Multiuser Diversity with Adaptive Modulation in Non-Identically Distributed Nakagami Fading Environments

Anlei Rao, Student Member, IEEE and Mohamed-Slim Alouini, Fellow, IEEE

Abstract—In this paper, we analyze the performance of adaptive modulation with single-cell multiuser scheduling over independent but not identical distributed (i.n.i.d.) Nakagami fading channels. Closed-form expressions are derived for the average channel capacity, spectral efficiency, and bit-error-rate (BER) for both constant-power variable-rate and variable-power variable-rate uncoded/decoded M-ary quadrature amplitude modulation (M-QAM) schemes. We also study the impact of time delay on the average BER of adaptive M-QAM. Selected numerical results show that the multiuser diversity brings a considerably better performance even over i.n.i.d. fading environments.

Index Terms—Adaptive modulation; Nakagami fading; Multiuser diversity

I. INTRODUCTION

To meet the growing demand of services for wireless communications with relatively scarce spectrum, adaptive modulation [1] has been proposed for the last two decades as an effective method of increasing the spectral efficiency. When the channel side information can be estimated and this estimation sent back to the transmitter, the transmit rate and power can be adapted relative to the channel characteristics to achieve better performance. In [2], the channel capacity with channel side information was derived for both constant-power variable-rate (CPVR) and variable-power variable-rate (VPVR) transmission. In [3], a VPVR M-ary quadrature amplitude modulation (M-QAM) scheme was proposed for data transmission over fading channels. This paper showed that there is a constant power gap between the channel capacity and the spectral efficiency for this scheme, and this gap is a function of the maintained bit-error-rate (BER). In [4], variable-rate adaptation for M-QAM with constant-power allocation has been analyzed over the Nakagami fading channels.

In a single-cell multiuser environment, multiuser diversity [5] takes advantage of the variability in fading channels to improve the overall spectral efficiency. For instance, consider that a reasonably large number of users experiencing independent time-varying fading conditions are actively transmitting or receiving data in a given cell. By granting the channel access only to the user with the best transmission quality, we can achieve a diversity gain so that an increase in the channel capacity and spectral efficiency can be reached. Different multiuser scheduling algorithms were proposed in [5], [6] to choose the best user and reduce the feedback load. In this paper, we simply select the user with the largest instantaneous singal-to-noise ratio (SNR) to perform data transmission. With the best user selected, we analyze the average channel capacity, spectral efficiency and BER for both CPVR and VPVR schemes with M-QAM over an independent but not necessarily identical distributed (i.n.i.d.) Nakagami fading channels.

The operation of the CPVR and VPVR schemes needs accurate channel estimation at the receiver, a reliable feedback path between the estimator and transmitter, and a negligible time delay. So an efficient error control scheme is therefore needed to insure the transmitter’s access of the channel side information. In this paper, we assume perfect channel estimation and a reliable feedback, and we also quantify the effect of time delay on the proposed system under consideration. In addition, it has been shown in [4] that the time delay affects the average BER due to the variability of the channel conditions. As such, the goal is to keep the time delay under a threshold to limit the impact on the BER.

The rest of this paper is organized as follows. In section II, the multiuser model is introduced. Next, section III first reviews the channel capacity and spectral efficiency of both CPVR and VPVR schemes, and then gives closed-form expressions for them with best user selection over i.n.i.d. Nakagami fading channels. The performance of the adaptive M-QAM is also analyzed in this section. In section IV, we consider the impact of time delay on the average BER. Finally, some numerical results are shown in section V and conclusions are drawn in the last section.

II. SYSTEM MODEL

Consider a single-cell multiuser environment with $N$ users uniformly distributed in a cell of radius $R$ centered around a base station. In this case, the average SNR $\bar{\gamma}$ for each user is assumed to be log-normal distributed with a shadow standard deviation of $\theta$ dB and an average SNR $\bar{\gamma}_R$ at distance $R$, and a mean value following an exponentially decreasing path loss model with an exponent $\zeta$. The probability density function (PDF) for the average SNR $\bar{\gamma}$ is given by [7], [8]:

$$f_{\bar{\gamma}}(\bar{\gamma}) = \frac{2}{c} \exp \left\{ \frac{2\theta^2 - 2c(\bar{\gamma} - \bar{\gamma}_R)}{c^2} \right\} Q \left( \frac{2\theta^2 - c(\bar{\gamma} - \bar{\gamma}_R)}{c \cdot \theta} \right),$$

(1)
where $c = 10 \log_{10} (e)$ is the exponential path loss parameter, $Q(\cdot)$ is the Gaussian $Q$-function defined as the tail probability of the standard normal distribution, i.e., $Q(x) = \frac{1}{\sqrt{2\pi}} \int_{x}^{\infty} \exp \left( -\frac{t^2}{2} \right) dt$.

