Spectral Efficiency Analysis of Multi-Cell Massive MIMO Systems with Ricean Fading

Pei Liu†, Kai Luo†, Da Chen†, Tao Jiang‡, and Michail Matthaiou§
†School of Electronic Information and Communications, Huazhong University of Science and Technology, Wuhan 430074, P. R. China (e-mail: {peil, kluo, chenda, taojiang}@hust.edu.cn)
‡Institute of Electronics, Communications and Information Technology (ECIT), Queen’s University Belfast, BT3 9DT, Belfast, United Kingdom (e-mail: m.matthaiou@qub.ac.uk)

Abstract—This paper investigates the spectral efficiency of multi-cell massive multiple-input multiple-output systems with Ricean fading that utilize the linear maximal-ratio combining detector. We firstly present closed-form expressions for the effective signal-to-interference-plus-noise ratio (SINR) with the least squares and minimum mean squared error (MMSE) estimation methods, respectively, which apply for any number of base-station antennas \( M \) and any Ricean \( K \)-factor. Also, the obtained results can be particularized in Rayleigh fading conditions when the Ricean \( K \)-factor is equal to zero. In the following, novel exact asymptotic expressions of the effective SINR are derived in the high \( M \) and high Ricean \( K \)-factor regimes. The corresponding analysis shows that pilot contamination is removed by the MMSE estimator when we consider both infinite \( M \) and infinite Ricean \( K \)-factor, while the pilot contamination phenomenon persists for the rest of cases. All the theoretical results are verified via Monte-Carlo simulations.

Index Terms—Effective SINR, multi-cell massive multiple-input multiple-output, pilot contamination, Ricean fading, spectral efficiency.

I. INTRODUCTION

In the design of future communication systems, spectral efficiency (SE) becomes one of the dominant targets [1]. Massive multiple-input multiple-output (MIMO) [2], where the base station (BS) is equipped with hundreds of antennas to serve tens of users in the same time-frequency resource block, has attracted substantial attention from academia and industry thanks to the the high SE gains provided by the massive array [3]–[5]. The fundamental problem of placing a massive number of antennas in a confined space, can be addressed by pushing the operating frequency in the millimeter wave band, where the wavelengths become inherently small. We recall that mm-wave channels are typically modeled via the Ricean fading distribution due to the presence of line-of-sight (LOS) or specular components [6]. Hence, the SE analysis in massive MIMO systems with Ricean fading becomes a problem of practical relevance.

We will now review the relevant state-of-the-art in the massive MIMO literature. We firstly note the work of [3], which derived useful lower bounds on the uplink ergodic rate for different classical linear receivers and analyzed the SE performance over a single-cell massive MIMO system operating in Rayleigh fading for perfect channel state information (CSI) and imperfect CSI, respectively. The authors in [8] considered the achievable uplink rates based on the least squares (LS) and minimum mean squared error (MMSE) estimation methods in multi-cell massive MIMO systems and investigated the pilot power allocation scheme to maximize the minimum SE. As a general comment, most issues pertaining to the achievable SE of massive MIMO, massive MIMO systems in Rayleigh fading have been largely and extensively characterized. Apart from that, some recent investigations have analyzed the SE performance in massive MIMO systems based on Ricean fading. For example, for a single-cell environment, [6] examined the scaling law and obtained the approximate expressions of uplink ergodic rate by considering Ricean fading and both perfect CSI and imperfect CSI. Moreover, for multi-cell massive MIMO systems, [9] obtained a closed-form approximation and [10] provided a lower bound of the achievable uplink rate with imperfect CSI, respectively. Also, [11] considered a similar scenario as [9] and pursued an asymptotic analysis of the achievable rate. Recently, [12], [13] systematically investigated the SE performance with spatially correlated Ricean fading channels and obtained closed-form expressions of uplink/downlink SE for different channel estimation methods. However, it appears that those expressions are untraceable to gain a very clear insight. Also, the corresponding asymptotic analysis in massive antenna regime is obtained at the cost of the tougher conditions of the spatial covariance matrix and LOS component. From the above discussion, it becomes apparent that a straightforward and exact theoretical analysis of massive MIMO systems with Ricean fading is missing from the open literature.

In this paper, we firstly introduce a general analytic framework for investigating the achievable uplink SE with a linear maximal-ratio combining (MRC) detector. Then, we derive two exact closed-form expressions of the effective signal-to-interference-plus-noise ratio (SINR) based on the LS and
MMSE estimation methods, respectively, which apply for any number of antennas $M$ and any Ricean $K$-factor. These expressions are particularly tractable for admitting fast and effective computation. When the Ricean $K$-factor is equal to zero, i.e., Rayleigh fading, our results are substantially simplified. Also, based on the proposed expressions, we investigate in detail the performance of the effective SINR precisely by making either the $M$ or the Ricean $K$-factor becomes infinite. In both cases, it is shown that the effect of pilot contamination cannot be removed for both LS and MMSE estimation methods; however, if both the $M$ and Ricean $K$-factor continue to grow unbound, the pilot contamination issue is eliminated only for the MMSE estimation method. Finally, a set of Monte-Carlo simulations is conducted to validate the above mentioned analytical results.

