Communication Using a Large-Scale Array of Ubiquitous Antennas: A Geometry Approach
Kaibin Huang, Senior Member, IEEE, Jiayi Chen, Member, IEEE, and Vincent K. N. Lau, Fellow, IEEE

Abstract—The recent trends of densification and centralized signal processing in radio access networks suggest that future networks may comprise ubiquitous antennas coordinated to form a network-wide gigantic array, referred to as the ubiquitous array (UA). In this paper, the UA communication techniques are designed and analyzed based on a geometric model. Specifically, the UA is modeled as a continuous circular/spherical array enclosing target users and free-space propagation is assumed. First, consider the estimation of multiuser UA channels induced by user locations. Given single pilot symbols, a novel channel estimation scheme is proposed that decomposes training signals into Fourier/Laplace series and thereby translates multiuser channel estimation into peak detection of a derive function of location. The process is shown to suppress noise. Moreover, it is proved that estimation error due to interference diminishes with the increasing minimum user-separation distance following the power law, where the exponent is 1/3 and 1 for the circular and spherical UA, respectively. If orthogonal pilot sequences are used, channel estimation is found to be perfect. Next, consider channel-conjugate data transmission that maximizes received signal power. The power of interference between two users is shown to decay with the increasing user-separation distance sub-linearly and super-linearly for the circular and spherical UA, respectively. Furthermore, a novel multiuser precoding design is proposed by exciting different phase modes of the UA and controlling the mode weight factors to null interference. The number of available degrees of freedom for interference nulling using the UA is proved to be proportional to the minimum user-separation distance.

Index Terms—Beamforming, channel estimation, distributed antenna systems, free space propagation, massive MIMO, multiuser precoding.

I. INTRODUCTION

The explosive growth of mobile traffic is driving the rapid network densification to provide high-speed wireless access to coverage regions. To rein in the escalating network cost and interference, wireless access networks are evolving towards having an architecture with centralized signal processing and minimum on-site hardware comprising merely antennas and RF units, called either cloud radio access networks or base-station virtualization [1]. The network evolution as well as the advancements of other technologies such as small cells [2] and massive MIMO [3] will lead to future networks where antennas are ubiquitous and have network-wide coordination to form a gigantic array, which is called the ubiquitous array (UA) and forms the theme for the paper.

A. Prior Work and New Challenges

The UA system is equivalent to a distributed antennas system (DAS) with dense antennas and without cells. DASs refer to cellular systems where in each cell, antennas are distributed over the cell region and connected to a central processing unit [4]. This technology was initially developed to reduce transmission power and improve network coverage by either simulcast over all distributed antennas or serving each user using the nearest antennas [5], [6]. Recent research on DASs focuses on increasing the sum rate based on multiuser MIMO transmission using the distributed antennas as a virtual array, addressing issues such as inter-cell interference distribution [7], resource allocation [8], capacity analysis [9]–[11] and multi-cell coordination [12]. Each MIMO channel in such systems has coefficients corresponding to heterogeneous path loss depending on antenna locations and cannot be modeled as i.i.d. random variable as for the case of co-located antennas (see e.g., [13]). This complicates the distributions of signals and interference [7] and provides extra degrees of freedom, namely antenna locations, for sum-rate optimization [14], [15]. Motivated by recent advancements in massive MIMO [3], dense DAS systems are analyzed in [16] for scattering channels and based on a stochastic single-cell model with randomly distributed antennas, users and scatterers. Based on this model, the channel variance matrix is shown to have a finite rank regardless of the number of distributed antennas, providing the possibility of solving the well-known pilot contamination problem. In the current work, dense DAS systems are also designed and analyzed, however, assuming free-space channels and based on a deterministic model with fixed locations for distributed antennas and users. Based a different model from that in [16], the current work quantifies the advantages of extremely dense distributed antennas for channel estimation and multiuser transmission over free-space channels.

Given the proximity between the UA and users, the UA channel is typically over free space or at most contains sparse scatterers. Combining free space channels and ubiquitous antennas shifts the paradigm of MIMO communications in several aspects. First, a new approach is needed for analyzing the
capacity of the UA channel. Rich scattering is commonly assumed in a conventional MIMO channel, allowing the channel to be modeled as a random matrix and its capacity analyzed using probability theory and linear algebra (see e.g., [13]). This approach, however, is unsuitable for the UA channel with free space propagation since the channel capacity depends on the array geometry and the user locations. Thus, analyzing the capacity of the UA channel should rely on an approach merging geometry, electromagnetic wave theory and information theory in the same vein as [17], [18]. Next, despite the massive number of elements in the UA, the UA channel over free space is determined only by a few parameters such as the user locations and this fact can be exploited to dramatically reduce the complexity of channel estimation. In contrast, given rich scattering, deploying more antennas leads to continuous growth of the number of degrees of freedom (DoF) in the channel, which makes channel estimation a key challenge in designing massive MIMO systems [19]. Last, the classic technique for multiuser beamforming for free space channels computes nulls by solving a linear system where the number of variables is equal to that of transmit antennas [20], and thus is inefficient for the UA system with a massive number of transmit antennas. This calls for the design of efficient transmission algorithms for the UA systems.

B. Contributions and Organization

The paper represents the first attempt on designing the UA communication systems and focuses on the signal-processing aspect, namely channel estimation and data transmission. For tractability, the work considers a particular coverage region represented by a simple geometric model which comprises a continuous circular UA communicating with $U$ single-antenna users at fixed locations near the UA center. In the model, propagation is constrained to be within the horizontal plane. The results are subsequently extended to the case with propagation in the three-dimensional space and a continuous spherical UA. Channels are assumed to be free space, narrow band, and reciprocal. The elements of the UA and user antennas are all assumed to be omnidirectional. The layout of the above model is similar to that of some existing ones for single-cell DASs (see e.g., [21]) that, however, assumes discrete antennas and rich scattering. The continuity of the UA, modeling dense antennas, is a typical technique for avoiding consideration of antenna placement (see e.g., [17]). More important, it is instrumental for the new findings and the algorithmic designs as summarized in the sequel.

First, consider estimation of multiuser UA channels using only single pilot symbols. The channels are determined by user locations and thus called location induced channels (LI-channels). The channel responses are non-linear functions of the locations. This makes the conventional linear (minimum-mean-square-error or zero-forcing) estimation unsuitable and the optimal maximum-a-posteriori estimation intractable since it requires solving a set of non-linear equations [22]. To address this issue, a novel low-complexity channel-estimation technique is proposed based on decomposing the receive multiuser circular training signal into a Fourier series for the circular UA or spherical harmonics for the spherical UA. This leads to a derived function of location, called the user-location profile. The proposed estimation method is to detect the locations of the peaks of the profile that yield estimated user locations. The estimation procedure is shown to suppress noise by averaging and incur estimation errors only due to interference between multi-channel estimation. The error is shown to decay with the minimum user-separation distance following the power law with the exponent 1/3 and 1 for the circular and spherical UA, respectively. Therefore, without orthogonal pilot sequences, multiuser channel estimation in the UA system is enabled by sufficiently large user-separation distances instead of differentiation in multiuser angles of arrival as in the conventional MIMO systems (see e.g., [23]). In addition, applying the method to the scenario where users deploy orthogonal pilot sequences leads to almost-perfect channel estimation.

Next, consider channel conjugate data transmission using the UA. The signal-to-interference-and-noise ratios (SINRs) are derived in closed-form. In particular, the power of interference between any two users is shown to be proportional to their separation distance (in wavelength) raised to the power of 2/3 and 2, corresponding to the circular and spherical UA, respectively. Therefore, even given single-user transmission, interference can be suppressed by increasing users’ separation distances. Moreover, the path loss is shown to be inversely proportional to the propagation distance or fixed regardless of the distance for the circular and spherical UAs, respectively. In contrast, the loss for a conventional array with collocated elements is inversely proportional to the squared distance.

Last, a novel low-complexity precoding design is proposed for multiuser transmission using the UA. Specifically, the precoders are designed in the form of Fourier series for the circular UA and spherical harmonics for the spherical UA. Their coefficients are controlled as derived to excite different phase modes of the circular array so as to null multiuser interference. For this sub-optimal design, the number of DoF available for interference nulling is shown to be proportional to the minimum user separation distance or its square for the circular and spherical UAs, respectively.

The reminder of the paper is organized as follows. The UA system model is described in Section II. Algorithms for channel estimation and data transmission for the circular-UA system are presented in Sections III and IV, respectively. The results are extended to the spherical-UA system in Section V. Simulation results are provided in Section VI followed by Appendix A where Bessel functions and spherical harmonics are defined and their properties discussed. Last, Appendix B contains the proofs for lemmas.

II. System Model

As illustrated in Fig. 1, the UA is modeled as either a circular array or a spherical array, denoted as $C_0$ and $S_0$, respectively, having the same radius denoted as $r_0$. The dense UA is assumed to be continuous for tractable analysis but this assumption is relaxed in simulation. The communication system comprises the UA centered at the origin and a set of $U$ single-antenna users enclosed by the UA, represented by their fixed locations.
where \( \lambda \) denotes the carrier wavelength. Based on (1),
\[
  h_u (A) = \frac{1}{\sqrt{4\pi r_0}} e^{-j\frac{\lambda}{2r_0} |X_u - A|} + \frac{\alpha_u}{r_0}, \quad A \in \mathbb{C}^2
\]  

Channel estimation at the UA is assisted by pilot signals transmitted by users. Time is divided into slots with unit symbol durations. Since the effective aperture for a semi-directional antenna is \( \lambda^2/4\pi \) [24], the total pilot signal received at antenna \( A \) in an arbitrary slot, denoted as \( q (A) \), is
\[
  q (A) = \frac{\lambda}{4\pi r_0} \sum_{u=1}^{U} h_u (A) s_u + z (A), \quad A \in \mathbb{C}^2
\]  

where \( s_u \) is a pilot symbol transmitted by user \( u \) and the noise \( z (A) \) is a spatial sample of the additive white Gaussian noise \( \mathcal{C}\mathcal{N} (0, \sigma^2) \) process at location \( A \). The noise processes are assumed to be spatially white as follows.

Assumption 2: The noise processes \( z (X) \) and \( z (Y) \) at two locations \( X \) and \( Y \) are independent if \( X \neq Y \).