At the beginning of each time slot, the user with the largest instantaneous SNR is selected to perform data transmission in this time slot. Let $\gamma_i$ denote the instantaneous SNR for the $i$-th user and user $b$ as the selected best user such that $b = \arg \max_{i \in R} \{ \gamma_i \}$, where $R = \{1, 2, \ldots, N\}$. We can then write $\gamma_b = \max \{ \gamma_1, \gamma_2, \ldots, \gamma_N \}$.

Now suppose that the channels between the users and the base station are slowly-varying flat-fading such that the channel condition changes at a rate much slower than the symbol data rate. In this case, the channel remains roughly constant over hundreds of symbols. For Nakagami fading channels, the SNR of each user $\gamma_i$ is Gamma distributed with the parameters $(\gamma_i, \mu_i)$ and the PDF of $\gamma_i$ is thus given by

$$f_{\gamma_i}(\gamma) = \frac{m_i^{m_i} \gamma^{m_i-1}}{\Gamma(m_i)} \exp \left( -\frac{m_i}{\gamma_i} \gamma \right), \quad \gamma > 0,$$

where $\Gamma(\cdot)$ is the Gamma function.

It is easy to show the cumulative distribution function (CDF) of the instantaneous SNR of the best user can be written as

$$F_{\gamma_b}(\gamma) = \prod_{i=1}^{N} F_{\gamma_i}(\gamma) = \prod_{i=1}^{N} \left( 1 - \frac{\Gamma(m_i, \frac{m_b}{\gamma_i})}{\Gamma(m_i)} \right), \quad \gamma > 0,$$

where $F_{\gamma_i}(\cdot)$ is the upper incomplete gamma function. Restricting $m_i$'s to integers and using the relation $\Gamma(n, x) = \Gamma(n) e^{-x} \sum_{i=0}^{n-1} \frac{x^i}{i!}$, the CDF of $\gamma_b$ can then be expressed for integer $m_i$'s as

$$F_{\gamma_b}(\gamma) = \sum_{S,L} (-1)^{K} C_{S,L} \gamma^I_{S} e^{-E_{S} \gamma}, \quad \gamma > 0,$$

where $S = \{s_1, s_2, \ldots, s_K\}$ is a subset of $R$, $\sum_{S}$ is the sum over all possible $S$, $\sum_{S}$ is the sum defined by $\sum_{S} = \sum_{S} \sum_{i_{s_1}=0}^{m_{s_1}-1} \sum_{i_{s_2}=0}^{m_{s_2}-1} \cdots \sum_{i_{s_K}=0}^{m_{s_K}-1}$. In (3), $I_S$, $E_S$ and $C_{S,L}$ are given by $I_S = I_{s_1} + I_{s_2} + \cdots + I_{s_K}$, $E_S = E_{s_1} + E_{s_2} + \cdots + E_{s_K}$, and $C_{S,L} = \prod_{i=1}^{K} \frac{1}{1 + \frac{m_{s_i}}{\gamma_i}}$. When $S = \phi$ is an empty set, we have $I_S = 0$, $E_S = 0$ and $C_{S,L} = 1$. By simply differentiating (3) over $\gamma$, the PDF of $\gamma_b$ can then be written as

$$f_{\gamma_b}(\gamma) = \sum_{S,L} (-1)^{K} C_{S,L} (I_S - E_S \gamma) \gamma^{I_S-1} e^{-E_S \gamma}, \quad \gamma > 0.$$

III. PERFORMANCE ANALYSIS

Assuming perfect channel estimation and reliable feedback, the channel capacity of a fading channel $< C > [\text{Bits/Sec}]$ with variable power allocation and rate adaptation is given by [2]

$$< C > = \max_{P(\gamma)} \left\{ \int_{0}^{\infty} W \log_2 \left( 1 + \frac{P(\gamma)}{P} \right) f_{\gamma}(\gamma) d\gamma \right\},$$

where $W$ [Hz] is the channel bandwidth, $f_{\gamma}(\gamma)$ is the distribution of the received SNR, $P$ is the average transmit power, and $P(\gamma)$ is the instantaneous transmit power chosen relative to $\gamma$, which is subject to the power constraint

$$\int_{0}^{\infty} P(\gamma) f_{\gamma}(\gamma) d\gamma \leq P,$$

If we choose uncoded M-QAM for the adaptive modulation scheme, the BER with variable power allocation can be then well approximated by [3]

$$\text{BER}(M, \gamma) \simeq a_0 \exp \left( -\frac{b_0 \gamma}{(M - 1) \frac{P}{P_0}} \right),$$

where $a_0 = 1/5$, and $b_0 = 3/2$.