Notation: Lower-case and upper-case boldface letters denote vectors and matrices, respectively; $\mathbb{C}^{M \times N}$ denotes the $M \times N$ complex space; $A^T$, $A^\dagger$, and $A^{-1}$ denote the transpose, the Hermitian transpose, and the inverse of the matrix $A$, respectively; $\mathbf{I}_M$ denotes an $M \times M$ identity matrix and $\mathbf{0}_{M \times N}$ denotes an $M \times N$ zero matrix. The expectation operation is $E\{\cdot\}$. A complex Gaussian random vector $\mathbf{x}$ is denoted as $\mathbf{x} \sim \mathcal{C}\mathcal{N}(\bar{x}, \Sigma)$, where the mean vector is $\bar{x}$ and the covariance matrix is $\Sigma$, while $\|\cdot\|_2$ denotes the 2-norm of a vector. Finally, $\text{diag}(a)$ denotes a diagonal matrix where the main diagonal entries are the elements of vector $a$.

II. System Model

In this paper, we consider a typical uplink cellular communication system with $L$ hexagonal cells. Each cell contains a BS and $N$ single-antenna users. Each BS has a uniform linear array with $M (M \gg N)$ antennas. In the following, the time-division duplex (TDD) mode is adopted and we assume that the BSs and users in this system are perfectly synchronized. For each channel use, the $M \times 1$ received signal vector of the BS in cell $j$ is given by

$$y_j = \sqrt{\rho_u} \sum_{l=1}^{L} \mathbf{H}_{jl} \mathbf{x}_l + \mathbf{n}_j,$$

where $\mathbf{x}_l$ denotes the $N \times 1$ vector containing the transmitted signals from all the users in cell $l$, which satisfies $E\{\mathbf{x}_l\} = \mathbf{0}_{N \times 1}$ and $E\{\mathbf{x}_l\mathbf{x}_l^\dagger\} = \mathbf{I}_N$, $\rho_u$ is the average transmitted power of each user, and $\mathbf{n}_j \in \mathbb{C}^{M \times 1} \sim \mathcal{C}\mathcal{N}(\mathbf{0}_{M \times 1}, \mathbf{I}_M)$ is the additive white Gaussian noise (AWGN) vector.

Also, $\mathbf{H}_{jl} \in \mathbb{C}^{M \times N}$ represents the channel matrix between the users in cell $l$ and the BS in cell $j$, whose $n$th column, $\mathbf{h}_{jln} \in \mathbb{C}^{M \times 1}$, is the channel vector between the user $n$ in cell $l$ and the BS in cell $j$. Here, the block-fading model [3] is utilized where the large-scale fading coefficients are kept fixed over lots of coherence time intervals and the small-scale fading coefficients remain fixed within a coherence time interval and change between any two adjacent coherence time intervals. Moreover, the large-scale fading coefficients are assumed to be perfectly known at the BS side due to their slow-varying nature. We herein consider the Ricean fading model in [9] where both a LOS path and a non LOS (NLOS) component exist in the channels between the users and BS in the same cell, while only a NLOS component exists in the channels between the users and BS in different cells. The above mentioned model is reasonable since a LOS component is more likely to kick in when the users and the BS are in the same cell. Hence, $\mathbf{h}_{jln}$ can be rewritten as

$$\mathbf{h}_{jln} = \begin{cases} \sqrt{\frac{K_{jn}}{K_{jn}+1}} \mathbf{h}_{jln,\text{LOS}} + \sqrt{\frac{1}{K_{jn}+1}} \mathbf{h}_{jln,\text{NLOS}}, & l = j, \\ \mathbf{h}_{jln,\text{NLOS}}, & l \neq j, \end{cases}$$

where $K_{jn}$ denotes the Ricean $K$-factor for the user $n$ in cell $l$, $\mathbf{h}_{jln,\text{LOS}} \sim \mathcal{C}\mathcal{N}(\mathbf{0}_{M \times 1}, \beta_{jln} \mathbf{I}_M)$ (forall, $n$) denotes the NLOS component between the user $n$ in cell $l$ and the BS in cell $j$, while $\beta_{jln}$ is the corresponding large-scale fading coefficient. Also, $\mathbf{h}_{jln,\text{LOS}} \in \mathbb{C}^{M \times 1}$ denotes the LOS part between the user $n$ and the BS in cell $j$, whose $m$th entry $[\mathbf{h}_{jln,\text{LOS}}]_m$ is given by

$$[\mathbf{h}_{jln,\text{LOS}}]_m = \beta_{jln}^2 \rho_{jn} e^{-j(m-1)\frac{2\pi}{\lambda}d \sin(\theta_{jn})},$$

where $\lambda$ is the wavelength, $d$ is the antenna spacing, $\theta_{jn} \in [0, 2\pi)$ is the angle of arrival of the $n$th user in cell $j$, and $i$ denotes imaginary unit. Based on the fact that $K_{jn}$ and $\theta_{jn}$ can be obtained through a feedback link [14], we can also assume that the Ricean $K$-factor and LOS path can be perfectly obtained by the BS and users.

