

# An Electromagnetic Information Theory Based Model for Efficient Characterization of MIMO Systems in Complex Space

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**Abstract**—It is the pursuit of a multiple-input–multiple-output (MIMO) system to approach and even break the limit of channel capacity. However, it is always a big challenge to efficiently characterize the MIMO systems in complex space and get better propagation performance than the conventional MIMO systems considering only free space, which is important for guiding the power and phase allocation of antenna units. In this article, an electromagnetic-information-theory (EMIT)-based model is developed for the efficient characterization of MIMO systems in complex space. The group- $T$ -matrix-based multiple scattering fast algorithm, the mode-decomposition-based characterization method, and their joint theoretical framework in complex space are discussed. First, key informatics parameters in free EM space based on a dyadic Green’s function are derived. Next, a novel group- $T$ -matrix-based multiple scattering fast algorithm is developed to describe a representative inhomogeneous EM space. All the analytical results are validated by simulations. In addition, the complete form of the EMIT-based model is proposed to derive the informatics parameters frequently used in EM propagation, by integrating the mode analysis method with the dyadic Green’s function matrix. Finally, as a proof-of-concept, microwave anechoic chamber measurements of a cylindrical array are performed, demonstrating the effectiveness of the EMIT-based model. Meanwhile, a case of image transmission with limited power is presented to illustrate how to use this EMIT-based model to guide the power and phase allocation of antenna units for real MIMO applications.

**Index Terms**—Complex space, electromagnetic information theory (EMIT), group  $T$  matrix, mode analysis, multiple-input–multiple-output (MIMO) system.

## I. INTRODUCTION

Typically, for antenna design, it is promising to maximize the channel capacity via a multiple-input–multiple-output (MIMO) system to approach the limit of channel

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capacity during propagation. Thus, under the demand for high accuracy and low latency nowadays, basic research on the efficient characterization of MIMO systems is very important [1]. On this basis, we can carry out further work such as the optimization solutions for the power and phase allocation of antenna units.

Previous works for MIMO characterization can be roughly clarified into two categories: electromagnetic (EM) methods and information theory (IT). The former mainly focuses on the radio frequency (RF) front-end design by solving Maxwell’s equations under different boundary conditions, consisting of the descriptions of the complex EM space [2], [3], while the latter mainly analyzes the channel properties under different probability models by using Shannon IT [4], [5]. The above two frameworks are faced with a major challenge in practical application: how to efficiently model MIMO systems in complex space to achieve better propagation performance than MIMO analysis that only consider free space.

For EM methods, the core step of intelligent designs nowadays is reconstructing the MIMO systems’ radiation patterns [6], [7], [8]. To consider the effect of the EM propagation space, full-wave numerical algorithms have been used to incorporate the RF front-end design and environment perception into the EM framework, consuming a lot of time [9]. To greatly reduce the calculation time of modeling EM space, some studies have proposed to use approximate methods like ray tracing (RT) [10], [11], [12]. However, this is often not acceptable due to a lack of high accuracy. Besides, the  $T$ -matrix can be easily used to characterize efficient MIMO in complex EM space, via combining multiple scattering equations [13], [14]. However, practical wireless communication often focuses on some informatic parameters (such as channel capacity), while the EM-only framework is incapable of efficiently extracting the informatic parameters in the complex EM space.

For IT, the common statistic model, such as the Rayleigh fading model, is a mathematical tool based on the assumption of rich scattering [15]. When it evolves to cluster models like geometry-based stochastic models (GBSMs), the EM space is equivalent to the clusters with different shapes or distributions for convenient characterization [16], [17]. However, the accuracy of those models will be reduced due to the

EM properties of the MIMO system. For example, the work in [18] complements numerical methods to make up for the problem of using only Fresnel approximation in airborne antenna design. Moreover, the main idea of the emerging intelligent reflective surface (IRS) is to lay out the controllable surfaces in free or complex EM space [19], [20], [21], [22], [23]. Due to the lack of efficient MIMO characterization, this technology is still in the trial stage. This suggests that many basic assumptions of the information-only framework need to be reconsidered.