To maintain a fixed BER of $\text{BER}_0$, the constellation size $M$ is adjusted as

$$M = 1 + \frac{\gamma}{K_0} \frac{P(\gamma)}{P},$$

where $K_0 = -\frac{1}{b_0} \ln(\frac{\text{BER}_0}{a_0})$.

A. Constant-Power Variable-Rate Adaptation

1) Channel Capacity: With a constant transmit power $P(\gamma) = P$, the channel capacity $< C >_{cpvr}$ is given as

$$< C >_{cpvr} = W \int_{0}^{\infty} \log_2 (1 + \gamma) f_{\gamma}(\gamma) d\gamma.$$

Substituting (4) in (9) yields the channel capacity per unit bandwidth over i.n.i.d. Nakagami fading channels with the best user selection

$$< C >_{cpvr} = \log_2(e) \sum_{S,L} (-1)^{K} C_{S,L} \times (I_S T_0(E_S) - E_S T_{0+1}(E_S)),$$

where $T_n(\mu)$ is defined in [10, (32)] as

$$T_n(\mu) = \int_{0}^{\infty} t^{n-1} \ln(1+t)e^{-\mu t}; \quad \mu > 0; n = 1, 2, \cdots$$

and its closed-form evaluation for a positive integer $n$ is also derived in [10, (78)] as

$$T_n(\mu) = (n-1)! e^\mu \sum_{k=1}^{n} \frac{(-\mu)^{k-n}}{\mu^k}.$$

2) Spectral Efficiency: With the constant power scheme and $M$ chosen in (8), the spectral efficiency for continuous-rate M-QAM can be approximated by [4]

$$\frac{R}{W} >_{cpvr} = \int_{0}^{\infty} \log_2 \left( 1 + \frac{\gamma}{K_0} \right) f_{\gamma}(\gamma) d\gamma.$$

Substituting (4) in (12), the above spectral efficiency per unit bandwidth can be written as

$$\frac{R}{W} >_{cpvr} = \log_2(e) \sum_{S,L} (-1)^{K} C_{S,L} K_0^I_{S} \times (I_S T_0(E_S K_0) - E_S K_0 T_{0+1}(E_S K_0)).$$
3) Constellation Restriction 1: For practical use, the constellation size $M$ is often restricted to $2^n$ for a positive integer $n$. We divide the whole SNR range into $T+1$ fading regions, and assign $M_n$ as the constellation size to the $n$-th region. When the estimated SNR falls in the $n$-th region, $M_n$ is chosen for M-QAM. For uncoded M-QAM, the boundary SNRs are chosen as

$$
\gamma_n = \begin{cases} 
\text{erfc}^{-1}(2\text{BER}_0)^2, & n = 1 \\
K_0(2^n - 1), & n = 0, 2, 3, \cdots, T; \\
+\infty & n = T + 1;
\end{cases}
$$  \hspace{1cm} (14)

where \(\text{erfc}(\cdot)\) is the complementary error function. When $M = 2$, the modulation reduces to BPSK, the BER of which is $\frac{1}{2} \text{erfc}(\sqrt{\gamma})$.

For coded M-QAM, such as trellis codes in [11], [12], the BER is approximated by

$$\text{BER}(M_n, \gamma) \approx a_n \exp \left( \frac{b_n \gamma}{M_n} \right),$$  \hspace{1cm} (15)

where $a_n$ and $b_n$ for different constellation size $M_n$ are given in [11]. The boundary SNRs for $n = 2, 3, \cdots, T$ in this case are chosen as $\gamma_n = K_n 2^n$, where $K_n = -\frac{1}{b_n} \ln (\text{BER}_0)$.