In practical communication systems, the BS side does not have perfect CSI and the channel needs to be estimated at the BS. In TDD mode, uplink pilot training is adopted for obtaining the estimated channel information before data transmission. The worst-case of pilot sequence allocation is adopted here, where each cell’s users utilize the same set of pilot sequences [4], which we denote as $\Phi \in \mathbb{C}^{\tau \times N}$. Also, $\tau$ is the length of the pilot sequence, which is larger than or equal to $N$, and $\Phi$ satisfies $\Phi^\dagger \Phi = \mathbf{I}_N$. Then, the $M \times \tau$ received pilot sequence signal matrix at the BS in cell $j$ is given by

$$\mathbf{Y}_{j,\text{train}} = \sum_{l=1}^{L} \mathbf{H}_{jl}(\Omega_l + \mathbf{I}_N)^\frac{1}{2} \mathbf{P}_l^\frac{1}{2} \Phi^\dagger + \mathbf{N}_j,$$

where $\mathbf{P}_l$ is the $N \times N$ diagonal matrix with the power of the pilot sequence sent by user $n$ in cell $l$ is $[\mathbf{P}_l]_{nn} = \rho_{tn}$, $\mathbf{N}_j \in \mathbb{C}^{M \times \tau}$ is the AWGN matrix with the independent and identically distributed zero-mean and unit-variance elements, and $\Omega_l = \text{diag}([K_1, \ldots, K_{ln}, \ldots, K_{1N}]) \in \mathbb{C}^{\tau \times N}$. Particularly, the reason for the pilot matrix $\mathbf{P}_l^\frac{1}{2} \Phi^\dagger$ multiplied by $(\Omega_l + \mathbf{I}_N)^\frac{1}{2}$, which is sent by all users in cell $j$, is to estimate the NLOS component in a simplified manner; note that a similar method followed in [6], [9]. By removing the known LOS part in (4) and utilizing the LS and MMSE estimation methods [4], [15], $\hat{\mathbf{h}}_{jln,\text{NLOS}}$ is estimated as

$$\hat{\mathbf{h}}_{jln,\text{NLOS}} = \frac{1}{\sqrt{\rho_{jn}}} \mathbf{h}_{jln,\text{NLOS}} + \hat{\mathbf{n}}_{jn},$$

$$\hat{\mathbf{h}}_{jln,\text{NLOS}} = \chi_{jn} \hat{\mathbf{h}}_{jln,\text{NLOS}},$$

(5)
respectively, where $\hat{h}_{jn,NLOS}^{LS}$ and $\hat{h}_{jn,NLOS}^{MMSE}$ denote the estimators of $h_{jn,NLOS}$ based on the LS and MMSE estimation methods, respectively, as well as, $\tilde{h}_{jn} = N_j \phi_n$. Also, $\phi_n$ is the $n$th column of $\Phi$ and

$$\chi_{jn} = \frac{\rho_{jn} \beta_{jn}}{\rho_{jn} \beta_{jn} + \sum_{l \neq j} \rho_l (K_n + 1) \beta_{jl} + 1}.$$  \hfill (7)

For convenience, we denote the estimator of $h_{jn,NLOS}$ as $\hat{h}_{jn,NLOS}$ and the estimator of $h_{jn}$ as

$$\hat{h}_{jn} = \sqrt{\frac{K_n}{K_n + 1}} h_{jn,NLOS} + \sqrt{\frac{1}{K_n + 1}} \hat{h}_{jn,NLOS},$$  \hfill (8)

respectively, for both the LS and MMSE estimation methods.

After channel estimation, we use the standard linear detector MRC [4] to detect the received data signal in (1). Hence, for the BS in cell $j$, (1) is separated into $N$ streams by multiplying it with the MRC detector, that is,

$$r_j = \hat{H}_j y_j \in \mathbb{C}^{N \times 1},$$  \hfill (9)

where $\hat{H}_j \triangleq [\hat{h}_{j1}, \ldots, \hat{h}_{jn}, \ldots, \hat{h}_{jN}] \in \mathbb{C}^{M \times N}$. Then, for the $n$th user in cell $j$, we have

$$r_{jn} = \sqrt{\rho_n} \hat{h}_{jjn}^\dag x_{jn} + \sqrt{\rho_u} \sum_{(l,t) \neq (j,n)} \hat{h}_{jjn}^\dag h_{jt} x_{lt} + \hat{h}_{jjn}^\dag x_{jn} \in \mathbb{C}^{N \times 1}.$$  \hfill (10)

In this section, we obtain the closed-form expressions of the effective SINR for both LS and MMSE estimation methods and, thereafter, aim to analyze the SINR performance with the respect of the number of BS antennas $M$ and the Ricean $K$-factor, respectively.

### A. Closed-Form of SINR$_{jn}$

Since we want to obtain a computable expression of the achievable uplink SE and investigate it by a simple way, it is convenient to follow the methodology of [4] that assumes that the term $\mathbb{E}\{\hat{h}_{jjn}^\dag h_{jjn}\}$ is known at the BS in cell $j$ perfectly. Hence, (10) can be rewritten as

$$r_{jn} = \sqrt{\rho_n} \mathbb{E}\{\hat{h}_{jjn}^\dag h_{jjn}\} x_{jn} + \sqrt{\rho_u} \sum_{(l,t) \neq (j,n)} \hat{h}_{jjn}^\dag h_{jt} x_{lt} + \sqrt{\rho_u} \hat{h}_{jjn}^\dag h_{jjn} x_{jn} + \hat{h}_{jjn}^\dag x_{jn} \in \mathbb{C}^{N \times 1}.$$  \hfill (11)

Then, by using the definition of the effective SINR in multicell massive MIMO systems as in [4, Eq. (18)], the achievable uplink SE of the $n$th user in cell $j$, in units of bit/s/Hz, is given by

$$R_{jn} = \frac{T - \tau}{T} \log_2 (1 + \text{SINR}_{jn}).$$  \hfill (12)

where $T$ denotes the channel coherence time interval, in terms of the number of symbols, while $\tau$ symbols are utilized for channel estimation, and the SINR$_{jn}$ is defined as (13) (at the top of the next page).