Nowadays, the EMIT for the MIMO characterization attracts more attention, which is expected to solve the challenges mentioned above [24], [25], [26], [27]. Researchers point out that with the wide layout of the antenna array (e.g., internet of vehicles), it is expected to eliminate the step of channel estimation with the help of rich environmental information [28]. Some works have been done from the perspective of EM fields to study the degree of freedom of MIMO systems [29], [30]. There are also mathematical methods to model the source region and field region as two sets of orthogonal bases in Hilbert space, and then construct some characteristic parameters of the MIMO system [31], [32]. To integrate the RF front-end design in the EMIT framework, the surface currents of antenna elements are modeled as the point sources with orthogonal bases. For example, a model was established to build a channel matrix from the angle of coordinate transformation and orthogonal decomposition of EM plane wave expansion, applied in a holographic MIMO system [33], [34]. Additionally, the work in [35] contains the idea of deriving the channel limit of a MIMO system by the EM field method. Nevertheless, the above research works on EMIT mainly focus on free space or revealing the parameter mapping between two theories; efficient characterization algorithms and clear EM information analysis methods for complex EM space are still unexplored.

In this article, we develop an EMIT-based model to conduct the efficient characterization of MIMO systems in complex EM space. The proposed EMIT-based model uses the group  $T$  matrix algorithm and dyadic Green's function-based mode analysis method, filling the research gap of efficient characterization algorithms and clear EM information analyses. The main contributions of this article are described as follows.

- 1) The key parameters of the MIMO systems are extracted through the dyadic Green's function and matrix mode analysis. The information characteristics of the MIMO systems are described by the EM method, revealing some important conclusions and deducing the key information parameters and valuable conclusions of IT by means of EM methods.
- 2) A fast algorithm based on the group  $T$  matrix is developed to model the complex EM space. Since the algorithm has semianalytical characteristics and the classical  $T$  matrix can be stored, which provides a faster calculation compared with the traditional full-wave algorithm. In contrast to the RT and pilot-based methods for channel estimation, our EM algorithm can be easily integrated into EMIT due to its higher accuracy and

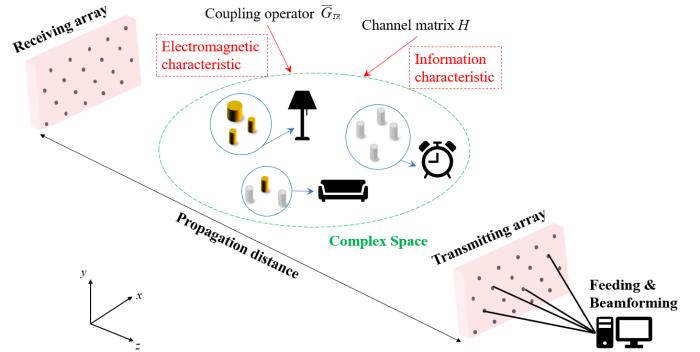


Fig. 1. System model of MIMO analysis in a complex space for indoor communication.

efficiency. In other words, the benefit of our proposed method is generated from the fast characteristics of the group  $T$  matrix and the EM analysis of the channel matrix (without the help of statistics).

- 3) The efficient EMIT-based model is proposed to characterize the MIMO systems in complex space. As a proof-of-concept, a microwave anechoic chamber measurement of a cylindrical array is taken as an example, demonstrating the effectiveness of the EMIT-based model for the MIMO mode analysis. Meanwhile, a case of image transmission with limited power is presented to illustrate how to guide the MIMO feeding based on the model, bringing a new insight into extracting information parameters using the basis of computational electromagnetism.

This article is organized as follows. The key informatics parameters based on the dyadic Green's function are derived in Section II. Then, the proposed EMIT-based model is analyzed in Section III. Experimental verification and an image transmission case were conducted in Section IV. Finally, the conclusion is drawn in Section V.

## II. SYSTEM MODEL AND KEY PARAMETERS

As shown in Fig. 1, consider a typical MIMO system including the transmitting and receiving array for indoor communication, whose overall communication performance will be affected by the propagation distance and the properties of complex space. In this section, the EM propagation space is designated as a free space for extracting key parameters of a MIMO system. More complex EM space is characterized in Section III.

To combine the coupling operator  $\bar{G}_{TR}$  and the channel matrix  $H$ , a series of isotropic point sources are placed in the transmission volume and the receiving volume, with the position vectors  $\mathbf{r}_T$  and  $\mathbf{r}_R$  respectively. The EM wave received is defined as  $\psi_{outR}$ , thus the Helmholtz wave equation is given by

$$\nabla \times \nabla \times \psi_{outR} - k_0^2 \psi_{outR} = i\omega \mu_0 \mathbf{J}_{incT} \quad (1)$$

where  $\mathbf{J}_{incT}$  is the transmitted source and  $k$  is the wave vector. To solve this equation, the dyadic Green's function  $\bar{G}$  operator