Substituting (2) into (3) gives
\[
  q (A) = \frac{\lambda}{4\pi r_0} \sum_{u=1}^{U} e^{-j\frac{\lambda}{2r_0} r_u \cos \psi_u (A)} s_u + z (A) + \frac{\alpha_u}{r_0}
\]  

with \( A \in \mathbb{C}^2 \).

Next, consider downlink data transmission. The data symbol intended for user \( u \) is denoted as \( x_u \) and assumed to be distributed as a \( \mathcal{C}\mathcal{N} (0, 1) \) random variable. The symbol is precoded by a continuous precoder represented by \( f_u : \mathbb{C} \rightarrow \mathbb{C} \).

Let \( P_t \) denote the transmission power per user. Then the incident field at location \( u \) in an arbitrary slot, denoted as \( g (X_u) \), can be written as
\[
  g (X_u) = \sqrt{P_t} \int_A h_u (A) \sum_{k=1}^{U} f_k (A) \, dA \, x_k
\]  

where \( A \in \mathbb{C}^2 \), \( h_u (A) \) is the channel response in (2) and \( \partial A \) represents the circumference of a circular UA (\( A = 0 \)) or the surface area of a spherical UA (\( A = \mathbb{C}^2 \)). Note that the integral in (5) yields the accumulated field due to transmissions by different elements of the continuous UA and the integral term is the field generated by radiation of an infinitesimal region of the UA. It follows that the corresponding received signal is
\[
  y_u = \frac{\lambda}{\sqrt{4\pi}} g (X_u) + z_u, \quad u = 1, 2, \ldots, U
\]  

where \( \{ z_u \} \) are i.i.d. \( \mathcal{C}\mathcal{N} (0, \sigma^2) \) random variables representing channel noise.

Last, the backhaul network coordinating the elements of the UA is assumed to be perfect as follows.

Assumption 3: Signals propagating through the backhaul links connecting the UA antennas encounter no delay and distortion.

The practical issues of backhaul networks, such as delay and limited capacity, can degrade the performance of channel estimation and transmission using the UA. Quantifying the resultant performance loss is outside the scope of this work.
III. COMMUNICATION USING THE CIRCULAR UA: CHANNEL ESTIMATION

Estimation of the LI-channels is to infer the user locations from the training signal, namely
\[ \{q(A) \mid A \in \mathcal{O}\} \Rightarrow \{X_u\} \]
with \(\{q(A)\}\) given in (4). As mentioned, the linear or MAP estimation techniques are intractable since \(q(A)\) is a nonlinear function of \(\{X_u\}\). To address this issue, a simple estimation scheme is proposed in the sub-sections, which reduces channel estimation to the detection of the peaks of a given function.

A. LI-Channel Estimation With Single Pilot Symbols

Consider the scenario where users simultaneously transmit single pilot symbols \(\{s_u\}\) to facilitate channel estimation at the UA. Without loss of generality, assume that the pilot symbols are all ones: \(s_u = 1 \forall u\). Let the training signal \(q(A)\) in (4) be re-denoted as \(q(\varphi)\) since \(A = (r_1, \varphi, \frac{\pi}{2})\).

The LI-channel estimation scheme is designed as follows. First, the proposed scheme is based on decomposing the received training signal \(\{q(\varphi)\}\) into a Fourier series:
\[ q(\varphi) = \sum_{k = -\infty}^{\infty} Q_k e^{j k \varphi}, \quad \varphi \in [0, 2\pi] \tag{7} \]
where the Fourier coefficients \(\{Q_k\}\) are defined as
\[ Q_k = \frac{1}{2\pi} \int_{0}^{2\pi} q(\varphi) \, d\varphi. \tag{8} \]
The Fourier coefficients contain all information on the signal and thus can replace it in channel estimation. To facilitate the algorithmic design, the structure of the coefficients is characterized as follows. To this end, a few notations are introduced. Let \(Q\) represent the infinite sequence: \(\ldots, Q_{-1}, Q_0, Q_1, \ldots\). The product between two sequences, \(V_1\) and \(V_2\), is denoted and defined as \(V_1 \odot V_2 = \sum_{n = -\infty}^{\infty} [V_1]_n [V_2]_n\) where \([.]_n\) yields the element with index \(n\). Moreover, \(J_n : \mathbb{R} \rightarrow \mathbb{R}\) represents Bessel function with an integer index \(n\) as defined and discussed in Appendix A.

Lemma 1 (Training Signal Decomposition): The sequence \(Q\) can be decomposed as
\[ Q = \frac{\lambda}{4\pi r_0} e^{-j \frac{2\pi}{\lambda} r_0} \sum_{u=1}^{U} V (X_u) + \frac{\alpha_r}{r_0}, \quad \text{a.s.} \tag{9} \]
where \(V (X_u) = \ldots \cdot V_1 (X_u), V_0 (X_u), V_1 (X_u) \ldots\) with \(V_n(Y)\) defined for a given location \(Y = (r_Y, \varphi_Y, \frac{\pi}{2})\) as
\[ V_n(Y) = j^n e^{j n \varphi_Y} J_n \left(\frac{2\pi}{\lambda} r_Y\right). \tag{10} \]
The proof is provided in Appendix B-A.

Remark 1: The Fourier coefficients \(\{Q_k\}\) of the received training signal are noiseless. The noise suppression is the combined result of the noise spatial whiteness in Assumption 2 and the integral operation in (8).

Next, based on the decomposition of \(Q\) in Lemma 1, the key component of the proposed scheme for LI-channel estimation is a function \(\Phi : \mathbb{R}^2 \rightarrow \mathbb{R}_+\) defined as
\[ \Phi(Y) = \frac{4\pi r_0}{\lambda} |V(Y) \odot Q| \tag{11} \]
and called the channel observation profile. To derive a closed-form expression for \(\Phi(Y)\), a useful property for \(V\) directly follows from the Addition Theorem in Property (B2) of Bessel functions in Appendix A as shown below.

Lemma 2: Given two locations \(X, Y \in \mathbb{R}^2\), the product between the sequences \(V(X)\) and \(V(Y)\) satisfies
\[ |V(X) \odot V(Y)| = J_0 \left(\frac{2\pi}{\lambda} |X - Y|\right). \]

Combining Lemmas 1 and 2 and the definition of \(\Phi(Y)\) in (11) gives the following theorem.

Theorem 1 (Channel Observation): As \(r_0 \rightarrow \infty\), the channel observation profile for the circular UA is noiseless and given as
\[ \Phi(Y) = \sum_{u=1}^{U} J_0 \left(\frac{2\pi}{\lambda} |X_u - Y|\right) + \alpha_r, \quad \text{a.s.} \tag{12} \]

An example of \(\Phi(Y)\) is illustrated in Fig. 2.

LI-channel estimation scheme: One can observe from Theorem 1 that the \(U\) Bessel functions in \(\Phi(Y)\) have their peaks at corresponding user locations since \(J_0(x)\) is maximized at \(x = 0\). Motivated by this fact, the proposed scheme for estimating the user locations is to detect the peaks in the channel observation profile \(\Phi(Y)\). Since the profile is noiseless, the only source for estimation errors is the coupling (interference) between the Bessel functions in \(\Phi(Y)\) due to finite separation distances between user locations.

Remark 2 (Channel Estimation Error): Channel estimation using the proposed scheme is close to perfect for a single-user system since the corresponding channel observation profile \(\Phi(Y) \approx J_0 \left(\frac{2\pi}{\lambda} |X_1 - Y|\right)\) that is maximized at \(Y = X_1\)
with $\Phi(X_1) \approx 1$. Next, consider the estimation of multiuser LI-channels. The accuracy for estimating the channel corresponding to user $X_u$ can be evaluated by the difference between $\Phi(X_u)$ and the value of 1 for its single-user counterpart. This measures the interference due to the presence of multiuser channels and can be obtained from Theorem 1 as follows:

$$|\Phi(X_u) - 1| \leq \sum_{k \neq u} J_0\left(\frac{2\pi}{\lambda}|X_k - X_u|\right) + o_r, \quad \text{a.s.}$$

Based on Property (B4) of Bessel functions in Appendix A,

$$|\Phi(X_u) - 1| \leq \frac{1}{\nu} \sum_{k \neq u} \left(\frac{2\pi}{\lambda}|X_k - X_u|\right)^{-\frac{1}{2}} + o_r, \quad \text{a.s.}$$

$$\leq \frac{U - 1}{\nu} \left(\frac{2\pi}{\lambda \nu \min_k|X_k - X_u|}\right)^{-\frac{1}{2}} + o_r. \quad (13)$$

The result shows that the interference magnitude diminishes with the increasing minimum user-separation distance approximately following a sub-linear function. Given a pair of users, setting the upper bound in (13) to a small value e.g., 0.1, a rule of thumb for the user-separation distance sufficient large for accurate channel estimation can be computed as $77\lambda$, which is 23 m for a carrier of 1 GHz and 2.3 m for 10 GHz.

**Remark 3 (Effect of Wavelength on Channel Estimation):**

One can infer from (13) that with user locations fixed, the accuracy of channel estimation can be improved by reducing the carrier wavelength $\lambda$. However, this leads to a denser UA in practice since its elements are required to be separated by no more than $\lambda/2$.

**Remark 4 (Scattering Channels):** Assuming the reflection at scatterers is isotropic, the channels between the UA and users are determined not only by the scatterers’ locations but also by the channel coefficients that combine the gains of the channels between scatterers and users and reflection attenuation at scatterers. Estimation of the channels can follow a two-phase procedure. First, the scatter locations can be estimated using a similar scheme as discussed earlier for estimating user locations. This is possible due to that by reflecting signals from users, a scatter becomes a virtual signal source. Next, resolving the scatter locations allows the estimation of the training signals reflected by individual scatterers. However, further estimating the individual channel coefficient between each pair of scatterer and user requires the use of pilot sequences and exploiting their orthogonality. Therefore, unlike that of free-space channels, channel estimation using single pilot symbols is infeasible for scattering channels.

**Remark 5 (Blind Channel Estimation):** Replacing the known single pilot symbols with unknown data symbols with unit magnitude leads to the same channel observation profile as given in Theorem 1. In other words, it is possible to blindly estimate the multiuser LI-channels using the UA by observing the users’ transmissions.

