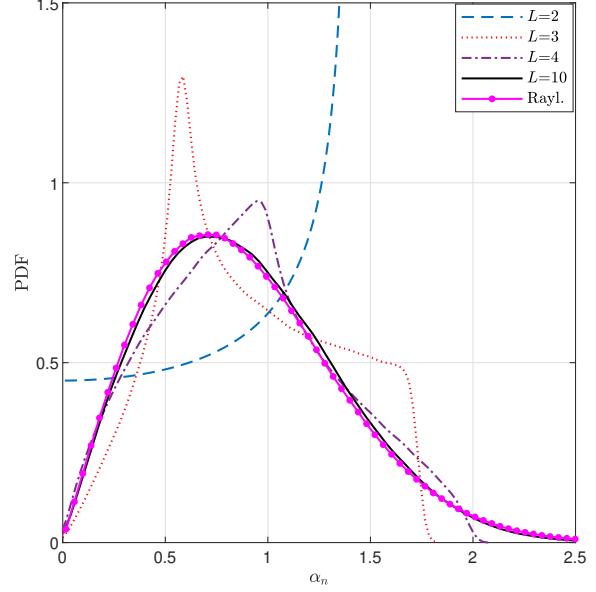


Fig. 2. The distribution of the hyper-Rayleigh with normalized power, i.e.,  $E[\alpha_n^2] = 1$ , for different values of  $L$ .

to 1, i.e.,  $E[\alpha_n^2] = 1$ , and  $a_i = 1 \forall i$  is considered. As can be observed from this figure, the PDF for low values of  $L$  is very different from Rayleigh, denoted as Rayl., distribution, which implies that the well-known Rayleigh model is not suitable when few components dominate the received signal. However, the PDF of a  $\alpha_n$  becomes closer to Rayleigh distribution as the value of  $L$  increases.; for example, the distribution when  $L = 10$  is almost identical with Rayleigh distribution.

### C. Receiver Model

For coherent detection, the channel phase  $\arg\{h_n\} \triangleq \theta_n$  can be estimated and compensated separately from  $\alpha_n$ . Given that the channel phase  $\theta_n$  is estimated and compensated perfectly at the  $n$ th UE receiver, then the received signal after phase compensation can be written as

$$\check{r}_n = r_n e^{-j\theta_n}, \quad \check{r}_n = \alpha_n x + \check{w}_n \quad (7)$$

where  $\check{w}_n = w_n e^{-j\theta_n}$ . Given that  $w_n$  is circularly symmetric, then  $\check{w}_n \sim w_n \sim \mathcal{CN}(0, \sigma_n^2)$ . Without loss of generality, we assume that  $\alpha_1 < \alpha_2 < \dots < \alpha_N$ , and thus, the power allocation coefficients  $\beta_i$ 's are assigned by the BS in the opposite order of channel gains, i.e.,  $\beta_1 > \beta_2 > \dots > \beta_N$ , to enable satisfying the quality of service (QoS) requirements for far users. Fig. 3 shows an example for a two users NOMA network. Therefore, the channel model for each user follow the order statistics [20]

$$f_{\alpha_n}(\alpha_n) = K_n f(\alpha) [F(\alpha)]^{n-1} [1 - F(\alpha)]^{N-n} \quad (8)$$

where  $K_n = \frac{N!}{(n-1)!(N-n)!}$ ,  $f(\alpha)$  and  $F(\alpha)$  are the PDF and CDF of  $\alpha$ , respectively, which corresponds to the hyper-

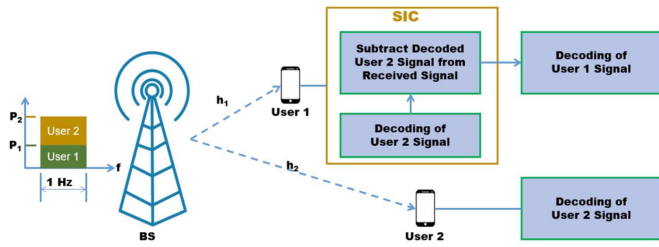


Fig. 3. An example of a BS employs power-domain NOMA to transmit data to two user.