The following theorem presents a new general framework for the closed-form expression of SINR$_{jn}$, which applies for both the LS and MMSE estimation methods. This constitutes a key contribution of this paper.

**Theorem 1:** The exact SINR$_{jn}$, for both the LS and MMSE estimation methods, can be analytically evaluated as

$$\text{SINR}_{jn} = \begin{cases} \text{SINR}_{jn}^{LS}, & \text{LS}, \\ \text{SINR}_{jn}^{MMSE}, & \text{MMSE}, \end{cases}$$  \hfill (14)

where

$$\text{SINR}_{jn}^{LS} \triangleq \frac{M \rho_{jn} (K_n + 1) \beta_{jn}^2}{M \psi_{jn} + \zeta_{jn} \theta_j + \rho_n \hat{h}_{jn,NLOS}^\dag \psi_{jn} - \rho_n \hat{h}_{jn,NLOS}^\dag \psi_{jn}},$$  \hfill (15)

and $\text{SINR}_{jn}^{MMSE}$ is defined as in (16) (at the top of the next page). Also, $\psi_{jn}$, $\zeta_{jn}$, $\theta_j$, and $\varsigma_{jn}$ are denoted as

$$\psi_{jn} \triangleq \sum_{l \neq j} \rho_l (K_n + 1) \beta_{jl},$$  \hfill (16)

$$\zeta_{jn} \triangleq \sum_{l \neq j} \rho_c (K_{jn} + 1) \beta_{jc} + 1,$$  \hfill (17)

$$\theta_j \triangleq \sum_{l = 1}^L \sum_{n = 1}^N \beta_{jl} + \frac{1}{\rho_u},$$  \hfill (18)

$$\varsigma_{jn} \triangleq \sum_{l \neq j} \sum_{l = 1}^L \sum_{n = 1}^N \frac{K_{jt}}{K_{jt} + 1} - \frac{2}{\rho_u},$$  \hfill (19)

respectively, where

$$\phi_{nt} \triangleq \frac{\sin \left( \frac{M\pi}{2} \left( \sin(\theta_{jn}) - \sin(\theta_{jn}) \right) \right)}{\sin \left( \frac{\pi}{2} \left( \sin(\theta_{jn}) - \sin(\theta_{jn}) \right) \right)}.$$  \hfill (20)

**Proof:** See Appendix A.

It is important to note that the expressions in Theorem 1 can be easily evaluated since they involve only the pilot sequence power, uplink data power, Ricean $K$-factor, and large-scale fading coefficients, for all cases of interest. Moreover, from Theorem 1, we see that the obtained effective SINRs based on the LS and MMSE channel estimation methods with MRC detector are different. Note that when $K_{jn} = 0 (\forall l, n)$, SINR$_{jn}$ reduces to the special case of Rayleigh fading channel. After performing some simplifications, for both LS and MMSE estimation methods, we have

$$\text{SINR}_{\text{Rayleigh},jn} = \frac{M \rho_{jn} \beta_{jjn}^2}{M \sum_{l \neq j} \rho_l \beta_{jn}^2 + \sum_{l = 1}^L \rho_c (K_{jn} + 1) \beta_{jc} + 1 - \theta_j},$$  \hfill (21)

Interestingly, (22) is the effective SINR in Rayleigh fading channels given by [4, Theorem 1]. Note that Theorem 1 gives a universal formula for the SINR$_{jn}$ when Ricean fading is considered.
\[
\text{SINR}_{jn} \triangleq \frac{\rho_u \mathbb{E} \left\{ |\hat{h}_{jn}^j h_{jn}^j| \right\}^2}{\rho_u \sum_{l=1}^{L} \sum_{t=1}^{N} \mathbb{E} \left\{ |\hat{h}_{jn}^l h_{jn}^l| \right\} - \rho_u \mathbb{E} \left\{ |\hat{h}_{jn}^j h_{jn}^j| \right\}^2 + \mathbb{E} \left\{ |\hat{h}_{jn}^j h_{jn}^j| \right\}^2}.
\]

(13)

\[
\text{SINR}^{\text{MMSE}}_{jn} \triangleq \frac{M \rho_j (K_j + \chi_j)^2 \beta_{jn}^2}{M \chi_j^2 (K_j + 1) \psi_j + \rho_j (K_j + \chi_j) (K_j + 1) \beta_{jn}^2 \varphi_j + \rho_j K_j (K_j + 1) \beta_{jn}^2 \varphi_j}.
\]

(16)

B. Analysis of SINR\(_{jn}\)

Now, we consider the SINR\(_{jn}\) limit when \(M\) grows without bound. To the best of our knowledge, this result is also new.

**Corollary 1:** If \(M \to \infty\), the exact analytical expression of the SINR\(_{jn}\) in (14) approaches

\[
\lim_{M \to \infty} \text{SINR}_{jn} = \begin{cases} 
\text{SINR}^{\text{LS}}_{jn}, & \text{LS,} \\
\text{SINR}^{\text{MMSE}}_{jn}, & \text{MMSE}, 
\end{cases}
\]

(23)

where

\[
\text{SINR}^{\text{LS}}_{jn} \triangleq \frac{\rho_j (K_j + 1) \beta_{jn}^2}{\psi_j},
\]

(24)

\[
\text{SINR}^{\text{MMSE}}_{jn} \triangleq \frac{\rho_j (K_j + \chi_j)^2 \beta_{jn}^2}{\chi_j^2 (K_j + 1) \psi_j}.
\]

(25)