### B. LI-Channel Estimation With Pilot Sequences

In this section, the results in the preceding section for single pilot symbols are extended to the case with pilot sequences. Let the pilot sequence for the $u$-th user be denoted as $s_u = [s_u(1), \cdots, s_u(L)]^T$ with length $L$. In the channel training phase, the UA receives sequentially $L$ infinite sequences over $L$ symbol durations, denotes as $Q(1), Q(2), \cdots, Q(L)$, is modified from their single-symbol counterpart in Lemma 1 as

$$Q(\ell) = \frac{\lambda e^{-\frac{i2\pi r_0}{\lambda \ell}}}{4\pi r_0} \sum_{u=1}^{U} s_u(\ell) \mathbf{V}(X_u) + o_r, \quad \text{a.s.} \quad (14)$$

To estimate the user location $X_u$, $\{Q(\ell)\}$ are coherently combined using $s_u$ as $\sum_{\ell=1}^{L} s_u(\ell) Q(\ell)$. The result, denoted as $Q_u$, follows from (14) as

$$Q_u = \frac{\lambda e^{-\frac{i2\pi r_0}{\lambda}}}{4\pi r_0} \sum_{k=1}^{U} s_u s_k \mathbf{V}(X_k) + o_r, \quad \text{a.s.} \quad (15)$$

A corresponding channel observation profile $\Phi_u(Y)$ can be defined similarly as in (11):

$$\Phi_u(Y) = \frac{4\pi r_0}{\lambda} |\mathbf{V}(Y) \circ Q_u|. \quad (16)$$

The profile $\Phi_u(Y)$ is decomposed into the desired and interference terms as follows:

$$\Phi_u(Y) = J_0\left(\frac{2\pi}{\lambda}|X_u - Y|\right) + \sum_{k \neq u} s^*_k s_k J_0\left(\frac{2\pi}{\lambda}|X_k - Y|\right) + o_r.$$  

Then the deviation of $\Phi_u(Y)$ from its single-user counterpart can be bounded as:

$$\left|\Phi_u(Y) - J_0\left(\frac{2\pi}{\lambda}|X_u - Y|\right)\right| \leq \sum_{k \neq u} |s^*_k s_k| J_0\left(\frac{2\pi}{\lambda}|X_k - Y|\right) + o_r,$$

for all $u$. This leads to the following main result of this section.

**Theorem 2 (Effect of Pilot Sequences):** Given $L \geq U$ and orthogonal pilot sequences, the LI-channel estimation using the circular UA is almost perfect since the channel observation profile is approximately equal to the single-user counterpart:

$$\left|\Phi_u(Y) - J_0\left(\frac{2\pi}{\lambda}|X_u - Y|\right)\right| \leq o_r, \quad \text{as } r_0 \to \infty$$

for $u = 1, 2, \cdots, U$.

Comparing Theorems 1 and 2, the advantage of pilot sequences over single pilot symbols lies in their capability of decoupling the estimation of multiuser LI-channels, thereby providing close-to-perfect channel estimation.

### C. Effect of Finite Elements in the UA

Consider a discrete circular UA with $N$ antennas uniformly placed on the circle centered at the origin and with the radius $r_0$. The discrete UA can be interpreted as a quantized version of the continuous UA with the quantization error bounded by $\pi/N$. Based on this interpretation and assuming $N$ is large, the analysis for the continuous UA can be straightforwardly extended to the case of discrete UA by including the quantization error. As a result, for channel estimation with single pilot symbols,
the channel observation profile for the discrete UA, denoted as \( \hat{\Phi}(Y) \), can be written as

\[
\hat{\Phi}(Y) = \Phi(Y) + O\left(\frac{1}{N}\right)
\]

(17)

where \( \Phi(Y) \) for the continuous UA is given in Theorem 1. Moreover, the result in Theorem 2 for the case of pilot sequences can be modified by replacing \( c_p \) with \( c_p + O(1/N) \).

Similarly, for data transmission using the discrete UA, the receive SINR at user \( u \), denoted as \( \text{SINR}_u \), can be shown to be \( \text{SINR}_u = \text{SINR}_u + O(1/N) \) with \( \text{SINR}_u \) corresponding to the continuous UA. The above results suggest that with respect to the continuous counterpart, the discrete UA causes additional fluctuation in the channel observation profile and receive SNRs, which can potentially degrade the performance of channel estimation and data transmission.

IV. COMMUNICATION USING THE CIRCULAR UA: DATA TRANSMISSION

In this section, two precoding techniques are designed for the UA, namely the channel conjugate and the multiuser phase mode (MU-PM) precoding. For simplicity, it is assumed that the UA has perfect knowledge of the LI-channels.

A. Channel Conjugate Transmission

For channel conjugate transmission, the precoder applies a phase shift to each antenna for compensating for propagation delay to achieve coherent combining at the target user location, which is similar to beamforming using a phase array. Specifically, the precoder \( f_u(\varphi) \) is given as

\[
f_u(\varphi) = \frac{h_u^*(\varphi)}{|h_u(\varphi)|}, \quad \varphi \in [0, 2\pi)
\]

(18)

where \( h(\varphi) \) is given in (2). It follows that

\[
f_u(\varphi) = e^{j \frac{\varphi}{2} (r_0 - r_u \cos \psi_u(\varphi))}, \quad \varphi \in [0, 2\pi).
\]

The normalization \( |f_u(\varphi)|^2 = 1 \) facilitates the UA implementation under the per-element power constraint.

The channel conjugate precoder \( \{f_u(\varphi)\} \) shapes the distribution of the field power density such that transmission power is concentrated in a small region centered at the target location \( X_u \). Characterizing the distribution is useful for analyzing the precoder performance. To this end, consider the transmission of an unmodulated wave using the UA after precoding using \( \{f_u(\varphi)\} \) in (19). Conditioned on the precoder, let \( g(X_k | X_u) \) denote the resultant field measured at the location \( X_k \). With the propagation distance in (1), it can be obtained that

\[
g(X_k | X_u) = \sqrt{\frac{P_i}{2\pi r_0}} \times \frac{1}{\sqrt{4\pi r_0}} \times \int_{0}^{2\pi} e^{j \frac{\varphi}{2} (-r_0 + r_u \cos \psi_u(\varphi))} f_u(\varphi) r_0 d\varphi + \frac{c_p}{\sqrt{r_0}}.
\]

(20)

At the right-hand side of (20), the three factors of the dominant term correspond to the density of transmission power uniformly distributed over the UA, the propagation loss and the wave superposition at \( X_u \), respectively. Substituting the precoder in (19) into (20) yields

\[
g(X_k | X_u) = \sqrt{\frac{P_i}{2\pi r_0}} \times \frac{1}{\sqrt{4\pi r_0}} \times \int_{0}^{2\pi} e^{j \frac{\varphi}{2} (-r_0 + r_u \cos \psi_u(\varphi))} \sqrt{\frac{P_i}{2\pi r_0}} \times \frac{1}{\sqrt{4\pi r_0}} \times \int_{0}^{2\pi} e^{j \frac{\varphi}{2} (-r_0 + r_u \cos \psi_u(\varphi))} f_u(\varphi) r_0 d\varphi + \frac{c_p}{\sqrt{r_0}} = \sqrt{\frac{P_i}{2\pi r_0}} \times \frac{1}{\sqrt{4\pi r_0}} \times \int_{0}^{2\pi} d\varphi.
\]

(21)

The field power density at location \( X_k \) can be represented by

\[
p(X_k | X_u) = g(X_k | X_u)^2.
\]

Using (21), a closed-form expression for \( p(X_k | X_u) \) is obtained as shown in the following lemma proved in Appendix B-B.

**Lemma 3 (Field Power Density Distribution):** Given the circular UA and channel conjugate precoding targeting user \( X_u \), the field power density at the user location \( X_k \) is

\[
p(X_k | X_u) = \frac{P_i}{2\pi r_0} \int_{0}^{2\pi} \frac{2\pi}{\lambda} |X_u - X_k| + \frac{c_p}{\sqrt{r_0}}.
\]

(22)

as \( r_0 \to \infty \).

The result shows that the field power density function \( p(X_k | X_u) \) depends only on the distance \( |X_k - X_u| \) and thus can be rewritten as \( p(d) \) with \( d > 0 \) denotes the distance from the location targeted by the precoder. Then the distribution of the field power density can be characterized by the function \( p(d) / p(0) = J_0\left(\frac{2\pi d}{\lambda}\right) \) that is plotted in Fig. 3. It can be observed from the figure that the channel-conjugate precoder shapes the field distribution such that most power is concentrated within a circular region centered at the target location and having a radius of half wavelength. The tail of the distribution function is undesirable as it causes interference to nearby unintended receivers. However, the envelop of the tail decays with distance \( d \), allowing interference suppression by spatial separation as further discussed in the sequel.

With the field distribution in Lemma 3, the performance of the channel-conjugate precoding is readily analyzed in terms of receive SINRs as follows. Let \( P_i \) and \( P_i \) denote the signal and...
interference powers at user $X_u$, respectively. Since the effective aperture of the receive omni-directional antenna is $\lambda/4\pi$, $P_r$ and $P_t$ are given as

$$P_r = \frac{\lambda}{4\pi} p(X_u | X_u) \quad \text{and} \quad P_t = \frac{\lambda}{4\pi} \sum_{k \neq u} p(X_u | X_k). \quad (23)$$

The receive SINR for user $X_u$ can be written as

$$\text{SINR}_u = \frac{P_t \lambda}{8\pi \sigma^2 r^2} + \frac{\sigma_t}{r_0}. \quad (24)$$

Substituting (22) and (23) into (24) yields the following main result of this section.

**Theorem 3 (Receive SINRs):** For channel conjugate transmission using the circular UA, the receive SINRs are

$$\text{SINR}_u = \frac{1}{\sum_{i \neq u} j_0^2 \left( \frac{|t_i|}{X_u - X_i} \right) + \frac{1}{\text{SIR}}} + \sigma_t, \quad \text{as} \quad r_0 \to \infty$$

for $u = 1, 2, \cdots, U$ where the receive SNR is given as

$$\text{SNR} = \frac{P_t \lambda}{8\pi \sigma^2 r^2} + \frac{\sigma_t}{r_0}. \quad (25)$$

**Remark 6 (High SNR):** Applying the bound on the Bessel function in Property (B4) in Appendix A, for a high SNR, the signal-to-interference ratio (SIR) at user $u$ scales with the distance to the nearest interferer, namely $\min_{k \neq u} |X_k - X_u|$, and the wavelength $\lambda$ as

$$\text{SIR}_u \geq \frac{\nu^2}{U - 1} \left( \frac{2 \min_{k \neq u} |X_k - X_u|}{\lambda} \right)^2 + \sigma_t, \quad \forall \ u. \quad (26)$$

Thus the receive SIRs increase with the increasing minimum user-separation distance (in wavelength) following a sub-linear function. Moreover, the result in (26) suggests that denser simultaneous users can be supported by reducing the wavelength without compromising the system throughput. Specifically, the user density can scale linearly with $1/\lambda^2$.