The spectral efficiency is the sum of the data rates $(\log_2(M_n))$ associated with each region multiplied by the probability of the instantaneous SNR $\gamma$ falling in that region

$$\frac{R}{W} > \text{Ber}_{\text{cprd}} = \sum_{n=1}^{T} r_n p_n,$$  \hspace{1cm} (16)

where $r_n = \log_2(M_n) = n$, $p_n = \int_{\gamma_n}^{\gamma_{n+1}} f_\gamma(\gamma) d\gamma = F_{\gamma_n}(\gamma_{n+1}) - F_{\gamma_n}(\gamma_n)$, which can be easily computed with (2).

For continuous-rate M-QAM, it always operates at the target BER0. When the constellation size is restricted to some integer values, however, the discrete M-QAM operates at an average $< \text{BER} >_{\text{cprd}}$ lower than BER0. Similar to the calculation of the spectral efficiency, the average BER with constellation restriction is the ratio of the average number of bits in error over the total average number of transmitted bits [4]

$$< \text{BER} >_{\text{cprd}} = \frac{\sum_{n=1}^{T} r_n \text{BER}_n}{\sum_{n=1}^{T} r_n p_n},$$  \hspace{1cm} (17)

where $\text{BER}_n$ is calculated by

$$\text{BER}_n = \int_{\gamma_n}^{\gamma_{n+1}} \text{BER}(M_n, \gamma) f_\gamma(\gamma) d\gamma.$$  \hspace{1cm} (18)

Substituting $\text{BER}(M_n, \gamma) = a_n \left( \frac{\text{BER}_n}{a_n} \right)^{\frac{1}{2}}$ into (18), we have

$$< \text{BER} >_{\text{cprd}} = \frac{\sum_{n=1}^{T} r_n \text{BER}_n}{\sum_{n=1}^{T} r_n p_n} > \text{BER}_n,$$  \hspace{1cm} (19)

where $\text{BER}_n$ is written as

$$\text{BER}_n = \sum_{S,L} (-1)^{K_n C_{S,L}} \beta_{S_n}^{\beta_{S_n}} \left( \int_{S} f_{S_n}(l_S, \beta_{S_n} \gamma_n, \beta_{S_n} \gamma_{n+1}) - \frac{E_S}{\beta_{S_n}} \Gamma(l_S + 1, \beta_{S_n} \gamma_n, \beta_{S_n} \gamma_{n+1}) \right),$$  \hspace{1cm} (19)

where $\beta_{S_n} = E_S + b_n K_n / \gamma_n$, and $\Gamma(n, a, b) = \Gamma(n, a) - \Gamma(n, b)$.

B. Variable-Power Variable-Rate Adaptation

1) Channel Capacity: By optimizing the channel capacity (5) with the power constraint (6), we have the channel capacity with optimal power allocation as [2]

$$< C >_{\text{vpvr}} = W \int_{\gamma_0}^{\infty} \log_2 \left( \frac{\gamma}{\gamma_0} \right) f_\gamma(\gamma) d\gamma,$$  \hspace{1cm} (20)

where $\gamma_0$ is the optimal cutoff SNR below which data transmission is suspended and can be determined by solving

$$\frac{1 - F_{\gamma_0}(\gamma_0)}{\gamma_0} = 1 + \sum_{S,L} (-1)^{K_n C_{S,L}} E_S^{1-l_S} \times \left( l_S \Gamma(l_S - 1, E_S \gamma_0) - \Gamma(l_S, E_S \gamma_0) \right).$$  \hspace{1cm} (21)

When $l_s = 0$, we have $l_S \Gamma(l_S - 1, E_S \gamma_0) = 0$ and $\Gamma(0, E_S \gamma_0) = E_1(E_S \gamma_0)$, where $E_1(\cdot)$ is the exponential integral of the first order [13], and $F_{\gamma_0}(\gamma_0)$ can be easily evaluated using (2).

By substituting (4) in (20), the channel capacity per unit bandwidth with optimal power allocation can be written as

$$< \frac{C}{W} >_{\text{vpvr}} = \log_2(e) \sum_{S,L} (-1)^{K_n C_{S,L}} \Gamma(l_S),$$  \hspace{1cm} (22)

where the definition of $J_n(\mu)$ is given in [10, (14)] as

$$J_n(\mu) = \int_{1}^{\infty} t^{n-1} \ln(t) e^{-\mu t} dt; \hspace{1cm} t > 0; n = 1, 2, \cdots$$  \hspace{1cm} (23)

and its closed-form evaluation for a positive integer $n$ is also derived in [10, (64)] as $J_n(\mu) = \Gamma(n) \mu^{-n} \sum_{k=0}^{n-1} \frac{1}{\Gamma(k+1)}$.