**Proof:** The proof is completed by calculating the limit of (14) when \(M \to \infty\).

**Corollary 1** indicates that if the number of users is kept fixed and the number of receive antennas at the BS side is increased, then, the asymptotic SINR\(_{jn}\) is saturated. Intuitively, this is due to the pilot contamination since the other cells’ users adopt the same pilot sequence as the user in the target cell \(j\). It is also worth noting that, based on (24) and (25), if \(K_{ln} = 0, \forall l, n\), the limit of SINR\(_{jn}\) as \(M \to \infty\) for both LS and MMSE estimation methods is given by

\[
\text{SINR}_{\text{Rayleigh},jn} = \frac{\rho_j \beta_{jn}^2}{\sum_{t \neq j}^{L} \rho_t \beta_{jn}^2}.
\]

(26)

Again, it is important to note that (26) is the limit of effective SINR with imperfect CSI in Rayleigh fading channels given by [4, Corollary 1]. Hence, (26) is a special case of (23) if the power of the LOS part of the Ricean fading channel is equal to zero. To gain more insights, the exact SINR\(_{jn}\) admits further simplifications in the large Ricean \(K\)-factor regime for both LS and MMSE estimation methods.

**Corollary 2:** If for any \(l\) and \(n\), \(K_{ln} = K \to \infty\), (14) converges to

\[
\lim_{K \to \infty} \text{SINR}_{jn} = \begin{cases} 
\text{SINR}^{\text{LS}}_{jn}, & \text{LS,} \\
\text{SINR}^{\text{MMSE}}_{jn}, & \text{MMSE}, 
\end{cases}
\]

(27)

where \(\text{SINR}^{\text{LS}}_{jn}\) and \(\text{SINR}^{\text{MMSE}}_{jn}\) are defined as

\[
\text{SINR}^{\text{LS}}_{jn} \triangleq \frac{M \rho_j \beta_{jn}^2}{M \sum_{l \neq j}^{L} \rho_l \beta_{jn}^2 + \sum_{t = 1}^{L} \rho_t \beta_{jn} \varphi_j + \rho_j \beta_{jn} \varphi_j},
\]

(28)

\[
\text{SINR}^{\text{MMSE}}_{jn} \triangleq \frac{M \beta_{jn}}{M \sum_{t = 1}^{L} \sum_{l \neq j}^{L} \beta_{jt} + \sum_{t = 1}^{L} \beta_{jt} \varphi_j},
\]

(29)

respectively. Also, \(\varphi_j\) is denoted as

\[
\varphi_j \triangleq \sum_{t = 1}^{L} \frac{\beta_{jt}^2}{M},
\]

(30)

**Proof:** The proof is completed by calculating the limit of (14) when \(K \to \infty\).

It is interesting to note from **Corollary 2** that as \(K\) increases, the SINR based on LS and MMSE will approach two constant values, respectively. Note that (29) is unbounded as \(M \to \infty\), whereas (28) is bounded for the same conditions. In other words, in the high Ricean \(K\)-factor regime with infinite \(M\), the pilot contamination effect will be completely eliminated for MMSE estimation, since this scheme accounts for the presence of a LOS component in (6). On the other hand, pilot contamination effect cannot be be removed when using the LS estimation method since the LS estimation method regards the co-channel interference in (5) as just noise. Note that, **Corollary 2** reflects that massive MIMO systems have unlimited achievable uplink SE via different estimation methods, the number of BS antennas, and Ricean \(K\)-factor, though in a slightly different way than in [16].

C. Numerical Results

In this section, we consider a hexagonal cellular network with a set of \(L\) cells and radius (from center to vertex) \(\varpi\) meters where users are distributed uniformly in each cell. Also, leveraging the large-scale fading model of [17], the large-scale fading coefficient between the user \(n\) in cell \(l\) and the BS in cell \(j\), \(\beta_{jn}\), is given by

\[
\beta_{jn} = \frac{v_{jn}}{1 + \left(\frac{\eta_{jn}}{\eta_{\text{min}}}\right)^\alpha},
\]

(31)

where \(v_{jn}\) is a log-normal random variable with standard deviation \(\xi\), \(\alpha\) is the path loss exponent, \(\eta_{jn}\) is the distance between the user \(n\) in cell \(l\) and the BS in cell \(j\), and \(\eta_{\text{min}}\) is
We see that in all cases the analytical curves (based on (14)) is equal to 30dB, and the uplink data power
of users’ large-scale fading coefficients.

In the following, we assess the accuracy of the achievable uplink SE given by (12) for both LS and MMSE estimation methods, respectively. Theorem 1
shows the validity of the results (based on (22)) in previous literature [4], but also does prove that Theorem 1 can be applied into the Rayleigh fading environment. Moreover, in Ricean fading conditions, the results in this figure show that the MMSE performance is always better than the LS.

Fig. 1(b) investigates the analytical results of sum achievable uplink SE \( R_{\text{sum}} \) in the whole \( M \) regime under the same parameter setting in Fig. 1(a). When \( M \to \infty \), we see that all \( R_{\text{sum}} \) results tend to different constants, which match the asymptotic expression (based on (23)) for different Ricean \( K \)-factor, respectively. In other words, it justifies the effectiveness of Corollary 1. Also, we note that, for the LS case with \( K = 0(-\infty \text{dB}) \), \( 3 \text{dB} \), \( 6 \text{dB} \), and \( 10 \text{dB} \), as well as, the MMSE case with \( K = 0(-\infty \text{dB}) \), the asymptotic results are identical. This phenomenon is caused by the following two reasons. First, if all users’ Ricean \( K \)-factors are identical, (24) is uncorrelated with the Ricean \( K \)-factor. In other words, for the LS case, different Ricean \( K \)-factor means the same asymptotic SE. Second, when Ricean \( K \)-factor is equal to zero, both the LS and MMSE cases have the same asymptotic SE since the current channel becomes the Rayleigh fading channel. Hence, for both the LS and MMSE estimation methods, the pilot contamination exists such that the \( R_{\text{sum}} \) saturates even when \( M \to \infty \).