**Remark 7 (Free Space vs. Scattering Channels):** Free-space channels allow the UA to focus signal energy at intended users such that the signal power decays rapidly with the distance from the target location. As a result, given single-user (channel-conjugate) transmission, interference can be suppressed by increasing user spatial separation as shown in (26). However, this is infeasible in the scenario of scattering channels as scattering introduces additional cross coupling between multiuser signals. Suppressing the resultant interference cannot rely on increasing users’ spatial separation and multiuser precoding has to be used for this purpose.

Last, the receive SNR in (25) implies the following result.

**Corollary 1 (Multipath Loss):** For channel-conjugate transmission using the circular UA, the propagation loss is

$$\frac{P_t}{P_t} = \frac{\lambda}{8\pi r^2} + \frac{\sigma_t}{r_0}, \quad \text{as} \quad r_0 \to \infty. \quad (27)$$

In other words, transmission using the circular UA reduces the loss such that it is inversely proportional to the propagation distance instead of its square as in the case of a conventional array of collocated antennas.

### B. Multiuser Phase Mode Precoding

In this section, MU-PM precoders are designed under the zero-forcing constraints to avoid multiuser interference. Exploiting the circular structure of the UA, the precoder for each user, say $f_u' (\varphi)$ for user $u$, can be expressed in terms of the Fourier series representing the sequence of phase modes:

$$f_u' (\varphi) = f_u (\varphi) \sum_{m = -\infty}^{\infty} c_{u,m} e^{-jm\varphi}, \quad \varphi \in [0, 2\pi) \quad (28)$$

where $f_u (\varphi)$ is the channel-conjugate precoder in (18) for compensating the channel phase shift and the summation is the said Fourier series. The precoder coefficients $\{c_{u,m}\}$ satisfy the power constraint: $\sum_{m = -\infty}^{\infty} |c_{u,m}|^2 \leq 1$ for all $u$.

The precoder coefficients are designed to avoid the inter-user interference as follows. Without multiuser interference, it is unnecessary to analyze the field spatial distribution as in the case of channel conjugate transmission and instead the remainder of the section focuses on the precoder design and the analysis of the received signal power. To this end, the received signal at user $u$ is obtained by substituting the precoder in (28) into (5):

$$y_u = \sqrt{\frac{\lambda P_t}{8\pi r_0}} \sum_{k=1}^{U} \left( \frac{1}{2\pi} \int_{0}^{2\pi} h^*_k (\varphi) h_u (\varphi) \frac{\rho_k (\varphi)}{|h_k (\varphi)|} d\varphi \right) x_k + z_u. \quad (29)$$

The substitution of the channel response in (2) yields

$$y_u = \sqrt{\frac{\lambda P_t}{8\pi r_0}} c_{u,0} x_u + \sqrt{\frac{\lambda P_t}{8\pi r_0}} \sum_{k \neq u, m = 0} c_{k,m} J_{u,k,m} x_k + z_u + \frac{\sigma_t}{r_0}. \quad (29)$$

where the first term and the summation represent the signal and interference, respectively, and $J_{u,k,m}$ is defined as

$$J_{u,k,m} = \frac{1}{2\pi} \int_{0}^{2\pi} e^{\frac{j2\pi}{\lambda} (r_u \cos \varphi_u - r_k \cos \varphi_k) - jm \varphi} d\varphi.$$

A closed-form expression for $J_{u,k,m}$ can be obtained as follows, following similar steps as in the proof for Lemma 3 with the details omitted for brevity.

**Lemma 4:** The coefficients $\{J_{u,k,m}\}$ defined earlier can be written as

$$J_{u,k,m} = e^{jm \zeta_{u,k,m}} J_m \left( \frac{2\pi}{\lambda} |X_u - X_k| \right) \quad (29)$$

where $1 \leq u, k \leq U$ and $m$ is an integer.

The interference term in the signal in (29) can be nulled by enforcing the following zero-forcing constraints:

$$\sum_{m = -\infty}^{\infty} c_{k,m} J_{u,k,m} = 0, \quad \forall k \neq u. \quad (30)$$

In practice, given a discrete UA and a constraint on computation complexity, it is infeasible to implement MU-PM precoders with an infinite number of phase modes and only a finite set of modes.
is considered in the design. Let the corresponding coefficient set be denoted as \( \{c_{u,m} \mid 1 \leq u \leq U, -M \leq m \leq M \} \) with \( M \) being a fixed integer and other coefficients set to zero. Under the zero-forcing constraints in (30), it is desirable to choose the precoder coefficients such that the correspond set of coefficients \( \{J_{u,k,m} \mid 1 \leq u, k \leq U, -M \leq m \leq M \} \) are significant. Then applying Property (B4) of Bessel functions in Appendix A gives

\[
M = 2\pi \min_{u,k} \left| \frac{|X_u - X_k|}{\lambda} \right| 
\]

(31)

where users are assumed to be separated by distances much larger than a single wavelength, yielding the following fact.

Remark 8 (Degrees of Freedom): Given MU-PM transmission using a circular array, the total number of degrees of freedom for interference avoidance is approximately equal to \((2M + 1)\) with \( M \) given in (31), which is proportional to the minimum user-separation distance.

Transmission over only \((2M + 1)\) phase modes allows the precoder coefficients for user \( u \) to be represented by the vector \( c_u = [c_{u,-M}, \ldots, c_{u,0}, \ldots, c_{u,M}]^T \). Furthermore, the zero-forcing constraints in (30) can be written in a matrix form:

\[
\mathcal{J}_u c_u = 0 
\]

(32)

where the matrix \( \mathcal{J}_u \) is defined as

\[
\mathcal{J}_u = \begin{bmatrix}
J_{u,1,-M}, & \cdots, & J_{u,1,0}, & \cdots, & J_{u,1,M} \\
J_{u,u-1,-M}, & \cdots, & J_{u,u-1,0}, & \cdots, & J_{u,u-1,M} \\
\vdots & \ddots & \vdots & \ddots & \vdots \\
J_{u,u+1,-M}, & \cdots, & J_{u,u+1,0}, & \cdots, & J_{u,u+1,M} \\
J_{u,U,-M}, & \cdots, & J_{u,U,0}, & \cdots, & J_{u,U,M}
\end{bmatrix}
\]

with the elements specified in Lemma 4. Under the constraints in (32), the main result of the section can be stated as follows.

Theorem 4 (Multiuser UA Precoding): For MU-PM transmission using the circular array, the precoder coefficients satisfy

\[
c_u \in \text{null} \left( \mathcal{J}_u \right), \quad u = 1, 2, \ldots, U
\]

(33)

where the matrix \( \mathcal{J}_u \) is given earlier and \( \text{null} \left( \mathcal{J}_u \right) \) denotes the null-space of \( \mathcal{J}_u \).

With interference avoided, the receive SNR for user \( u \) follows from (29) as

\[
\text{SNR}_u = \frac{\lambda P_1 |c_{u,0}|^2}{8\pi r_0 \sigma^2} + \frac{\alpha_r}{r_0} \quad \text{as } r_0 \to \infty 
\]

(34)

where \( e_0 = [0, \cdots, 0, 1, 0, \cdots, 0]^T \). Thus, to maximize \( \text{SNR}_u \), the precoder-coefficient vector \( c_u \) should be chosen as the projection of \( e_0 \) onto \( \text{null} \left( \mathcal{J}_u \right) \) that contains \( c_u \) according to Theorem 4. This gives the following corollary.

Corollary 2: For MU-PM transmission using the circular UA, the maximum receive SNRs are given as

\[
\text{SNR}_u = \frac{\lambda P_1 |e_u| b_u|^2}{8\pi r_0 \sigma^2} + \frac{\alpha_r}{r_0} \quad \text{as } r_0 \to \infty 
\]

(35)

for \( u = 1, 2, \ldots, U \) where \( b_u \) is a basis of \( \text{null} \left( \mathcal{J}_u \right) \).

With respect to the case of channel-conjugate transmission, the term \( |e_u| b_u|^2 \) represents the loss in receive SNR due to interference nulling.

V. COMMUNICATION USING THE SPHERICAL UA

In the preceding section, the UA is modeled as the circular array. The results are extended in this section to the spherical UA system, which shows its performance improvements with respect to the circular-UA counterpart. Essentially, the difference in analysis arises from the use of spherical harmonics as the tool in place of Fourier series.

A. Spherical LL-Channel Estimation

Assume that users transmit single pilot symbols: \( \{s_u\} = \{1\} \). The estimation scheme in Section III is extended to the spherical UA as follows. To this end, the training signal in (4), re-denoted as \( q(\varphi, \theta) \), is expanded as a Laplace series as follows by using the harmonic functions \( \{Y^m_{\ell}\} \), defined in (57) in Appendix A, as the basis:

\[
q(\varphi, \theta) = \sum_{\ell=0}^{\infty} \sum_{m=-\ell}^{\ell} Q^m_{\ell} Y^m_{\ell}(\varphi, \theta) \sin \varphi d\varphi d\theta. \quad (36)
\]

(37)

To derive a closed-form expression for \( Q^m_{\ell} \), a useful result is given as follows.

Lemma 5: Given two points \( (r, \varphi, \theta), (r_u, \varphi_u, \theta_u) \in \mathbb{R}^3 \),

\[
e^{j \frac{2\pi}{\lambda} r \cos \psi_u} = (2\pi)^{\frac{3}{2}} \sum_{\ell=0}^{\infty} \sum_{m=-\ell}^{\ell} \left( \frac{j^\ell J_{\ell+\frac{1}{2}} \left( \frac{2\pi}{\lambda} r_u \right)}{(2\pi)^\frac{1}{2} \frac{2\pi}{\lambda} r_u} \right)^{\frac{m}{2}} Y^m_{\ell}(\varphi_u, \theta_u) Y^m_{\ell}(\varphi, \theta)
\]

where \( \psi_u \) denotes the angle between the vectors corresponding to the two points.