based on the impulse function idea is introduced

$$\bar{\mathbf{G}}(\mathbf{r}_R, \mathbf{r}_T) = \left( \bar{\mathbf{I}} + \frac{\nabla \nabla}{k^2} \right) \frac{\exp[ik|\mathbf{r}_R - \mathbf{r}_T|]}{4\pi|\mathbf{r}_R - \mathbf{r}_T|} \quad (2)$$

where  $\bar{\mathbf{I}}$  is the unit tensor. Since the dyadic Green's function tensor  $\bar{\mathbf{G}}$  contains the scalar Green's functions

$$\bar{\mathbf{G}} = \begin{bmatrix} G_{xx} & G_{xy} & G_{xz} \\ G_{yx} & G_{yy} & G_{yz} \\ G_{zx} & G_{zy} & G_{zz} \end{bmatrix}. \quad (3)$$

When it comes to the two independent single-polarization situations at the far-field, the coupling operator is able to be simplified into the scalar Green's function without loss of accuracy. Therefore, the element  $h_{ij}$  in the channel matrix  $\mathbf{H}$  will be changed to the following form in the case of a single polarized source:

$$h_{ij} = \frac{\exp[ik|\mathbf{r}_{Ri} - \mathbf{r}_{Tj}|]}{4\pi|\mathbf{r}_{Ri} - \mathbf{r}_{Tj}|} = g_{0(ij)} \quad (4)$$

where  $g_0$  is the scalar Green's function, that is, the special form of  $\bar{\mathbf{G}}$  in the case of single polarization. It is worth mentioning that the channel matrix  $\mathbf{H}$  at this time is not normalized, so it contains path loss.

Assume that the number of transmitting source points is  $N_T$ , and the number of receiving field points is  $N_R$ . Therefore, the transmitting source can be expressed as a  $N_T * 1$  matrix, the receiving electric field as a  $N_R * 1$  matrix, and the coupling operator  $\bar{\mathbf{G}}$  of the EM space as a  $N_R * N_T$  matrix. By introducing the Dirac notation, the MIMO propagation relation of free space is expressed as

$$|\psi_{outR}\rangle = \bar{\mathbf{G}}|\mathbf{J}_{incT}\rangle. \quad (5)$$

To normalize the channel matrix, we defined the normalized coupling operator  $\bar{\mathbf{G}}_{TR}$  as  $\bar{\mathbf{G}}_{TR} = \alpha \bar{\mathbf{G}}$ , where  $\alpha$  is a normalization factor, making  $E[\|\bar{\mathbf{G}}_{TR}\|_F^2] = N_T N_R$ .  $E[\cdot]$  denotes the expectation and  $\|\cdot\|_F$  means the Frobenius norm. The physical meaning of this normalization is that every subchannel should have a unity average channel gain.

According to the Hermitian nature  $\bar{\mathbf{G}}_{TR} \bar{\mathbf{G}}_{TR}^\dagger$ , singular value decomposition (SVD) of the coupling operator could conduct mode analysis of MIMO EM propagation, where  $\dagger$  denotes the conjugate transpose

$$\bar{\mathbf{G}}_{TR} = \mathbf{U}_R \mathbf{S} \mathbf{V}_T^\dagger \quad (6)$$

where  $\mathbf{V}_T$  ( $\mathbf{U}_R$ ) is a  $N_T * N_T$  ( $N_R * N_R$ ) matrix, and each column represents the EM eigenvector of the transmitting sources (receiving fields). Because of the unitary nature of the SVD eigenmatrix, it is known that each column is strictly orthogonal, which is called the EM space mode. The weight of each pattern is determined by the corresponding element in the diagonal matrix  $\mathbf{S}$ . The combinations of those orthogonal modes form two Hilbert spaces, and therefore the coupling operator  $\bar{\mathbf{G}}_{TR}$  builds a mapping between the transmitting Hilbert space and the receiving Hilbert space, which is an important property in the subsequent discussion.

As we all know, the upper limit of information transmission per bandwidth in MIMO systems is also limited by Shannon's

formula [36]

$$\begin{aligned} C &= E \left\{ \log_2 \left[ \det \left( \mathbf{I} + \frac{\rho}{n} \frac{1}{N_t} \bar{\mathbf{G}}_{TR} \bar{\mathbf{G}}_{TR}^\dagger \right) \right] \right\} \\ &= \sum_i \log_2 \left( 1 + \frac{\rho}{n} \sigma_i^2 \right) \end{aligned} \quad (7)$$

where  $\mathbf{I}$  is the identity matrix and  $\sigma_i$  are the singular values of  $(1/\sqrt{N_t})\bar{\mathbf{G}}_{TR}$ . Apparently,  $\sigma_i^2$  is the decisive parameter of key information-carrying capacity in the MIMO system at a given signal-to-noise ratio (SNR)  $\rho/n$ . Besides, we can drop the expectation  $E[\cdot]$  in (7) and no longer need to make a special distinction for large-scale and small-scale path loss and fading, because the amplitude and phase changes of the electric field have been included in the operator  $\bar{\mathbf{G}}_{TR}$ .