In Fig. 2, we investigate the impact of the Ricean \( K \)-factor on the sum achievable uplink SE performance for \( M = 125, 250, 500, \) and \( 1000 \), with the LS and MMSE estimation

\( \eta \) and \( \omega \) denote the reference distance. In our simulations, we choose \( \omega = 500 \text{m} \), \( \xi = 8 \text{dB} \), \( \alpha = 3.8 \), \( d = \frac{\lambda}{\lambda} \), and \( \eta_{\text{min}} = 200 \text{m} \), which follow the methodology of [6], [17]. Here, the observed cell is the center cell and call it cell 1, which is given as

\[ R_{\text{sum}} \triangleq \sum_{n=1}^{N} R_{1n} \]  

which proves the validity of Theorem 1. Moreover, for all cases, when \( M \) increases, \( R_{\text{sum}} \) increases. Also, when the power of the LOS path becomes zero, \( R_{\text{sum}} \) for both the LS and MMSE estimation methods are identical for both the analytical and simulated curves, respectively, which not only shows the validity of the results (based on (22)) in previous literature [4], but also does prove that Theorem 1 can be applied into the Rayleigh fading environment. Moreover, in Ricean fading conditions, the results in this figure show that the

we define the metric called “Sum achievable uplink SE” in target cell 1, which is given as

the results in Fig. 1(b).

the users in same cell have the same Ricean \( K \)-factor, denoted by \( K \), and \( \tau = N = 10 \) OFDM symbols. Finally, all the simulation results are obtained by averaging 100 realizations of all users’ large-scale fading coefficients in all cells over 100 independent small-scale fading channels for each realization of users’ large-scale fading coefficients.

In the following, we assess the accuracy of the achievable uplink SE given by (12) for both LS and MMSE estimation methods, the closed-form expression given in Theorem 1, and the results in Corollary 1 and Corollary 2. For comparison, we define the metric called “Sum achievable uplink SE” in target cell 1, which is given as

Fig. 1(a) gives the analytical and Monte-Carlo simulated sum achievable uplink SE \( R_{\text{sum}} \), with the LS and MMSE estimation methods, respectively, in moderate-to-high \( M \) regime. Results are shown for different Ricean \( K \)-factor, and pilot sequence power \( \rho_p = 30 \text{dB} \) with uplink data power \( \rho_u = 20 \text{dB} \). We see that in all cases the analytical curves (based on (14)) match precisely with the simulated curves (based on (13)), which proves the validity of Theorem 1. Moreover, for all cases, when \( M \) increases, \( R_{\text{sum}} \) increases. Also, when the power of the LOS path becomes zero, \( R_{\text{sum}} \) for both the LS and MMSE estimation methods are identical for both the analytical and simulated curves, respectively, which not only shows the validity of the results (based on (22)) in previous literature [4], but also does prove that Theorem 1 can be applied into the Rayleigh fading environment. Moreover, in Ricean fading conditions, the results in this figure show that the MMSE performance is always better than the LS.

Fig. 1(b) investigates the analytical results of sum achievable uplink SE \( R_{\text{sum}} \) in the whole \( M \) regime under the same parameter setting in Fig. 1(a). When \( M \to \infty \), we see that all \( R_{\text{sum}} \) results tend to different constants, which match the asymptotic expression (based on (23)) for different Ricean \( K \)-factor, respectively. In other words, it justifies the effectiveness of Corollary 1. Also, we note that, for the LS case with \( K = 0(-\infty \text{dB}) \), \( 3 \text{dB} \), \( 6 \text{dB} \), and \( 10 \text{dB} \), as well as, the MMSE case with \( K = 0(-\infty \text{dB}) \), the asymptotic results are identical. This phenomenon is caused by the following two reasons. First, if all users’ Ricean \( K \)-factors are identical, (24) is uncorrelated with the Ricean \( K \)-factor. In other words, for the LS case, different Ricean \( K \)-factor means the same asymptotic SE. Second, when Ricean \( K \)-factor is equal to zero, both the LS and MMSE cases have the same asymptotic SE since the current channel becomes the Rayleigh fading channel. Hence, for both the LS and MMSE estimation methods, the pilot contamination exists such that the \( R_{\text{sum}} \) saturates even when \( M \to \infty \).

In Fig. 2, we investigate the impact of the Ricean \( K \)-factor on the sum achievable uplink SE performance for \( M = 125, 250, 500, \) and \( 1000 \), with the LS and MMSE estimation
methods, respectively. In this figure, the pilot sequence power is 30dB and the uplink data power is 20dB. It shows that the analytical values and simulation values are almost indistinguishable for both the LS and MMSE estimation methods, respectively. If both the Ricean $K$-factors become infinite or $M \to \infty$, it can be shown that the pilot contamination phenomenon disappeared for the MMSE case.