2) Spectral Efficiency: With the optimal power allocation and $M$ chosen in (8), the spectral efficiency is given by [3]

$$\frac{R}{W} > \text{vpvr} = \int_{\gamma_K}^{\infty} \log_2 \left( \frac{\gamma}{\gamma_K} \right) f_\gamma(\gamma) d\gamma,$$  \hspace{1cm} (24)

where $\gamma_K$ is the cutoff fade depth under which no power is allocated to transmit data, which is determined by $K_0 \int_{\gamma_K}^{\infty} \left( \frac{1}{\gamma_K - \frac{1}{\gamma} \frac{1}{\gamma} f_\gamma(\gamma) d\gamma = 1.$$  \hspace{1cm} (25)

and its closed-form evaluation for a positive integer $n$ is also derived in [10, (64)] as $J_n(\mu) = \Gamma(n) \mu^{-n} \sum_{k=0}^{n-1} \frac{1}{\Gamma(k+1)}$.

The spectral efficiency with optimal power allocation over i.n.i.d. Nakagami fading channels can be derived by simply replacing $\gamma_0$ in (22) with $\gamma_K$

$$\frac{R}{W} > \text{vpvr} = \log_2(e) \sum_{S,L} (-1)^{K_n C_{S,L}} \Gamma(l_S),$$  \hspace{1cm} (25)

$$(l_S J_S(E_S \gamma_K) - E_S \gamma_K J_{S+1}(E_S \gamma_K)).$$  \hspace{1cm} (25)
<BER_{\tau} >_{cprv} = a_0 \sum_{S,L} (-1)^K C_{S,L} \left( l_S F_{m_b}^{0,\infty} (\alpha, \beta, l_S - 1, E_S) - E_S F_{m_b}^{0,\infty} (\alpha, \beta, l_S, E_S) \right), \tag{28}

<BER_{\tau} >_{vprv} = a_0 \sum_{S,L} (-1)^K C_{S,L} \left( l_S F_{m_b}^{\gamma,\infty} (\alpha, \beta, l_S - 1, E_S) - E_S F_{m_b}^{\gamma,\infty} (\alpha, \beta, l_S, E_S) \right). \tag{30}

3) Constellation Restriction: Similar to the constellation restriction in the CPVR scheme, we also divide the whole SNR range into \( T + 1 \) fading regions, but restrict our constellation size as \( M_0 = 0, M_1 = 2 \), and \( M_n = 2^{2(n-1)}, n = 2, \ldots, T \). The boundary SNR \( \gamma_n \) for coded M-QAM is determined by \( \gamma_n = \gamma^* M_n \) with a parameter \( \gamma^* > 0 \) which satisfies the power constraint (26) if we adopt the same power allocation scheme as in [3]

\[
\sum_{n=1}^{T} K_n (M_n - 1) \int_{\gamma^* M_n}^{\gamma^* M_n + 1} \frac{f_{\gamma}(\gamma)}{\gamma} d\gamma = 1. \tag{26}
\]

Replacing \( K_n \) in (26) with \( K_0 \), we can find \( \gamma^* \) by using (26) for the uncoded M-QAM.

The spectral efficiency can also be calculate by (16) with \( p_n = F_{\gamma_n}(\gamma^* M_n + 1) - F_{\gamma_n}(\gamma^* M_n) \) and \( r_1 = 1, r_n = 2(n - 1), n = 2, 3, \ldots, T \). And the BER will always operate at \( \text{BER}_0 \), i.e. \(< \text{BER} >_{vph} = \text{BER}_0 \).

IV. Effects of Time Delay

Recall that the choice of the constellation size \( M \) and power allocation \( P(\gamma) \) are based on a channel estimation at time \( t \), while the data are sent over the channel at time \( t + \tau \), where \( \tau \) is the time delay for the channel estimation and feedback. If the time delay \( \tau \) degrades the BER significantly, then the adaptive technique will not work. Assuming perfect channel estimation and reliable feedback, we analyze in this section the impact of time delay on BER with the best user selection in an i.n.i.d. Nakagami fading environment.