It is seen from (6) and (7) that the singular value of EM propagation space determines the number and weight of independent modes, which establishes a corresponding relationship with the number of independently available channels and path loss of wireless communication. We give the key informatics parameters of a MIMO system by referring to the effective rank idea of existing work [21]

$$C_{eff} = \prod_{i=1}^{\min(N^R, N^T)} \exp(-\sigma'_i \ln(\sigma'_i)) \quad (8)$$

where  $C_{eff}$  represents the EM effective capacity,  $\sigma'_i = \sigma_i / (\sum \sigma_i)$  represents the normalized singular values of  $\bar{\mathbf{G}}_{TR}$ . Hence, (8) establishes the mapping relationship between the dyadic Green's function matrix and typical informatics parameters, which is an important tool for the MIMO mode analysis.

To understand how this approach works, both mathematically and physically, we set up an  $N * N$  MIMO system with the same EM space properties, as shown in Fig. 1. In fact, in practical engineering applications, the mutual coupling is concerned not because it affects the EM equivalent capacity, but because it affects the radiation efficiency and SNR of the antennas. The transmitting and receiving antennas are modeled as isotropic point sources/receivers (delta function basis), which is a widely used assumption in EMIT. From the EM perspective, the antennas can also be modeled as continuous surface (equivalent) currents by the rooftop or Rao-Wilton-Glisson (RWG) basis, as frequently utilized in the methods of moments (MoM). Different basis representations of the currents, related to different antenna designs, will not influence the estimations of the effective degree of freedom limit. Since we want to focus the analysis in this work on solving dyadic Green's function and extracting informatics parameters in a complex space, we choose the model carefully to avoid mutual coupling. To construct the basic framework of EMIT, three key parameters (the number of sources, communication distance, and the size of the antenna aperture) are considered to illustrate the relationship between RF front-end devices' design and the effective capability of the MIMO system.

In Fig. 2, the relationship between the EM effective capacity and the number of sources is presented at a given aperture ( $6\lambda * 6\lambda$ ), showing clearly that with the increase of  $N$ , the EM effective capacity under different communication distances

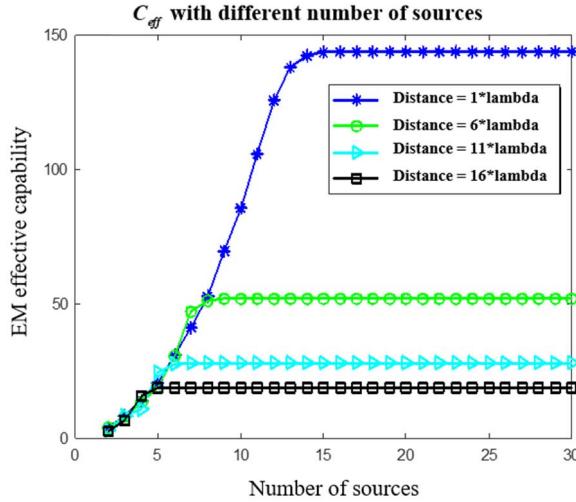


Fig. 2. EM effective capability with the change of number of sources in four communication distances.

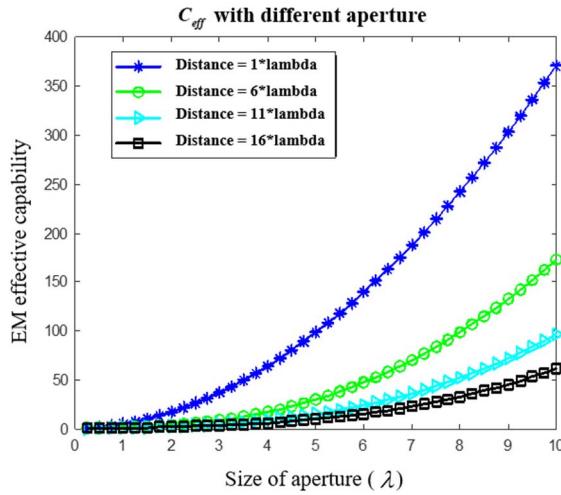


Fig. 3. EM effective capability with the change of aperture sizes in four communication distances.

will increase with the same slope, but it converges to the channel capacity. In this case, considering that the change of the total power of the transmitting array will lead to different channel capacities, we fixed the total transmitting power at  $P_0$ , satisfying  $\sum_{i=1}^{N_T} P_i = P_0$ .