IV. CONCLUSION

In this paper, a detailed statistical characterization of the SE for the multi-cell massive MIMO system with Ricean fading was presented. In order to evaluate the SE performance, we first proposed two exact closed-form expressions for the effective SINR based on LS and MMSE estimation methods, respectively, which also can be adopted in Rayleigh fading. Then, we analyzed the asymptotic properties of the effective SINR when the number of BS antennas $M$ and the Ricean $K$-factor became infinite. It was shown that, when the Ricean $K$-factor became infinite or $M \to \infty$, the SINR performance for the LS and MMSE estimation methods was saturated, which underlines the pilot contamination phenomenon. However, if both the Ricean $K$-factor and $M$ grow asymptotically large, the pilot contamination phenomenon disappeared for the MMSE estimation method, but persists for the LS estimation method.

APPENDIX A

PROOF OF THEOREM 1

To evaluate the SINR$_{jk}$ in (13), we define six terms

\begin{align}
\mathfrak{A} & \triangleq \mathbb{E} \left\{ \left| \hat{h}_{jjn} h_{jjn} \right|^2 \right\}, \\
\mathfrak{B} & \triangleq \mathbb{E} \left\{ \left| \hat{h}_{jjn} \right|^2 ; (t \neq n) \right\}, \\
\mathfrak{C} & \triangleq \mathbb{E} \left\{ \left| \hat{h}_{jjn} \right|^2 ; (l \neq j) \right\}, \\
\mathfrak{D} & \triangleq \mathbb{E} \left\{ \left| \hat{h}_{jjn} \right|^2 ; (l \neq j \& t \neq n) \right\}, \\
\mathfrak{E} & \triangleq \mathbb{E} \left\{ \| \hat{h}_{jjn} \|^2 \right\}, \\
\mathfrak{F} & \triangleq \mathbb{E} \left\{ \| h_{jjn} \|^2 \right\}.
\end{align}

Although $\hat{h}_{jjn}$ has different expressions for the LS and MMSE estimation methods, the corresponding proofs for SINR$_{jn}$ are similar. Hence, it is convenient to only study the case of the LS estimation.

- Calculate $\mathfrak{A}$: Substituting (3), (5) and (8) into (33), after some algebraic manipulations, it can be shown that $\mathfrak{A}$ reduces to

\begin{equation}
\mathfrak{A} = M^2 \beta_{jjn}^2.
\end{equation}

- Calculate $\mathfrak{B}$: Substituting (2), (5), and (8) into (34), after some manipulations, it is easy to obtain

\begin{align}
\mathfrak{B} & = \mathbb{E} \left\{ \left\| h_{jjn} \right\|^2 \right\} \\
& \times \frac{1}{K_{jn} + 1} + \frac{\mathbb{E} \left\{ \| h_{jjn} \|^2 \right\}}{B_2},
\end{align}

where the closed-form expression of $B_1$ can be obtained based on $h_{jjn}$ is uncorrelated with the rest of terms in $B_1$ and the distribution of $h_{jjn}$. Also, $B_2$ can be obtained based on the properties of non-central Wishart matrices [6, Eq. (123)]. After some algebraic manipulations, we write $\mathfrak{B}$ as follows

\begin{align}
\mathfrak{B} & = M^2 \beta_{jjn}^2 + M \beta_{jjn}^2 \frac{K_{jn}}{(K_{jn} + 1)^2} \\
& + M \beta_{jjn} \sum_{\ell \neq j} \rho_{jn} \frac{K_{jn} + 1}{(K_{jn} + 1)^2}.
\end{align}

- Calculate $\mathfrak{C}$: Since the $\hat{h}_{jjn}$ is uncorrelated with $h_{jjt}$ when $t \neq n$, we substitute (2), (5), and (8) into $\mathfrak{C}$, as

\begin{align}
\mathfrak{C} & \triangleq \mathbb{E} \left\{ \left| \hat{h}_{jjn} \right|^2 ; (l \neq j \& t \neq n) \right\}.
\end{align}

In this figure, the asymptotic case when Ricean $K$-factor approaches to 0$(-\infty$ dB) has not been shown since the Rayleigh fading case has been examined in Fig. 1.
where \( \phi_{nt} \) has been defined as in (21).

- Calculate \( \mathcal{D} \): Substituting (2), (5), and (8) into (36), and performing some basic simplifications to obtain

\[
\mathcal{D} = \mathbb{E} \left\{ |D_1|^2 \right\} + \mathbb{E} \left\{ |D_2|^2 \right\},
\]

where \( D_1 \) is denoted as in (44) (at the top of this page) and \( D_2 \) is denoted as

\[
D_2 \triangleq \left( \frac{\rho_{ln} (K_{jn} + 1)}{\rho_jn (K_{jn} + 1)} \right)^{1/2} \| \mathbf{h}_{jn, \text{NLOS}} \|^2_2,
\]

respectively. Based on the similar way for obtaining \( \mathcal{C} \), \( \mathbb{E} \left\{ |D_1|^2 \right\} \) reduces to

\[
\mathbb{E} \left\{ |D_1|^2 \right\} = M \beta_{jn} \sum_{c \neq j} L \rho_{cn} (K_jn + 1) \left( 1 + \beta_{jn} \right) \rho_{jn} (K_jn + 1)
\]

By using the properties of Wishart matrices [18, Lemma 2.9], we obtain

\[
\mathbb{E} \left\{ |D_2|^2 \right\} = M \beta_{jn} \rho_{jn} (K_jn + 1) M (M + 1) \beta_{jn}.
\]