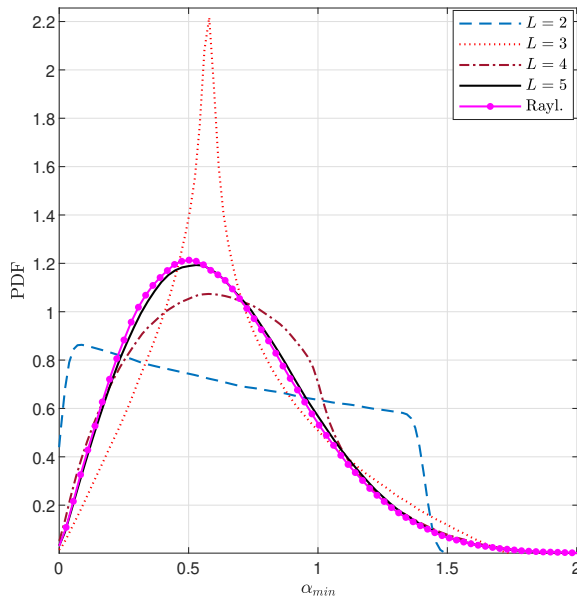


Fig. 4. The distribution of the the ordered channel amplitude  $\alpha_{\min}$ .

Rayleigh model. Unfortunately, deriving closed-form expressions for  $f_{\alpha_n}(\alpha_n)$  given that  $f_{\alpha}(\alpha)$  is hyper-Rayleigh is intractable, and hence, numerical analysis or Monte Carlo simulation can be used. Figs. 4 and 5 show  $f_{\alpha_1}(\alpha_1)$  and  $f_{\alpha_2}(\alpha_2)$ , respectively. As can be noted from the figures, the order statistics of the hyper-Rayleigh is significantly different from the Rayleigh for  $L \leq 3$ .

The detection process at each receiver can be performed using a maximum likelihood detector (MLD), or a SIC detector, which has lower complexity and equivalent performance, and hence, SIC is considered in this work. To extract the information symbols  $s_n$  at the  $n$ th user equipment, SIC is applied where the signal for the  $n$ th user is detected after detecting and subtracting the signals of the first  $n - 1$  users. Therefore, the maximum likelihood detection (MLD) is applied  $n$  times, however, the constellation size in each MLD is equal to the

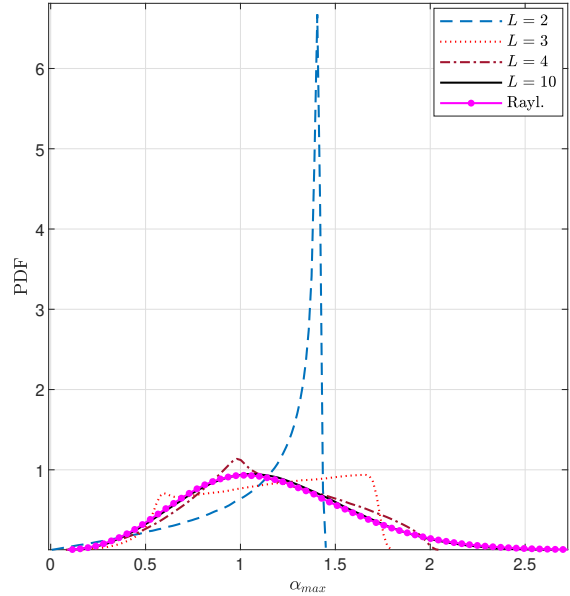


Fig. 5. The distribution of the the ordered channel amplitude  $\alpha_{\max}$ .

modulation order of the  $n$ th user signal, and thus,

$$\hat{s}_n = \arg \min_{\tilde{s}_n} \left| r_n - \hat{h}_n \sum_{i=1}^{n-1} \sqrt{\beta_i} \hat{s}_i - \hat{h}_n \sqrt{\beta_n} \tilde{s}_n \right|^2. \quad (9)$$

For the two users scenario, as shown in Fig. 3, the far user can detect its own signal directly by considering the multi-user signal as an unknown additive noise. For UE-1, the scenario is different because it has to apply SIC to be able to detect its own signal. According to this detection approach, UE-1 detects the signal of UE-2 first, and then the interference cancellation is performed by subtracting UE-2's signal, and finally UE-1 detects its own signal based on the obtained signal after the interference cancellation.