We denote \( \rho \) as the correlation factor between the SNR \( \gamma \) at time \( t \) and the SNR \( \gamma_{r} \) at time \( t + \tau \). \( \rho \) can be expressed in terms of time delay \( \tau \), the mobile speed \( v \), and the wavelength of the carrier frequency \( \lambda_c \) by \( \rho = J_0^2(2\pi f_D \tau) \) [14], where \( J_0(\cdot) \) is the zero-order Bessel function of the first kind, and \( f_D = v/\lambda_c \) is the maximum Doppler frequency shift. With \( \rho \), the conditional PDF of \( \gamma_{r} \) over \( \gamma_b \) is given by [9]

\[
f_{\gamma_{r}|\gamma_b}(\gamma_{r}|\gamma_b) = \frac{m_b}{(1 - \rho)^{\gamma_b}} \left( \gamma_{r}/\rho \right)^{m_b - 1} \exp \left( - \frac{m_b \rho \gamma_{r} + \gamma_{r}}{1 - \rho} \right) \times I_{m_b - 1} \left( \frac{2 m_b \sqrt{\rho \gamma_{r}}}{1 - \rho} \right), \tag{27}
\]

where \( m_b \) and \( \gamma_b \) are the Nakagami parameter and average SNR for the best user \( b \), respectively, and \( I_n(\cdot) \) is the modified Bessel function of the first kind.

A. Constant Power Scheme

With constant power allocation, the instantaneous BER for uncoded M-QAM without constellation restriction is given by

\[
\text{BER}(\gamma, \gamma_r) = a_0 \left( \frac{\text{BER}_0}{\alpha} \right)^{\gamma}.
\]

Averaging it over (27) yields the conditional average BER on \( \gamma \) as

\[
\text{BER}(\gamma) = a_0 \left( \frac{\gamma}{\gamma + \alpha} \right)^{m_b} \exp \left( - \frac{\beta \gamma}{\gamma + \alpha} \right),
\]

where \( \alpha = \frac{b_b(1 - \rho)K_0 \gamma_b}{\gamma_b} \) and \( \beta = b_b K_0 \rho \). Averaging \( \text{BER}(\gamma) \) over the PDF of \( \gamma_b \) given in (4) yields the average BER for CPVR scheme accounting for the impact of the time delay given by (28) shown on the top of this page, where \( F_{m,b}(\alpha, \beta, a, b) \) is defined as \( F_{m,b}(\alpha, \beta, a, b) = \int_{\gamma}^{x} \left( \frac{\gamma}{\gamma + \alpha} \right)^{\beta} e^{-\beta \gamma} d\gamma \).

Considering the restriction of the constellation, the instantaneous BER yields \( \text{BER}(\gamma_{r}, \gamma_{r}) = a_n \left( \frac{\text{BER}_n}{\alpha} \right)^{\gamma_{r}} \). Averaging this expression over the conditional PDF (27) yields the conditional average BER on \( \gamma \) for the discrete rate as

\[
\text{BER}_n(\gamma) = a_n \left( \frac{\gamma_n}{\gamma_n + \alpha_n} \right)^{m_b} \exp \left( - \frac{\beta_n \gamma}{\gamma_n + \alpha_n} \right),
\]

where \( \alpha_n = \frac{b_n(1 - \rho)K_0 \gamma_b}{\gamma_b} \) and \( \beta_n = b_b K_0 \rho \). The average BER is calculated by (17) with the \( \text{BER}_n \) calculated by (18), the result of which is given by

\[
\text{BER}_n = \frac{\left( \frac{\gamma_n}{\gamma_n + \alpha_n} \right)^{m_b}}{\alpha_n} \sum_{S,L} (-1)^K \frac{a_n C_{S,L}}{\alpha_n^S} \times I_{m_b - 1} \left( \frac{2 m_b \sqrt{\rho \gamma_{r}}}{1 - \rho} \right) \tag{29}
\]

\[
\times \left( l_S \Gamma(l_S, \alpha_n, \gamma_n, \gamma_n + 1) \right),
\]

where \( \alpha_n = E_S + \frac{\beta_n \gamma}{\alpha_n + \gamma_n} \) with \( \alpha_n \) and \( \beta_n \) given as above.