In addition, to illustrate the physical nature of the convergence, Fig. 3 shows the EM effective capacity corresponding to different aperture sizes with enough point sources (30\*30). Obviously, the size of the aperture plays a determinant role in the information capacity of MIMO systems. It suggests that the trend of antenna miniaturization is the weakening of maximum carrying information, which cannot be solved by multiantenna technology. Besides, in Fig. 4, we plot the curve of EM effective capacity changing with the communication distance, revealing the characteristics of wireless communication energy attenuated with the propagation distance from the perspective of dyadic Green's function. Besides, in Figs. 3 and 4, the variables we focus on are the aperture and distance respectively, so the number of point sources is a constant, and the total power  $P_0$  always remains a constant.

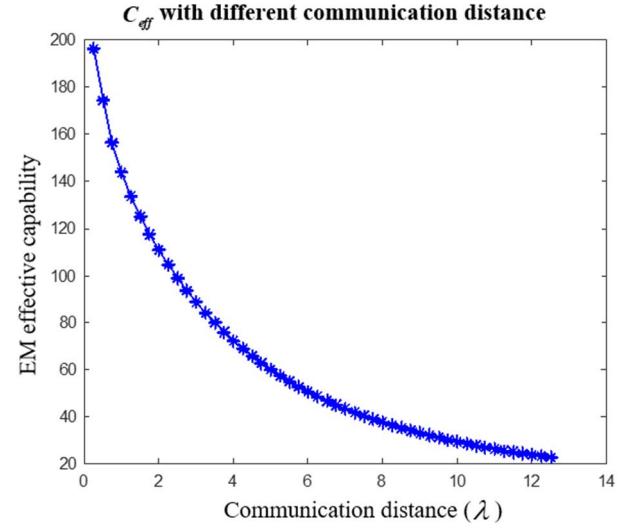


Fig. 4. Attenuation of EM effective capability with the change of communication distance.

It is worth mentioning that, due to the basis function decomposition method (such as RWG basis in MoM) commonly used in computational EMs, the specific RF front-end structure can be decomposed into the sum of point sources through grid partitioning, and the multichannel coupling effect will be considered in the coefficient term of the operator  $\bar{G}_{TR}$ . Therefore, when using the above method to perform theoretical modeling of EMIT, the coupling can be characterized by adding a coefficient term to the operator  $\bar{G}_{TR}$ , and the specific physical dimensions of the RF front-end can be numerically quantified by base function equivalence.

Essentially, changing the RF front-end or the channel will affect the value of  $C_{eff}$  in (8) by affecting the distribution of  $\sigma_i$  on the  $i$ th channel. In other words,  $C_{eff}$  and the distribution of  $\sigma_i$  are the inherent property of the communication system.

However, the core assumption of this part is based on free EM space, and the specific form of coupling operator  $\bar{G}_{TR}$  will change when numerous scatterers are introduced. The next section will demonstrate the fast algorithms for characterizing the complex EM space.

### III. PROPOSED EMIT-BASED MODEL FOR EM COMPLEX SPACE

In some typical wireless communication scenarios, objects in complex scattering environments are usually represented by some types of scatterers for convenient EM calculations, among which one of the commonly-used classical models is the cylindrical array, as described in Fig. 5. For example, a vehicle-to-vehicle channel is equivalent to a scattering cluster in the internet of vehicles channel modeling [16]. Due to the poor accuracy and long response time of traditional channel measurement schemes, this section proposes an EMIT-based model for efficient MIMO characterization in this typical scattering complex space based on the group  $T$  matrix.

#### A. Algorithm Description

$N$  cylindrical scatterers in MIMO EM propagation space are considered, which are centered at  $r_p$  ( $p = 1, 2, \dots, N$ )

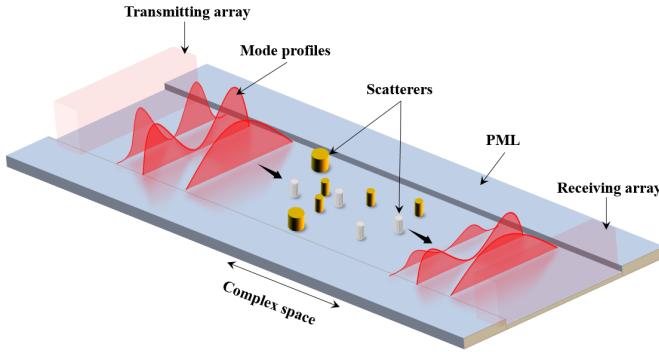


Fig. 5. Schematic of MIMO mode analysis of complex space. The material, position, quantity, and shape of the scatterers can be set arbitrarily in our algorithm.