The lemma is proved in Appendix B-C. Using the orthogonality of the basis functions \( \{Y^m_{\ell}(\varphi, \theta)\} \), substitution of the expression for \( q(\varphi, \theta) \) in (4) and Lemma 5 into (37) leads to the following lemma.

Lemma 6 (Training Signal Decomposition): The training signal received at the spherical UA, namely \( \{q(\varphi, \theta)\} \) in (36), has the Laplace coefficients given as follows:

\[
Q^m_{\ell} = \frac{\lambda (2\pi)^{\frac{3}{2}}}{4\pi^2 \alpha} \sum_{u=1}^{U} \left( \frac{j^\ell J_{\ell+\frac{1}{2}} \left( \frac{2\pi}{\lambda} r_u \right)}{(2\pi)^\frac{1}{2} \frac{2\pi}{\lambda} r_u} \right)^{\frac{m}{2}} Y^m_{\ell}(\varphi_u, \theta_u) + \frac{\alpha_r}{r_0}
\]

as \( r_0 \to \infty \) where \( \ell = 1, 2, \ldots \) and \( -\ell \leq m \leq \ell \).

Define two sets \( Q \) and \( V(Y) \) as \( Q = \{Q^m_{\ell}\} \) and \( V(Y) = \{V^m_{\ell}(Y)\} \) with their elements \( Q^m_{\ell} \) given in Lemma 6 and \( V^m_{\ell} : \mathbb{R}^3 \to \mathbb{R} \) being a function defined as

\[
V^m_{\ell}(Y) = (2\pi)^{\frac{3}{2}} \left( \frac{j^\ell J_{\ell+\frac{1}{2}} \left( \frac{2\pi}{\lambda} r_u \right)}{(2\pi)^\frac{1}{2} \frac{2\pi}{\lambda} r_u} \right)^{\frac{m}{2}} Y^m_{\ell}(\varphi_u, \theta_u). \quad (38)
\]
Using these definitions, the channel observation profile corresponding to the spherical UA, represented by $\tilde{\Phi}(Y)$, can be defined similarly as in (11):

$$
\tilde{\Phi}(Y) = \frac{r_0}{\lambda} \left| \tilde{V}(Y) \circ \tilde{Q} \right|
$$

$$
= \frac{r_0}{\lambda} \sum_{\ell=1}^{U} \sum_{m=0}^{\infty} \sum_{m=-\ell}^{\ell} \left| [V^m(Y)]^* Q^m \right| .
$$

Substituting the result in Lemma 6 and (38) gives

$$
\tilde{\Phi}(Y) = \sum_{\ell=1}^{U} \sum_{m=0}^{\infty} \sum_{m=-\ell}^{\ell} \left| [V^m(Y)]^* Y^m(X_u) \right| + o_r
$$

where noise variances for the same reason as for the case of circular UA. By applying Addition Theorem in Property (S3) for spherical harmonics in Appendix A,

$$
\tilde{\Phi}(Y) = \sum_{\ell=1}^{U} \sum_{m=0}^{\infty} \sum_{m=-\ell}^{\ell} \left| [\tilde{V}^m(Y)]^* Y^m(X_u) \right| + o_r.
$$

Next, applying Addition Theorem in Property (B3) of Bessel functions in Appendix A further simplifies the expression as shown in the following theorem.

**Theorem 5 (Channel Observation):** As $r_0 \to \infty$, the channel observation profile for the spherical UA is given as

$$
\tilde{\Phi}(Y) = \sum_{u=1}^{U} \frac{\pi r_0}{\lambda} \left| J_\ell \left( \frac{2 \pi r_0}{\lambda} \right) \phi, \theta \right| + o_r, \ a.s.
$$

**Remark 9 (Channel Estimation Error):** As in Remark 2 for the circular UA, the accuracy of channel estimation can be evaluated using the difference $|\tilde{\Phi}(X_u) - 1|$. Since $|\sin(d)| \leq d^{-1}$, it can be obtained from Theorem 5 that

$$
|\tilde{\Phi}(X_u) - 1| \leq 2\pi (U - 1) \min_{u \neq k} \left( \frac{|X_k - X_u|}{\lambda} \right) + o_r, \ a.s.
$$

The error bound diminishes inversely with the increasing minimum user-separation distance (in wavelength) following an inverse function that is faster than the circular-UA counterpart in (13). This quantifies the gain of increasing the UA by one dimension from the perspective of channel estimation.

Next, consider the case where users transmit pilot sequences with length of $L$ symbols. Following the same procedure as for deriving Theorem 2 yields the following corollary.

**Corollary 3 (Effect of Pilot Sequences):** Given $L \geq U$ and orthogonal pilot sequences, the LI-channel estimation using the spherical UA is almost perfect since the channel observation profile is approximately equal to the single-user counterpart:

$$
\left| \tilde{\Phi}_u(Y) - \sin \left( \frac{2\pi}{\lambda} |X_u - Y| \right) \right| \leq o_r, \ a.s. \quad r_0 \to \infty
$$

**Remark 10:** If orthogonal pilot sequences are used, the spherical UA does not have an advantage over the circular counterpart in terms of channel estimation. However, the former improves the performance of channel estimation in the case of single pilot symbols.

**B. Data Transmission Using the Spherical UA**

1) **Channel Conjugate Transmission:** For channel conjugate transmission using the spherical UA, the precoder, denoted as $\tilde{J}_u(\varphi, \theta)$, is modified from the circular counterpart in (19) as

$$
\tilde{J}_u(\varphi, \theta) = \frac{h_u^*(\varphi, \theta)}{|h_u^*(\varphi, \theta)|}, \quad \theta \in [0, 2\pi], \varphi \in [0, \pi).
$$

The resultant receive SNRs are derived as follows. Let $\tilde{q}(X_u | X_k)$ and $\tilde{p}(X_u | X_k)$ denote the field and its power density measured at location $X_u$ given a precoder targeting user $X_k$. Then $\tilde{q}(X_u | X_k)$ can be obtained by modifying the circular-UA counterpart in (20) by replacing the integration over a circle with one over a sphere:

$$
\tilde{q}(X_u | X_k) = \sqrt{\frac{P_t}{4\pi}} \int \int e^{i \frac{2\pi}{\lambda} (r_u \cos \psi_u - r_k \cos \psi_k)} \times \sin \theta \, d\theta \, d\varphi + o_r.
$$

To facilitate analysis, a set of coefficients $\{J_{u,k,m,n}\}$ with integer indices $(u, k, m, n)$ are defined as

$$
J_{u,k,m,n} = \int \int e^{i \frac{2\pi}{\lambda} (r_u \cos \psi_u - r_k \cos \psi_k)} Y^m_{\ell} (\varphi, \theta) \times \sin \varphi \, d\varphi \, d\theta
$$

with $1 \leq u, k \leq U, n \geq 0$ and $-\ell \leq m \leq \ell$, which are also used for designing multiuser precoders in the next sub-section. They can be written in a closed form as shown in the following lemma.

**Lemma 7:** The coefficients $J_{u,k,m,n}$ defined in (43) can be written as

$$
J_{u,k,m,n} = \frac{(2\pi)^2}{4\pi} \int \int J_{\ell+1} \left( \frac{2\pi}{\lambda} |X_u - X_k| \right) \frac{1}{\sqrt{2\pi} |X_u - X_k|} Y^m_{\ell} (\varphi_u, \theta_u) \times \sin \varphi_u \, d\varphi_u \, d\theta_u
$$

where the angles $\varphi_u, k$ and $\theta_u, k$ are defined by:

$$
\sin \varphi_u, k \cos \theta_u, k = \frac{r_u \sin \varphi_u \cos \theta_u - r_k \sin \varphi_k \cos \theta_k}{r_{u,k}}
$$

$$
\sin \varphi_u, k \sin \theta_u, k = \frac{r_u \sin \varphi_u \sin \theta_u - r_k \sin \varphi_k \sin \theta_k}{r_{u,k}}
$$

$$
\cos \varphi_u, k = \frac{r_u \cos \varphi_u - r_k \cos \varphi_k}{r_{u,k}}
$$

The proof of Lemma 7 is provided in Appendix B-D. Using the fact that

$$
J_{\ell} (x) = \sqrt{\frac{2}{\pi}} \frac{\sin x}{x^\ell}, \quad Y_{0,0} (\varphi_u, k, \theta_u, k) = \frac{1}{\sqrt{\lambda \pi}}
$$

and Lemma 7, the field in (42) can be obtained in a closed form, yielding the following result.
Lemma 8 (Field Power Density Distribution): Given the spherical UA and channel conjugate precoding targeting user \( X_u \), the field power density at the user location \( X_k \) is

\[
\hat{p}(X_u \mid X_k) = P_t \sin^2\left(\frac{2\pi}{\lambda} |X_u - X_k| \right) + o_r.
\]

As in the case of circular UA, \( \hat{p}(X_u \mid X_k) \) can be rewritten as \( \hat{p}(d) \) with \( d \) being the distance from the location targeted by the precoder. Then the distribution of the field power density can be characterized by the function \( \hat{p}(d) / \hat{p}(0) = \sin^2\left(\frac{2\pi d}{\lambda} \right) \) that is plotted in Fig. 3. Like the circular-UA counterpart, by channel-conjugate precoding, the spherical UA focuses the transmission power into a region within a distance of half wavelength from the target location. The advantage of the spherical UA is reflected in that the tail of the distribution function has an envelope diminishing with the increasing distance much faster than that corresponding to the circular UA. This reduces multiuser interference and leads to substantial performance improvements as shown in the analysis and observed from simulation results.

Given the SINR defined similarly as in (24), the result in Lemma 8 leads to the spherical-UA counterpart of Theorem 3 as follows.