### III. NUMERICAL RESULTS

This section introduces the simulation results for a downlink NOMA system, where Monte Carlo simulation run consists of  $10^7$  realization. The BS and users use a single antennas and apply Gray coded QPSK constellation. The channel between the BS and each user is modeled as a ordered hyper-Rayleigh flat fading channel with unit power, i.e.,  $E[\alpha_i^2] = 1$ , and the results are compared with Rayleigh channel. The number of users  $N = 2$  and the power allocation is set to  $\beta_1 = 0.7$  and  $\beta_2 = 0.3$ , unless it is specified otherwise.

Fig. 6 presents the BER for the two-ray model,  $L = 2$ , and compares it to the conventional Rayleigh fading, i.e.,  $L = \infty$ . In this figure we used  $a_1 = a_2$  with four different values of  $a_i$ , i.e.,  $a_i \in \{0.25, 0.5, 0.75, 1\}$ . As can be noted from this figure, the BER of  $U_1$  is always worse than BER of  $U_2$  in the case of hyper-Rayleigh, while a switch-over occurs for the case of Rayleigh fading. For the cases of  $a_i = 0.75$  and  $a_i = 1$ , the

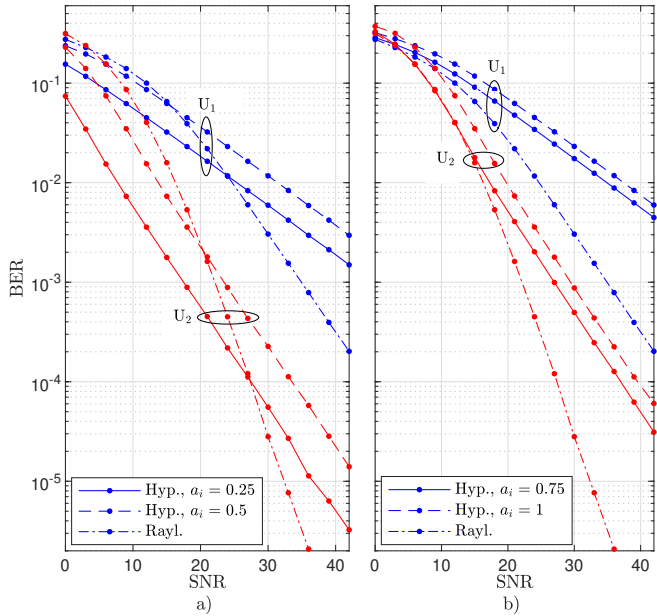


Fig. 6. The probability of bit error for the two-ray model different values of  $\{a_1, a_2\}$ .

BER under Rayleigh fading is lower than the hyper-Rayleigh for the entire SNR. However, it is not the case when  $a_i = 0.25$  or  $a_i = 0.5$ , where the relative difference depends on SNR. From this figure, it can be concluded that Rayleigh model could be inaccurate when the number of specular components is small.

Fig. 7 shows the BER for a two-user NOMA based communication system for different values of specular components  $L$ . The values of  $a_i$  for both  $h_1$  and  $h_2$  are taken from the vector  $\vec{a}_i = \{1, 0.75, 0.5, 0.25, 0.1\}$ , and the results are compared to the case of Rayleigh channel. As can be noted from figure, the BER of  $U_2$  is better than BER of  $U_1$ . In addition, the performance under two-ray channel conditions is very different from Rayleigh scenario; however, both models provide comparable BER as the value of  $L$  increases. Moreover, it can be noted from the figure that fading severity is not directly proportional to the number of specular components  $L$  as the BER fluctuates with different values of  $L$  as compared to Rayleigh fading.