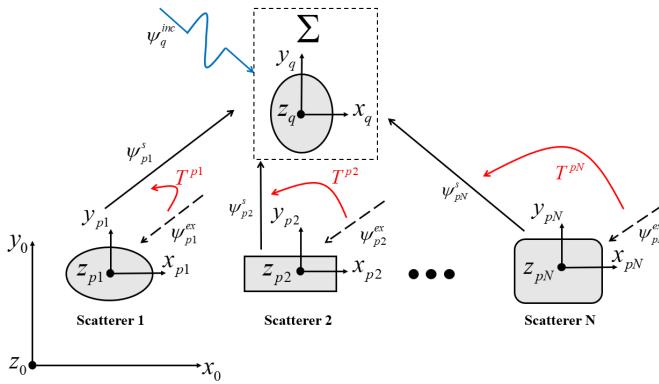


Fig. 6. Illustration of the group-T-matrix-based algorithm. The distribution of the field around the scatterer is decomposed, and the steady-state coefficient matching is carried out based on the cylindrical wave expansion without meshwork and time-domain iteration.

and are with radius  $a_p$  ( $p = 1, 2, \dots, N$ ). These parameters can be easily substituted to simulate different distributions and different shapes of scatterers. For the description of the RF front-end, we use a  $N_s \times 1$  dipole antenna array, coordinate  $r_s$  ( $s = 1, 2, \dots, N$ ), as a convenient MIMO model. The overall algorithm framework is shown in Fig. 6. To be clear, we focus on the scenarios where the transceivers and receivers are in the same horizontal plane (such as vehicle-to-vehicle communication and indoor point-to-point communication). In this case, we can regard the scatterer as a cluster of cylindrical scatterers, so conducting cylindrical wave expansion is reasonable and convenient. This benefits the convenience of calculation and the simplicity of the model.

To take the coupling between scatterers into account, we take the  $q$ th scatterer as the analysis object and decompose the total external field  $\psi_q^{\text{ex}}$  around it into the sum of the incident field  $\psi_q^{\text{inc}}$  and the scattering field  $\psi_p^s$  of the rest scatterers

$$\psi_q^{\text{ex}} = \psi_q^{\text{inc}} + \sum_{\substack{p=1 \\ p \neq q}}^N \psi_p^s. \quad (9)$$

For solving the scattered fields, the electric field is expanded as a vector cylindrical wave harmonic function

$$\psi_q^{\text{ex}} = \sum_n I_n^{(q)} Rg \psi_n(k(\mathbf{r} - \mathbf{r}_q)) \quad (10)$$

where  $k$  is the wave vector,  $I_n^{(q)}$  is the cylindrical wave coefficient, which is the unknown core quantity for solving the field distribution.

In addition, the specific mathematical form of cylindrical wave expansion in (10) is given as follows:

$$\begin{aligned} \psi_n(k(\mathbf{r} - \mathbf{r}_p)) &= H_n^{(1)}(k|\mathbf{r} - \mathbf{r}_p|) \exp(in\phi_{\mathbf{r}\mathbf{r}_p}) \\ Rg \psi_n(k(\mathbf{r} - \mathbf{r}_p)) &= J_n(k|\mathbf{r} - \mathbf{r}_p|) \exp(in\phi_{\mathbf{r}\mathbf{r}_p}) \end{aligned} \quad (11)$$

where  $\mathbf{r}$  represents the coordinate vector of the field point,  $\phi_{\mathbf{r}\mathbf{r}_p}$  represents the angle between the vectors  $\mathbf{r}$  and  $\mathbf{r}_p$ ,  $J_n$  is the Bessel function of order  $n$ ,  $H_n^{(1)}$  is the Hankel function of order  $n$ , and  $Rg$  means regularization. Later, we will use the symbol  $i$  to represent the imaginary unit.