Theorem 6 (Receive SINRs): For channel conjugate transmission using the spherical UA, the receive SINR for user \( u \), denoted as \( \overline{\text{SINR}}_u \), is given as

\[
\overline{\text{SINR}}_u = \sum_{k\neq u} \frac{1}{\overline{\text{SIR}}_u} \frac{1}{\sin^2\left(\frac{2\pi}{\lambda} |X_u - X_k| \right)} + o_r,
\]

for \( u = 1, 2, \ldots, U \) where the receive SNR is

\[
\text{SNR} = \frac{P_t \lambda^2}{4\pi \sigma^2}.
\]

Remark 11 (High SNR): Using the fact \( |\sin(d)| \leq d^{-1} \), the SIR lower bound in (26) for the circular UA can be modified for the current case as

\[
\overline{\text{SIR}}_u \geq \frac{1}{U-1} \left( \frac{2\pi \min_{k \neq u} |X_u - X_k|}{\lambda} \right)^2 + o_r.
\]

Thus, the receive SIRs increase with the minimum user-separation distance (in wavelength) at least following a super-linear function with the exponent 2, which is faster than the sub-linear function for the circular UA (see Remark 6). This specifies the gain of increasing the UA by one dimension from the perspective of received signal quality.

Last, the received SNR in (46) suggests the following result.

Corollary 4 (Propagation Loss): Given channel-conjugate transmission using the spherical UA, the propagation loss is approximately equal to the receive antenna aperture:

\[
P_r = \frac{\lambda^2}{4\pi} + o_r, \quad \text{as } r_0 \to \infty.
\]

In other words, the path loss is a constant and independent with the propagation distance \( r_0 \). In contrast, the loss corresponding to the circular UA and the conventional array is inversely proportional to \( r_0 \) (see Corollary 1) and \( r_0^2 \), respectively.

2) Multiuser Phase Mode Transmission: The phase modes for the spherical UA correspond to different spherical harmonics. Then the spherical MU-PM precoder for user \( u \), denoted as \( \hat{f}_u^m \), is modified from the circular counterpart in (28) as

\[
\hat{f}_u^m(\varphi, \theta) = \hat{f}_u(\varphi, \theta) \sum_{\ell=0}^{\infty} \sum_{m=-\ell}^{\ell} c_{u,m,\ell} Y_{\ell}^m(\varphi, \theta)
\]

where \( \{c_{u,m,\ell}\} \) are the precoder coefficients to be designed, \( \hat{f}_u(\varphi, \theta) \) is the channel conjugate precoder in (41) and the spherical harmonic function \( Y_{\ell}^m(\varphi, \theta) \) is defined in (57). By substitution of (49) into (5), the signal received at user \( X_u \) is

\[
\tilde{y}_u = \hat{y}_u = \lambda \sqrt{\frac{P_t}{4\pi}} \sum_{k=1}^{U} \sum_{\ell=0}^{\infty} \sum_{m=-\ell}^{\ell} c_{k,m,\ell} Y_{\ell}^m(\varphi, \theta) \sin \varphi \psi \psi \psi d\psi d\theta x_k + z_u.
\]

Substituting the channel response in (2) gives

\[
\tilde{y}_u = \lambda \sqrt{\frac{P_t}{4\pi}} c_{u,0,0} x_u + \lambda \sqrt{\frac{P_t}{4\pi}} \sum_{\ell=0}^{\infty} \sum_{m=-\ell}^{\ell} c_{k,m,\ell} Y_{\ell}^m(\varphi, \theta) \psi \psi \psi \psi \psi d\psi d\theta x_k + z_u + o_r.
\]

where the coefficients \( \{c_{u,m,\ell}\} \) are defined earlier in (43). It can be observed from Lemma 7 that \( J_{u,k,m,\ell} \) is proportional to \( J_{u,k,m,\ell} (\frac{2\pi}{\lambda} |X_u - X_k|) \). Therefore, following the same reasoning as for the circular UA, only the set of coefficients \( \{J_{u,k,m,\ell} \mid 0 \leq \ell \leq M \} \) have significant values. Considering only these values reduces the precoder coefficients to a finite set \( \{c_{k,m,\ell} \mid 0 \leq \ell \leq M \} \), yielding the following fact.

Remark 12 (Degrees of Freedom): Given MU-PM transmission using a spherical UA, the total number of degrees of freedom for interference avoidance is approximately equal to \((M+1)^2\) with \( M \) given in (31). As a result, since \( M \) is much greater than one, the number of degrees of freedom generated by the spherical UA is approximately proportional to \( M^2 \) that is much larger than that, namely \( 2M \), for the circular counterpart.

Next, the said finite set of precoder coefficients can be designed by applying the zero-forcing constraints that follow from (50) as

\[
\sum_{\ell=0}^{M} \sum_{m=-\ell}^{\ell} c_{k,m,\ell} J_{u,k,m,\ell} = 0, \quad \forall k \neq u.
\]

The constraints can be written in the matrix form. To this end, define the matrix \( \mathcal{H}_{u,\ell} \) as

\[
\mathcal{H}_{u,\ell} = \begin{bmatrix}
J_{u,1,-\ell,\ell}, & J_{u,1,-1+\ell,\ell}, & \cdots & J_{u,1,\ell,\ell} \\
J_{u,1,-\ell,\ell}, & J_{u,1,-1+\ell,\ell}, & \cdots & J_{u,1,\ell,\ell} \\
\vdots & \vdots & \ddots & \vdots \\
J_{u,1,-\ell,\ell}, & J_{u,1,-1+\ell,\ell}, & \cdots & J_{u,1,\ell,\ell} \\
J_{u,1,-\ell,\ell}, & J_{u,1,-1+\ell,\ell}, & \cdots & J_{u,1,\ell,\ell} \\
\vdots & \vdots & \ddots & \vdots \\
J_{u,1,-\ell,\ell}, & J_{u,1,-1+\ell,\ell}, & \cdots & J_{u,1,\ell,\ell} \\
\end{bmatrix}
\]
and the row vector \( \mathbf{a}_{u, \ell} = [c_{u, -\ell, \ell}, c_{u, -\ell+1, \ell}, \ldots, c_{u, \ell, \ell}] \) where \( \ell = 0, 1, \ldots \). Moreover, using these matrices/vectors as elements, define

\[
\mathbf{J}_u = [\mathbf{H}_{u,0}, \mathbf{H}_{u,2}, \ldots, \mathbf{H}_{u,M}], \\
\mathbf{c}_u = [\mathbf{a}_{u,1}, \mathbf{a}_{u,2}, \ldots, \mathbf{a}_{u,M}]^T.
\]

Using these definitions, the zero forcing constraints in (51) can be written as

\[
\mathbf{J}_u \mathbf{c}_u = \mathbf{0}, \quad u = 1, 2, \ldots, U.
\]

The main result of the section is summarized as follows.

**Theorem 7 (Multiuser UA Precoding):** For MU-PM transmission using the circular UA, the precoder coefficients under the zero-forcing constraints are given as

\[
\mathbf{c}_u \in \text{null}(\mathbf{J}_u), \quad u = 1, 2, \ldots, U
\]

where \( \text{null}(\mathbf{J}_u) \) denotes the null-space of \( \mathbf{J}_u \).

The spherical-UA counterpart of Corollary 2 is as follows.

**Corollary 5 (Receive SNRs):** For MU-PM transmission using the circular UA, the maximum receive SNR for user \( u \) is

\[
\text{SNR}_u = \frac{\eta^2 P_t |e_0^t b_u|^2}{\sigma^2} + o_r, \quad \text{as } r_0 \to \infty
\]

for \( u = 1, 2, \ldots, U \) where \( b_u \) is a basis of \( \text{null}(\mathbf{J}_u) \).

**Remark 13:** Theorem 7 and Corollary 5 are observed to have the same forms as Theorem 3 and Corollary 2, respectively. However, the space \( \text{null}(\mathbf{J}_u) \) corresponding to the spherical UA is much larger than its circular-UA counterpart \( \text{null}(\mathbf{J}_u) \). The extra degrees of freedom allow the spherical-UA system to support a larger number of simultaneous users and reduce the received SNR loss due to interference avoidance.

### VI. Simulation Results

In simulation, discrete UAs with finite numbers of antennas are considered. As illustrated in Fig. 4, the UA antennas are uniformly distributed on a circle for the case of circular UA or a sphere for the case of spherical UA, both of which have the fixed radius \( r_0 = 20 \text{ m} \) modeling a small deployment region. For benchmarking, the conventional *collocated array* is also included in simulation whose antennas are uniformly distributed in a horizontal disk with a radius denoted as \( r_d \) and right above the origin with a vertical distance equal to \( r_0 \) as shown in Fig. 4(c). Note that this location and orientation of the collocated UA are found by simulation to yield the best performance among other configurations with the same disk radius and distance to the origin. The radius of the collocated UA is set as \( r_d = 5 \text{ m} \) in Fig. 4 for ease of illustration and \( r_d = 1 \text{ m} \) for all other simulation results. In addition, relaxing Assumption 1, the users are uniformly distributed in the horizontal disk with a radius of \( 0.5r_0 \) instead of being near the origin. Moreover, the carrier frequency is 2.5 GHz and the noise variance is \(-100 \text{ dBm}\).

#### A. Channel Estimation

Consider channel estimation using single pilot symbols and algorithms from straightforward extension of those in Sections III and VA to discrete arrays. The average channel estimation error is defined as the difference between the estimated and
Consider LI-channel estimation with single pilot symbols. (a) Effect of the number of antennas on channel estimation error for 10 users. (b) Effect of the number of users on channel estimation error for 200 transmit antennas.

actual locations of a typical user as averaged over the random distributions of users and antennas. Fig. 5 displays the curves of average channel estimation error versus the number of antennas and those of average error versus the number of users in two separate sub-figures. Several observations can be made from the curves. As the number of antennas increases, the average errors for the circular and spherical UAs both diminishes rapidly and converges to a small constant corresponding to the continuous arrays analyzed in the preceding sections. Moreover, the errors grows rapidly as the number of users increases. For relatively small numbers of users (e.g., fewer than 9) or large numbers of antennas (e.g., larger than 200), the performance of channel estimation for the UAs is much better than that using a collocated UA; the spherical UA outperforms the circular UA. The conventional array’s incapability of accurate LI-channel estimation is mainly due to its confined geometry that is more suitable for estimating signals’ angles-of-arrival. For verification, it can observed from Fig. 5 that the average error for the collocated array is insensitive to the numbers of antennas and users, suggesting that they are not the performance limiting factors.