B. Variable Power Scheme

For the variable power scheme, the power allocation \( P(\gamma)/\bar{P} \) is chosen so that it always operates at the target BER. Taking time delay into consideration, the instantaneous BER yields \( \text{BER}(\gamma, \gamma_r) = a_0 \left( \frac{\text{BER}_0}{\alpha} \right)^{\gamma} \). For the continuous-rate VPVR scheme, the data is transmitted only when the instantaneous SNR is above the cutoff SNR, or \( \gamma > \gamma_K \). So the average BER for continuous-rate VPVR scheme with time delay effect is given by (30) which is shown on the top of this page.

With the constellation restriction, the average BER is calculated by (17), where \( r_1 = 1, r_n = 2(n - 1), n = 2, 3, \ldots, T \), \( p_n = F_{\gamma_n}(\gamma^* M_{n+1}) - F_{\gamma_n}(\gamma^* M_n) \), and \( \text{BER}_n \) is given in this case by

\[
\text{BER}_n = a_n \sum_{S,L} (-1)^K C_{S,L} \left( l_S F_{m_b}^{\gamma_n,\gamma_n+1} (\alpha_n, \beta_n, l_S, E_S) \right.
\]

\[
- E_S F_{m_b}^{\gamma_n,\gamma_n+1} (\alpha_n, \beta_n, l_S, E_S) \right) \tag{31}
\]
When the number of users increases, the cell radius remains too much complexity. To achieve a diversity gain without causing decreasing, so the number of users in the cell should be moderate in order to achieve a diversity gain without causing too much complexity.

The different number of users (\(\bar{\gamma}R\) for each user are supposed to be i.n.i.d following the distribution given in (1) with a cell-range SNR \(\bar{\gamma}R\), and generated in our simulation based on the procedure described in the Appendix. In our results, the Nakagami parameter for each user is fixed as \(m = 3\).

In Fig. 1 and Fig. 2, we plot the channel capacity and spectral efficiency for different number of users for the CPVR and VPVR schemes, respectively. Note that all the users are distributed in a single cell with the cell-range SNR \(\bar{\gamma}R = 0\) dB. When the number of users increases, the cell radius remains the same the \(\bar{\gamma}R\) is maintained unchanged, so we can see an improvement of the channel capacity and spectral efficiency for both coded and uncoded schemes because of the multiuser diversity.

In Fig. 1 and Fig. 2, the channel capacity and spectral efficiency for different cell-range SNR \(\bar{\gamma}R\) are shown for the CPVR and VPVR schemes with uncoded M-QAM, respectively. In Fig. 3, we see a saturation of the spectral efficiency at large SNR area for the discrete-rate CPVR scheme. For the large SNR range, the current constellation size is not enough for transmission with a higher efficiency, and larger constellation size is frequently used to achieve better performance. While in Fig. 4, the power allocation for discrete-rate VPVR scheme will be highly suboptimal for large SNR. In this circumstances, the power is far from being fully utilized and no \(\gamma^*\) is found as the solution for (26), so that no results for large SNR are shown in the figure. In order to utilize the power fully, larger constellation size is also needed to achieve a better performance.

In Fig. 5, we compare the BER of discrete-rate CPVR schemes with uncoded M-QAM for different regions. From the curves in this figure we see that the discrete-rate CPVR scheme operates at a BER smaller than the target BER_0, and the more regions the whole SNR is divided into, the higher BER we have. As the number of regions goes to infinity, the scheme will converge to the continuous-rate scheme and operate at the target BER_0.

Finally Fig. 6 analyzes the impact of time delay on BER for both the CPVR and VPVR schemes with uncoded M-QAM. From the figure we can see that, if the normalized delay is maintained under a value of \(10^{-2}\), the impact of delay is
Comparing the curves of BER $0 = 10^{-3}$ and BER $0 = 10^{-6}$, we can find that the smaller the BER of what we want to maintain, the more sensitive the BER is to the time delay.

### VI. CONCLUSION

In this paper, we analyze the CPVR and VPVR schemes over i.n.i.d. Nakagami fading channels when the best user is chosen in a multiuser environment. Close-form expressions have been derived for average channel capacity, spectral efficiency, and BER with uncoded-coded M-QAM. Numerical results show that by selecting the best user, we can still improve the performance considerably through the diversity gain. In addition, the impact of time delay on BER is also studied, and numerical results show that when the normalized time delay is below $10^{-2}$, the impact is tolerable. However, the BER degrades considerably when the normalized time delay exceeds $10^{-2}$.