For the mode matching, we perform the same cylindrical wave expansion for the incident field  $\psi_q^{\text{inc}}$  determined by the MIMO RF front-end (here is the  $N_s \times 1$  dipole array), obtaining

$$\psi_q^{\text{inc}} = \sum_{s=1}^{N_s} \frac{i}{4} H_0^{(1)}(k|\mathbf{r} - \mathbf{r}_s|). \quad (12)$$

To obtain the same expansion form as (10), the vector addition theorem is used to further expand (12) to obtain

$$\begin{aligned} \psi_q^{\text{inc}} &= \frac{i}{4} \sum_{n=1}^{N_s} \sum_{s=1}^{N_s} H_n^{(1)}(k|\mathbf{r} - \mathbf{r}_s|) \exp(-in\phi_{\mathbf{r}\mathbf{r}_q}) \\ &\quad \times Rg \psi_n(k(\mathbf{r} - \mathbf{r}_s)). \end{aligned} \quad (13)$$

In (13), as the RF front-end information is part of prior knowledge, the expansion coefficient of  $\psi_q^{\text{inc}}$  is determined, which is convenient for solving  $I_n^{(q)}$  in (10). Next, we write the scattering field  $\psi_p^s$  of the  $p$ th scatterer as follows:

$$\psi_p^s = \int_{dS_n} [\bar{\mathbf{G}}_{TR} \cdot i\omega\mu_0 \mathbf{J}(\mathbf{r}_p) - \nabla \times \bar{\mathbf{G}}_{TR} \cdot \mathbf{M}(\mathbf{r}_p)] dS' \quad (14)$$

where  $\bar{\mathbf{G}}_{TR}$  is dyadic Green's function illustrated in (2),  $\omega$  is the angular frequency,  $\mathbf{J}(\mathbf{r}_p)$  and  $\mathbf{M}(\mathbf{r}_p)$  are the current density and magnetic current density at the  $p$ th scatterer, respectively. The EM variation in the complex space is described by the action of the coupling operator  $\bar{\mathbf{G}}_{TR}$  on  $\mathbf{J}(\mathbf{r}_p)$  and  $\mathbf{M}(\mathbf{r}_p)$ . When  $\mathbf{J}(\mathbf{r}_p)$  and  $\mathbf{M}(\mathbf{r}_p)$  do not exist, (9) will then degenerate into the free space case shown in Section II.

In order to solve the  $\psi_p^s$  described in (14), we use the consistent mathematical form of  $\psi_q^{\text{ex}}$  on different scatterers to expand  $\psi_p^s$  into a cylindrical waveform by using (10), and the transformation relationship is shown in the red curve in Fig. 6. Thus, (14) can be rewritten as follows based on the group-T-matrix

$$\psi_p^s = \sum_m T_m^{(p)} I_m^{(p)} Rg \psi_m(k(\mathbf{r} - \mathbf{r}_s)) \quad (15)$$

where  $T^{(p)}$  is the group-T-matrix representing the relationship between the incident field and scattering field of the  $p$ th clustered scatterer, and its characteristics are only related to the shape and material of the current scatterer. Assuming the internal wave vector of the scatterer is  $k_p$ , the general form of group-T-matrix in the cylindrical coordinate system can be

obtained by using analytical methods

$$T_m^{(p)} = \frac{k_p J'_m(ka_p) - J_m(ka_p)k J'_m(ka_p)}{k H_m^{(1)}(ka_p)J_m(ka_p) - H_m^{(1)}(ka_p)k_p J'_m(ka_p)}. \quad (16)$$

The  $T$ -matrix of any shape object can be solved by numerical methods such as the MoM according to (14).

The basic purpose of this section is to illustrate the algorithm's efficiency, and thus we consider the model of dipole array with TM polarized waves incident on a perfect electric conductor (PEC). In this case, (16) evolves into

$$T_m^{(p)} = -\frac{J_m(ka_p)}{H_m^{(1)}(ka_p)}. \quad (17)$$

Substitute (17) into (15) to obtain the field distribution with  $I_m^{(p)}$  as the only variable. The matrix equation of the unknown coefficient  $I_m^{(p)}$  can be obtained by combining (9), (10), (13), and (15)

$$\mathbf{Z} \cdot \mathbf{I} = \mathbf{V}. \quad (18)$$

Here, in order to solve the coefficient  $I_m^{(p)}$ , the equations with different scatterers are written in matrix form, and the order of the Bessel function is truncated with the truncation number  $N_{\max}$ . Therefore,  $\mathbf{Z}$  is a square matrix of dimension  $(2N_{\max} + 1)N$ , while  $\mathbf{V}$  is a  $(2N_{\max} + 1)N \times 1$  vector. The specific form is

$$[\mathbf{Z}]_{(q-1) \times N+n, (p-1) \times N+m} = \begin{cases} -1, & p = q \\ T_m^{(p)} H_{n-m}^{(1)}(k|\vec{r}_p - \vec{r}_q|) \exp(-i(n-m)\phi_{\vec{r}_p \vec{r}_q}), & p \neq q \end{cases} \quad (19)$$

$$[\mathbf{V}]_{(q-1) \times N+n} = -\frac{i}{4} \sum_{s=1}^{N_s} H_n^{(1)}(k|\vec{r}_s - \vec{r}_q|) \exp(-i n \phi_{\vec{r}_s \vec{r}_q}). \quad (20)$$