Therefore, substituting (46) and (47) into (43) and simplifying, we get

\[
\mathcal{D} = M^2 \beta_{jn}^2 \sum_{c \neq j} L \rho_{cn} (K_jn + 1) \rho_{jn} (K_jn + 1) + M \beta_{jn} \beta_{jt} \rho_{jn} (K_jn + 1)
\]

- Calculate \( \mathcal{E} \): Based on the similar way for obtaining \( B_1 \) and \( \mathcal{C} \), \( \mathcal{E} \) is given by

\[
\mathcal{E} = M \beta_{jn} \beta_{jt} \rho_{jn} (K_jn + 1) + M \beta_{jn} \beta_{jt} \rho_{jn} (K_jn + 1)
\]

- Calculate \( \mathcal{E} \): With the help of (3), (5), and (8), we get

\[
\mathcal{E} = M \beta_{jn} + M \beta_{jn} \beta_{jt} \rho_{jn} (K_jn + 1)
\]

Finally, substituting (39), (41), (42), and (48)-(50) into (13) and simplifying, the closed-form expression for \( \text{SINR}_{jn}^{\text{LOS}} \) is obtained.

**References**

[1] Qualcomm, “The 1000x data challenge,” [Online]. Available: https://www.qualcomm.com/1000x

[2] T. L. Marzetta, “Noncooperative cellular wireless with unlimited numbers of base station antennas,” IEEE Trans. Wireless Commun., vol. 9, no. 11, pp. 3590-3600, Nov. 2010.

[3] H. Q. Ngo, E. G. Larsson, and T. L. Marzetta, “Energy and spectral efficiency of very large multiuser MIMO systems,” IEEE Trans. Commun., vol. 61, no. 4, pp. 1346-1449, Apr. 2013.

[4] P. Liu, S. Jin, T. Jiang, Q. Zhang, and M. Matthaiou, “Pilot power allocation through user grouping in multi-cell massive MIMO systems,” IEEE Trans. Commun., vol. 65, no. 4, pp. 1561-1574, Apr. 2017.

[5] P. Liu, K. Luo, D. Chen, and T. Jiang, “Spectral efficiency analysis in cell-free massive MIMO systems with arbitrary-rank forcing detector,” IEEE Trans. Commun., 2018, submitted. [Online]. Available: http://arxiv.org/abs/1805.10621

[6] Q. Zhang, S. Jin, K.-K. Wong, H. Zhu, and M. Matthaiou, “Power scaling of uplink massive MIMO systems with arbitrary-rank channel means,” IEEE J. Sel. Topics Signal Process., vol. 8, no. 5, pp. 966-981, Oct. 2014.

[7] H. Q. Ngo, E. G. Larsson, and T. L. Marzetta, “The multicell multiuser MIMO uplink with very large antenna arrays and a finite-dimensional channel,” IEEE Trans. Commun., vol. 61, no. 6, pp. 2350-2361, Jun. 2013.

[8] E. Björnson, E. G. Larsson, and M. Debbah, “Massive MIMO for maximal spectral efficiency: How many users and pilots should be allocated?” IEEE Trans. Wireless Commun., vol. 15, no. 2, pp. 1293-1308, Feb. 2016.

[9] Q. Zhang, S. Jin, Y. Huang, and H. Zhu, “Uplink rate analysis of multicell massive MIMO systems in Ricean fading,” in Proc. IEEE GLOBECOM, Dec. 2014, pp. 3279-3284.

[10] L. Wu, Z. Zhang, J. Dang, J. Wang, H. Liu, and Y. Yu, “Channel estimation for multicell multiuser massive MIMO uplink over Rician fading channels,” IEEE Trans. Veh. Technol., vol. 66, no. 10, pp. 8872-8882, Oct. 2017.

[11] L. Sanguinetti, A. Kammoun, and M. Debbah, “Asymptotic analysis of multicell massive MIMO over Rician fading channels,” in Proc. IEEE ICASSP, Mar. 2017, pp. 3539-3543.

[12] O. Ozdogan, E. Björnson, and E. G. Larsson, “Massive MIMO with spatially correlated Rician fading channels,” IEEE Trans. Commun., 2018, submitted. [Online]. Available: https://arxiv.org/abs/1805.07972

[13] ——, “Uplink spectral efficiency of massive MIMO with spatially correlated Rician fading,” in Proc. IEEE SPAWC, Jun. 2018. [Online]. Available: https://arxiv.org/abs/1805.07968

[14] M. A. Maddah-Ali and T. Tse, “Completely stale transmitter channel state information is still very useful,” IEEE Trans. Inf. Theory, vol. 58, no. 7, pp. 4418-4432, Jul. 2012.

[15] S. M. Kay, Fundamentals of Statistical Signal Processing: Estimation Theory. Prentice-Hall, Inc. Upper Saddle River, NJ, USA, 1993.

[16] E. Björnson, J. Hoydis, and L. Sanguinetti, “Massive MIMO has unlimited capacity,” IEEE Trans. Wireless Commun., vol. 17, no. 1, pp. 574-590, Jan. 2018.

[17] H. Q. Ngo, H. A. Suraweera, M. Matthaiou, and E. G. Larsson, “Multipair full-duplex relaying with massive arrays and linear processing,” IEEE J. Sel. Areas Commun., vol. 32, no. 9, pp. 1721-1737, Sept. 2014.

[18] A. M. Tulino and S. Verdú, “Random matrix theory and wireless communications,” Found. Trends Commun. Inf. Theory, vol. 1, no. 1, pp. 1-182, Jun. 2004.