B. Data Transmission

Consider data transmission assuming perfect channel-state-information at the transmitter. The curves of sum throughput versus transmission power per user are plotted in Fig. 6 with the number of users fixed at 10. Multiuser transmission using a UA and a conventional array use the multiuser phase-mode (PM) and the conventional zero-forcing (ZF) schemes, respectively. For single-user transmission, both the UA and conventional
array adopt the well-known channel-conjugate (CC) transmission scheme. The plots account for different numbers of antennas, namely 100 and 400. Several observations are made by comparing the curves. First, the combination of spherical UA and PM transmission yields much higher sum throughput than any other combination since distributing antennas over a larger area allows the spherical UA to generate a much higher number of degrees of freedom for avoiding interference and enhancing received signal power compared with other arrays (see Remark 12). Second, the sum throughputs for CC transmission using three types of arrays are comparable and higher than those achieved by multiuser transmission using the circular UA and collocated UA in the power range of $-70$ to $-45$ dBm; beyond this range, interference resulting from CC transmission dominates noise, causing the corresponding sum throughputs to saturate. Last, increasing the number of antennas from 100 to 400 contributes approximately the same throughput gain, about 20 b/s/Hz, for different combinations of array and transmission scheme.

The curves of sum throughput versus number of users are plotted in Fig. 7 with the transmission power per user fixed at $-40$ dBm. For multiuser transmission, the sum throughputs are observed to increase approximately linearly with the increasing number of users or equivalently the increasing number of simultaneous data streams. Furthermore, the curve corresponding to the spherical UA has a slope much larger than those for the other types of arrays that are comparable. In contrast, operating in the interference limiting regime, the sum throughputs for single-user (CC) transmission using 100 antennas saturate and are observed to be insensitive to the increase of the number of users. This issue can be alleviated by deploying more antennas (400) that leads to a substantial throughput gain e.g., about 20 b/s/Hz for the number of users equal to 18. Nevertheless, the gains for multiuser transmission are much larger and about 60 b/s/Hz at the same number of users.

VII. CONCLUSION

Techniques have been designed for channel estimation and data transmission in the UA system by modeling the UA as a gigantic continuous circular/spherical array and assuming free-space propagation. It has been shown that the UA enables accurate estimation of multiuser channels even if single pilot symbols are used provided that the user-separation distances are sufficiently large. If orthogonal pilot sequences are used, channel estimation is found to be always close to perfect. For single-user data transmission using the UA, inter-user interference can be suppressed by increasing user-separation distances. Alternatively, interference can be nulled at the UA based on a novel design of multiuser phase-mode precoders. The resultant number of available degrees of freedom for interference nulling is shown to be proportional to the minimum user-separation distance. Furthermore, the spherical UA provides performance gain compared with the circular counterpart.

This work opens up several interesting directions for further research. First, the current analysis is based on the model of a continuous circular/spherical UA and targets users near the UA center. Generalizing the model and user locations makes the analysis more challenging and requires the development of new analytical techniques. Second, it is important to address practical issues in the design and analysis of UA communication techniques such as delay and error in message exchange between the UA elements and the presence of sparse scatterers. Furthermore, it is also interesting to design algorithms/protocols for resource allocation, broadband transmission, and power control for the UA systems.

APPENDIX A

BESSEL FUNCTIONS AND SPHERICAL HARMONICS

Bessel functions and spherical harmonics are extensively used in the subsequent analysis. In this Appendix, the functions are
defined and some key properties useful for the analysis are summarized.

A. Bessel Functions and Their Properties: Only Bessel functions of the first kind are needed in the analysis and referred to simply as Bessel functions. A Bessel function with an integer order \( n \) can be defined in an integral form as [26]:

$$ J_n(x) = \frac{1}{2\pi} \int_0^{2\pi} e^{ix \sin \varphi} d\varphi. \quad (55) $$

Several useful properties of Bessel functions are described as follows.

(B1) The Jacobi-Anger expansion decomposes an exponential function of a trigonometric function into its harmonics as follows [26]:

$$ e^{i \pi \cos \varphi} = \sum_{n=-\infty}^{\infty} j^n J_n(x) e^{i k \phi}. $$

(B2) Addition Theorem I [27, (6.61)]:

$$ J_n(R) e^{i n \omega} = \sum_{k=-\infty}^{\infty} J_k(a) J_{n+k}(b) e^{i k \phi} $$

where \( R = \sqrt{a^2 + b^2 - 2ab \cos \varphi} \) and \( \sin \omega = \frac{a}{R} \sin \varphi \).

(B3) Addition Theorem II rewritten from [27, (6.62)]:

$$ \text{sinc}(R) = \frac{\pi}{2} \sum_{n=0}^{\infty} (2n+1) J_{n+\frac{1}{2}}(a) J_{n+\frac{1}{2}}(b) P_n(\cos \varphi) $$

where \( P_n(x) \) with \( x \in [-1, 1] \) is the Legendre polynomial defined as

$$ P_n(x) = \frac{1}{2^n n!} \frac{d^n}{dx^n} (x^2 - 1)^n. \quad (56) $$

(B4) The Bessel function \( J_0(x) \) can be upper bounded as \( J_0(x) \leq \nu x^{-\frac{1}{4}} \) where the constant \( \nu = 0.7857 \cdots \) [28].

(B5) Given \( 0 < z < 1 \), a Bessel function with a high order satisfies [29, 9.3.5 and 9.3.6]1

$$ J_n(zn) \sim \frac{c(z)}{n^\nu}, \quad n \to \infty $$

where \( c(z) \) is a positive constant whose value depends only on \( z \). Consequently, for \( x \gg 1 \) and \( |n| \geq x \), \( J_n(x) \approx 0 \) [17].

(B6) Gegenbauer’s generalization of Poisson’s integral [29, 10.1.14]:

$$ \sqrt{\frac{\pi}{2x}} J_{n+\frac{1}{2}}(x) = \frac{1}{2} (-j)^n \int_{-1}^{1} e^{ij\tau} P_n(\tau) d\tau $$

where \( P_n(x) \) is the Legendre polynomial defined in (56).

---

B. Spherical Harmonics and Their Properties: The spherical harmonic functions denoted as \( \{Y^m_\ell(\varphi, \theta)\} \) with integer indices \( \ell = 0, 1, \cdots \) and \( -\ell \leq m \leq \ell \) are defined as [26]:

$$ Y^m_\ell(\varphi, \theta) = \sqrt{\frac{2\ell + 1}{4\pi} \frac{(\ell - m)!}{(\ell + m)!} \frac{1}{P_\ell} P^m_\ell(\cos \theta) e^{im\varphi} \quad (57) $$

where \( \theta \in [0, \pi], \varphi \in [0, 2\pi] \) and \( P^m_\ell(\cos \theta) \) represents the associated Legendre polynomial defined as

$$ P^m_\ell(x) = \frac{(1-x^2)^{\frac{m}{2}}}{2^m \ell!} \frac{d^m}{dx^m} (x^2-1)^\ell, \quad -1 \leq x \leq 1. $$

Note that \( P^m_\ell(x) \) with \( m = 0 \) reduces to the Legendre polynomial in (56).

Several useful properties of the spherical harmonics are described as follows.

(S1) The functions \( \{Y^m_\ell(\varphi, \theta)\} \) are orthonormal over the spherical surface:

$$ \int_0^{2\pi} \int_0^\pi \! |Y^m_\ell(\varphi, \theta)|^2 Y^m_n(\varphi, \theta) \sin \theta \, d\theta d\varphi = \delta_{m,m'} \delta_{\ell,\ell'} $$

where \( \delta_{m,m'} \) is equal to 1 if \( m = m' \) and 0 otherwise.

(S2) Funk-Hecke Theorem [30, Theorem 3]: Let \( \psi_u(\varphi, \theta) \) denote the angle between the vectors \( X_u = (r_u, \varphi_u, \theta_u) \) and \( (1, \varphi, \theta) \in \mathbb{R}^3 \). The Laplace series of a function \( w(\cos \psi_u(\varphi, \theta)) \) is given as

$$ w(\cos \psi_u(\varphi, \theta)) = \sum_{\ell=0}^{\infty} \sum_{m=-\ell}^{\ell} c_{\ell}^m (\varphi_u, \theta_u) Y^m_\ell(\varphi, \theta) $$

where the coefficient \( c_{\ell}^m (\varphi_u, \theta_u) = c_{\ell} Y^m_\ell(\varphi_u, \theta_u) \) with

$$ c_{\ell} = 2\pi \int_{-1}^{1} (w(\tau) P_\ell(\tau)) d\tau. $$

(S3) Spherical Harmonic Addition Theorem [26]:

$$ \sum_{m=-\ell}^{\ell} [Y^m_\ell(A)]^* Y^m_\ell(X_u) = \frac{2\ell + 1}{4\pi} P_\ell(\cos \psi_u(A)) $$

where \( A, X_u \in \mathbb{R}^3 \) and \( \psi_u(A) \) denotes their angle.

---

APPENDIX B

PROOFS OF LEMMAS

A. Proof of Lemma 1: By substituting (4) and \( \psi_u(A) = \varphi_u - \varphi \) into (8),

$$ Q_k = \frac{\lambda e^{-\frac{1}{2} \varphi_0}}{4\pi r_0} \sum_{u=1}^{l} \int_0^{2\pi} e^{i \varphi_r \cos(\varphi - \varphi_u)} e^{i k \varphi} d\varphi $$

$$ + \frac{1}{2\pi} \int_0^{2\pi} \varphi \varphi e^{i k \varphi} d\varphi + \frac{\alpha_r}{r_0}. \quad (58) $$

The noise term can be written as

$$ \frac{1}{2\pi} \int_0^{2\pi} \varphi \varphi e^{i k \varphi} d\varphi = \lim_{N \to \infty} \sum_{n=1}^{N} \int_0^{2\pi} \left( \frac{2\pi n}{N} \right) e^{i k \frac{2\pi n}{N}}. \quad (59) $$

---

1 Two functions \( f \) and \( g \) are asymptotic equivalent, denoted as \( f(x) \sim g(x) \), if \( \lim_{x \to -\infty} f(x)/g(x) = 1 \).
For ease of notation, define $\tilde{z}_n = z \left( \frac{2\pi}{N} \right) e^{j \frac{2\pi}{N} n}$. Since $z_0, z_1, \ldots, z_N$ is an i.i.d. sequence of $CN\left(0, \sigma^2\right)$ random variables under Assumption 2, by applying the law of large numbers, it follows from (59) that

$$
\frac{1}{2\pi} \int_0^{2\pi} z(\varphi) e^{j k \varphi} d\varphi = 0, \text{ a.s.}
$$