### APPENDIX: RANDOM GENERATION OF THE AVERAGE SNR

In this paper, we describe the acceptance-rejection algorithm [15] used to generate the average SNR $\bar{\gamma}_R$‘s following the PDF given in (1) by the acceptance-rejection algorithm. This algorithm generates sampling values from an arbitrary PDF $f_X(x)$ by using an auxiliary PDF $f_Y(x)$. The PDFs $f_X(x)$ and $f_Y(x)$ should satisfy the following assumptions:

- It is simple to sample from $f_Y(x)$ (i.e. $f_Y(x)$ should be or be made of a common density function such as Gaussian or uniform, so that it is easy to generate the corresponding random samples);
- There exists $0 < \rho \leq 1$ so that $\rho f_X(x) \leq 1$, for all $x$.

As long as we find a proper $f_Y(x)$, the sampling procedure to generate samples for $f_X(x)$ yields

- **Step 1** Sample two independent random variables $y$ from $f_Y(x)$ and $u$ from the uniform distribution $U(0, 1)$;
- **Step 2** If $u \leq \rho f_X(y)$, then accept $x = y$ as a sample from $f_X(x)$, otherwise reject $y$, and go to Step 1.

For our case, the target sampling PDF is given in (1). Let us denote $x = \frac{2\omega^2 - c(\bar{\gamma} - \bar{\gamma}_R)}{2}$ so that $\bar{\gamma} = \bar{\gamma}_R + \frac{2\omega^2}{c} - \theta x$, and the PDF of $x$ is then given by

$$f_X(x) = \frac{2\theta}{c} \exp \left( \frac{2\theta}{c} x - \frac{2\theta^2}{c^2} \right) Q(x).$$  \hspace{0.5cm} (32)

Note that $Q(x) < 1$ and $Q(x) < \frac{1}{2} \exp \left( -\frac{x^2}{2} \right)$ for $x > 0$. We can find the auxiliary $f_Y(x)$ as

$$f_Y(x) = \begin{cases} 
\frac{2\theta}{c} \exp \left\{ \frac{2\theta}{c} \left( -\frac{\theta}{c} \right) \right\} & x \in (-\infty, \theta/c] \\
\frac{2\theta}{c} & x \in (\theta/c, 2\theta/c) \\
\frac{2\theta}{c} \exp \left\{ -\frac{1}{2} \left( -\frac{\theta}{c} \right)^2 \right\} & x \in [2\theta/c, +\infty) 
\end{cases} \hspace{0.5cm} (33)$$

where $\rho = \frac{1}{1+2\theta^2/c^2+2\theta^2/c^2 x}$ $< 1$ is a parameter to assure that $\int_{-\infty}^{+\infty} f_Y(x)dx = 1$.

With the above $f_X(x)$ in (32) and $f_Y(x)$ in (33), we can generate the samples for (1) through the following procedure:

- **Step 1** Sample $u_0$ from the uniform distribution $U(0, 1)$.
  For $u_0 \in (0, p_1)$, go to Step 2; for $u_0 \in (p_1, p_2)$, go to Step 3; and for $u_0 \in (p_2, 1)$, go to Step 4:
- **Step 2** Sample $x_0$ from the exponential distribution $Exp(\theta/c)$, and calculate $\omega = Q(y_0)$ with $y_0 = \theta/c - x_0$; and go to Step 5:
- **Step 3** Sample $x_0$ from the $U(0, 1)$, and calculate $\omega = \exp(2y_0\theta/c - 2\theta^2/c^2)Q(y_0)$ with $y_0 = (1 + x_0)\theta/c$; and go to Step 5;
- **Step 4** Sample $x_0$ from the normal distribution $N(0, 1)$, and calculate $\omega = \exp(2y_0\theta/c - 2\theta^2/c^2 + (y_0 - 2\theta/c)^2/2)Q(y_0)$ with $y_0 = 2\theta/c + |x_0|$; and go to Step 5;
- **Step 5** Sample $u$ from $U(0, 1)$. If $u \leq \omega$, then accept $y_0$ as a sample from $f_X(x)$, otherwise reject $y_0$ and go back to Step 1.
- **Step 6** $\bar{\gamma} = \bar{\gamma}_R + \frac{2\omega^2}{c} - \theta y_0$ is a sample from (1).

where $p_1 = \int_{-\infty}^{\theta/c} f_X(x)dx = 2Q(\theta/c)$ and $p_2 = \int_{-\infty}^{2\theta/c} f_X(x)dx = 0.5 + \exp(2\theta^2/c^2)Q(2\theta/c)$.

### REFERENCES