Fig. 8 shows the BER using different power allocation coefficients for  $L = 2$ , where  $a_1 = 1$  and  $a_2 = 0.75$ . It can be observed from the figure that for a given value of  $\beta_1$ , the BER of the near user, i.e.,  $U_2$ , is better than the BER of the far user, i.e.,  $U_1$ . For example, the BER of  $U_2$  is better when  $\beta_1 = 0.8$  although its allocated power is less than  $U_1$ , where the performance loss in the BER of  $U_1$  is due to the long-term fading caused by distance. Another interesting point is that when both user are allocated equal amount of power, i.e.,  $\beta_1 = 0.5$ , both users are not able to recover their signals. It should be noticed that this behavior is due to the large value of interference caused by allocating equal or adjacent values

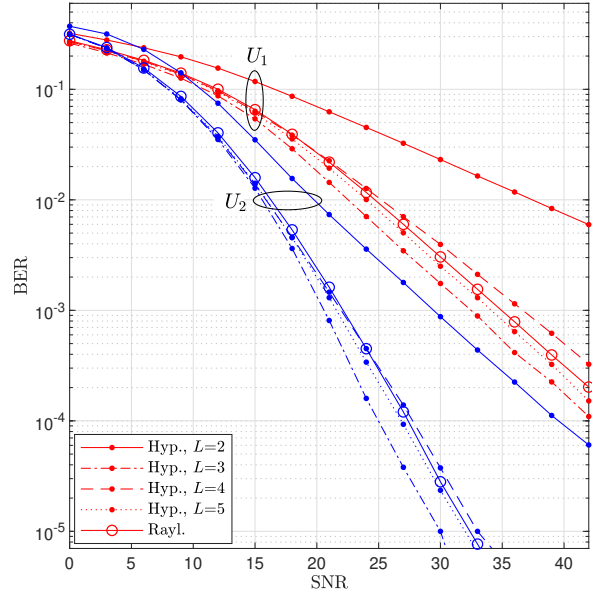


Fig. 7. The BER for a two-user NOMA system for different values of  $L$ .

of power for the users.

#### IV. CONCLUSION

In this paper, the BER performance of downlink NOMA system was studied under hyper-Rayleigh fading channel, which perfectly captures the situation in which few strong received components dominate the received signal. SIC based detection is applied by users for efficient signal extraction. It has been shown that there is considerable difference in the BER values when hyper-Rayleigh is compared to Rayleigh channel, and thus accurate channel modeling is required when designing NOMA based communication system; for example, Rayleigh model is impractical when the received signal is subject to two-ray (ground reflection) model. In addition, results showed that power allocation is very critical to provide users adequate QoS in terms of capacity and BER.

Future work may include the BER analysis of NOMA communication under hyper-Rayleigh fading channels. In addition, the optimum power allocation for user equipment which guarantee the QoS requirements for all users including the BER and transmission rate.

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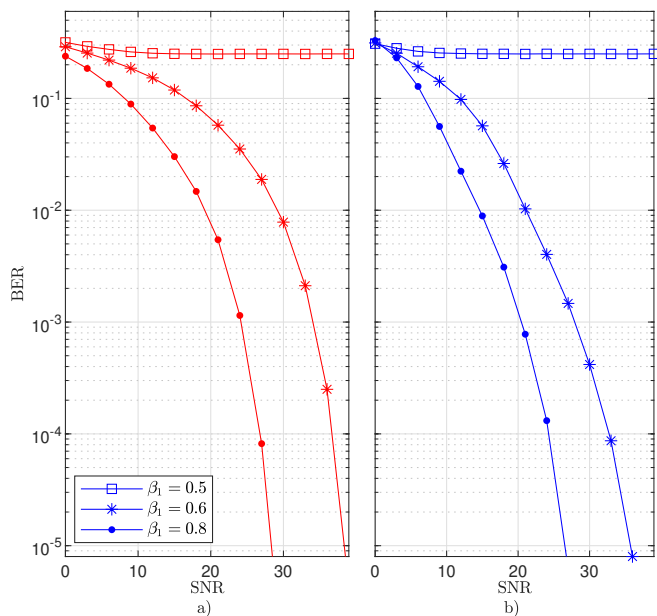


Fig. 8. The BER for different power allocation coefficients,  $L = 2$ : a) BER for user  $U_1$ , b) BER for user  $U_2$ .

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