(60)

Next, based on the Jacobi-Anger expansion in Property (B1) of Bessel functions in Appendix A, the first exponential term in (58) can be decomposed as

$$
e^{j \frac{2\pi}{N} r_u \cos (\varphi_u - \varphi)} = \sum_{n=0}^{\infty} \frac{(-1)^n}{2^n} J_n \left( \frac{2\pi}{\lambda} r_u \right) e^{j n (\varphi_u - \varphi)}.
$$

(61)

By substituting (60) and (61) into (58), it can be obtained that

$$Q_k = \frac{k e^{-j \frac{2\pi}{N} r_u}}{4 \pi r_0} \sum_{u=1}^{U} \sum_{n=0}^{\infty} J_n \left( \frac{2\pi}{\lambda} r_u \right) \times \int_0^{2\pi} e^{j n (\varphi_u - \varphi)} (\varphi_u - \varphi) d\varphi + \frac{\alpha_r}{r_0}, \text{ a.s.}
$$

The desired result follows since $\frac{1}{2\pi} \int_0^{2\pi} e^{j (k-n)\varphi} d\varphi = \delta_{k,n}$. □

**B. Proof of Lemma 3:** Using trigonometric identities, it can be obtained that

$$r_u \cos (\varphi_u - \theta) - r_k \cos (\varphi_u - \theta) = |X_u - X_k| \cos (\beta_{u,k} + \theta)
$$

where the angle $\beta_{u,k}$ is defined by

$$\tan \beta_{u,k} = \frac{r_u \cos (\varphi_u) - r_k \cos (\varphi_k)}{r_u \sin (\varphi_u) - r_k \sin (\varphi_k)}.
$$

It follows that

$$
\frac{1}{2\pi} \int_0^{2\pi} e^{j \frac{2\pi}{N} (r_u \cos (\varphi_u - \theta) - r_k \cos (\varphi_u - \theta))} d\theta = \frac{1}{2\pi} \int_0^{2\pi} e^{j \frac{2\pi}{N} |X_u - X_k| \sin (\beta_{u,k} + \theta + \frac{\pi}{2})} d\theta.
$$

Using the definition of Bessel function in (55),

$$\frac{1}{2\pi} \int_0^{2\pi} e^{j \frac{2\pi}{N} (r_u \cos (\varphi_u - \theta) - r_k \cos (\varphi_u - \theta))} d\theta = j e^{j \beta_{u,k}} J_0 \left( \frac{2\pi}{\lambda} |X_u - X_k| \right).
$$

(62)

Substituting (62) into (22) gives

$$g(X_k | X_u) = \sqrt{P_1 \frac{2\pi}{2r_0}} j e^{j \beta_{u,k}} J_0 \left( \frac{2\pi}{\lambda} |X_u - X_k| \right) + \frac{\alpha_r}{\sqrt{r_0}}.
$$

The desired result follows. □

**C. Proof of Lemma 5:** By substitution of the training signal in (4), its Laplace coefficients defined in (37) are obtained as

$$Q^m = \frac{\pi}{r_0} \sum_{u=1}^{U} \int_0^{2\pi} e^{j \frac{2\pi}{N} r_u \cos (\varphi_u, \theta)} \sin \varphi d\varphi d\theta \times \frac{\alpha_r}{r_0}, \text{ a.s.}
$$

where the vanishing of noise follows similar analysis as in the proof for Lemma 1. Using the Funk-Hecke Theorem in (52), the exponential term in the last equation can be also expanded into a Laplace series as

$$e^{j \frac{2\pi}{N} r_u \cos \psi} = \sum_{\ell=0}^{\infty} \sum_{m=-\ell}^{\ell} c_{\ell} Y^m_m (\varphi_u, \theta_u) Y^m_m (\psi, \theta)
$$

(63)

where

$$c_{\ell} = 2\pi \int_0^{2\pi} e^{j \frac{2\pi}{N} r_u \tau} P_1 (\tau) d\tau.
$$

(64)

Using Property (B6) in Appendix A, it follows from (64) that

$$c_{\ell} (r_u) = \frac{(2\pi)^2 j^\ell J_+ (\frac{2\pi}{\lambda} r_u)}{(2\pi)^2 r_u^2} Y^m_m (\varphi_u, \theta_u).
$$

(65)

Combining (63) and (65) gives the desired result. □

**D. Proof of Lemma 7:** Since $\psi_u$ is the angle between two unit vectors with spherical coordinates $(1, \varphi, \theta)$ and $(1, \varphi_u, \theta_u)$,

$$\cos \psi_u = \sin \varphi \cos \theta \sin \varphi_u \cos \theta_u + \sin \varphi \sin \theta \sin \varphi_u \sin \theta_u + \cos \varphi \cos \varphi_u.
$$

(66)

It follows that

$$r_u \cos \psi_u - r_k \cos \psi_k
$$

$$= \sin \varphi \cos \theta (r_u \sin \varphi_u \cos \theta_u - r_k \sin \varphi_k \cos \theta_k)
$$

$$+ \sin \varphi \sin \theta (r_u \sin \varphi_u \sin \theta_u - r_k \sin \varphi_k \sin \theta_k)
$$

$$+ \cos \varphi (r_u \cos \varphi_u - r_k \cos \varphi_k).
$$

(67)

To rewrite the right-hand side of (67) in a compact form, define the spherical coordinates $(u,k, \varphi_u, \theta_u) \in \mathbb{R}^3$ such that

$$r^2_{u,k} = (r_u \sin \varphi_u \cos \theta_u - r_k \sin \varphi_k \cos \theta_k)^2
$$

$$+ (r_u \sin \varphi_u \sin \theta_u - r_k \sin \varphi_k \sin \theta_k)^2
$$

$$+ (r_u \cos \varphi_u - r_k \cos \varphi_k)^2.
$$

(68)

and the angles $\varphi_{u,k}$ and $\theta_{u,k}$ as in the lemma statement. Moreover, let $\psi_{u,k}$ denote the angle between $(r_u, \varphi, \theta)$ and $(r_{u,k}, \varphi_{u,k}, \theta_{u,k})$. Using these definitions, (67) can be reduced to

$$r_u \cos \psi_u - r_k \cos \psi_k
$$

$$= r_{u,k} \cos \psi_{u,k}.
$$

(69)

It can be obtained from (68) that $r^2_{u,k} = |X_u - X_k|^2$. Thus,

$$r_u \cos \psi_u - r_k \cos \psi_k = |X_u - X_k| \cos \psi_{u,k}.
$$

(70)

Substituting this result into (43) yields

$$\mathcal{J}_{u,k,m,\ell} = \int \int e^{j \frac{2\pi}{N} |X_u - X_k| \cos \psi_{u,k}} \sin \varphi d\varphi d\theta.
$$

Applying Lemma 5 gives the desired result. □

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Kaibin Huang (S’05–M’08–SM’13) received the B.Eng. (first-class hon.) and the M.Eng. degrees from the National University of Singapore in 1998 and 2000, respectively, and the Ph.D. degree from the University of Texas at Austin (UT Austin) in 2008, all in electrical engineering.

Since January 2014, he has been an assistant professor in the Department of Electrical and Electronic Engineering (EEE) at The University of Hong Kong. He is an adjunct professor in the School of EEE at Yonsei University in South Korea. He used to be a faculty member in the Department of Applied Mathematics at The Hong Kong Polytechnic University (PolyU) and the Department of EEE at Yonsei University. He had been a Postdoctoral Research Fellow in the Department of Electrical and Computer Engineering at the Hong Kong University of Science and Technology from Jun. 2008 to Feb. 2009 and an Associate Scientist at the Institute for Infocomm Research in Singapore from Nov. 1999 to Jul. 2004. His research interests focus on the analysis and design of wireless networks using stochastic geometry and multi-antenna techniques.

He frequently serves on the technical program committees of major IEEE conferences in wireless communications. He chairs the Comm. Theory Symp. of IEEE GLOBECOM 2014 and the Adv. Topics in Wireless Comm. Symp. of IEEE/CIC ICC 2014 and has been the technical co-chair for IEEE Comm. Theory Workshop 2013, the track chair for IEEE Asilomar Conf. on Signals, Sys. and Computers 2011, and the track co-chair for IEEE Veh. Tech. Conf. Spring 2013 and IEEE Wireless Comm. and Networking Conf. 2011. He is a guest editor for the IEEE JOURNAL ON SELECTED AREAS IN COMMUNICATIONS, an editor for the IEEE TRANSACTIONS ON WIRELESS COMMUNICATIONS, IEEE WIRELESS COMMUNICATIONS LETTERS and also IEEE/KICS JOURNAL OF COMMUNICATION AND NETWORKS. He is an elected member of the SPCOM Technical Committee of the IEEE Signal Processing Society. Dr. Huang received the 2015 IEEE ComSoc Asia Pacific Outstanding Paper Award, Outstanding Teaching Award from Yonsei, Motorola Partnerships in Research Grant, the University Continuing Fellowship from UT Austin, and a Best Paper Award from IEEE GLOBECOM 2006 and PolyU AMA in 2013.

Jiayi Chen (S’08–M’16) received the B.S. and M.S. degrees from Sun Yat-sen University, Guangzhou, China, in 2006 and 2008, respectively, and the Ph.D. degree from City University of Hong Kong, Hong Kong, in 2013, all in electronic engineering. He was post-doctoral fellows at City University of Hong Kong and Hong Kong University in 2013 and 2014 respectively. Currently he is an assistant professor in Shenzhen University, Shenzhen, China.

His research interests include multi-user communications, information theory and advanced channel coding, communication signal processing and related topics of big data.

Vincent K. N. Lau (SM’04–F’12) obtained B.Eng. (Distinction 1st Hons.) from the University of Hong Kong (1989–1992) and Ph.D. from the Cambridge University (1995–1997). He joined Bell Labs from 1997–2004 and the Department of ECE, Hong Kong University of Science and Technology (HKUST) in 2004. He is currently a Chair Professor and the Founding Director of Huawei-HKUST Joint Innovation Lab at HKUST. He is also elected as IEEE Fellow, HKIE Fellow, Croucher Senior Research Fellow and Changjiang Chair Professor. His current research focus includes robust cross layer optimization for wireless systems, Massive MIMO, Compressed Sensing, Networked Control Systems as well as PHY Caching for Wireless Networks.