In (19) and (20),  $\mathbf{Z}$  is determined by the properties of the complex space, and  $\mathbf{V}$  is determined by the properties of the RF front-end. A joint solution can semianalytically describe the evolution of the MIMO coupling operator  $\bar{\mathbf{G}}_{EIT}$ . Therefore, due to the change in the EM space coupling operator, (5) will be rewritten as

$$|\psi_{outR}\rangle = \bar{\mathbf{G}}_{EIT} |\mathbf{J}_{incT}\rangle. \quad (21)$$

The subsequent analysis only needs to be carried out in the same way as (6)–(8) to complete efficient MIMO characterization in complex space. It should be noted that, when facing the time-varying channel scenario, it will be very convenient to rewrite the coupling operator  $\bar{\mathbf{G}}_{EIT}$  into the form based on the time-domain Green's function.

### B. Numerical Results

To verify the accuracy and efficiency of the proposed EMIT-based model, numerical calculations of some specific scenarios are carried out and compared with full-wave simulation results.

Fig. 7 presents the field distributions of three simple arrays. The effectiveness of the EMIT-based model is verified by comparing the full-wave FDTD algorithm [Fig. 7(a), (c), (e)]

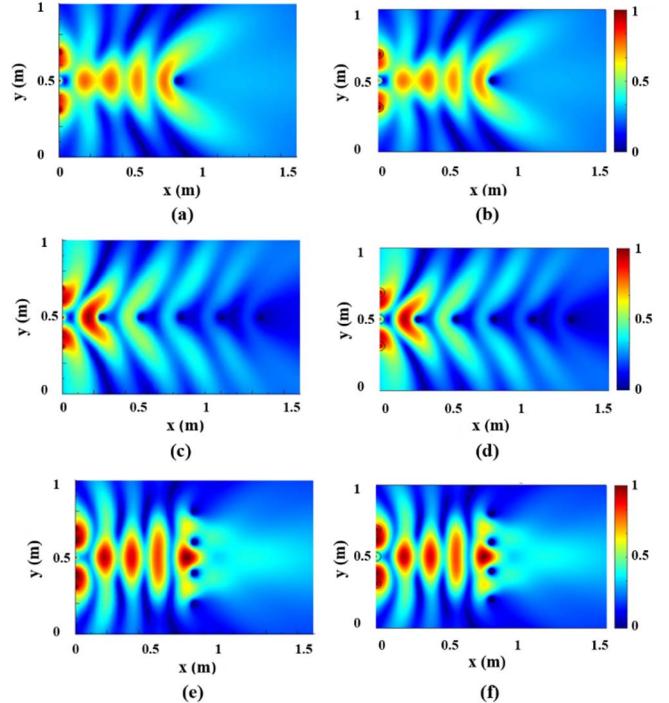


Fig. 7. Normalized electric field distribution on the validation plane. (a), (c), and (e) FDTD solver and (b), (d), and (f) proposed EMIT-based model for scatterers distribution of 1\*1, 1\*5 and 4\*1 respectively.

with the proposed algorithm [Fig. 7(b), (d), (f)]. We consider an EM space of 1\*1.5 m, where the MIMO system is modeled as a 3\*1 dipole array with an aperture of 0.75 m. The scatterer array element is modeled as a metal cylinder with a height of 0.25 m and a radius of 0.015 m, and the boundary is set as PEC. Due to the dense mesh division of the full-wave algorithm, its application is severely limited. However, the proposed semianalytic algorithm based on group- $T$ -matrix is suitable for various frequencies because it does not need mesh division. To obtain the comparison results, we first define the operating frequency at 915 MHz in this section.

Fig. 7 illustrates that the proposed EMIT-based model has achieved good results and can accurately describe the complex space. In order to further demonstrate its efficiency, the scatterer distribution was adjusted, and the solving time and error of the EMIT-based model and FDTD were calculated by analyzing the field intensity curve at the RF back-end, as shown in Table I. It is worth noting that the 10\*15 distribution cannot fully explain the difference between the two algorithms, because a large number of FDTD meshes converge extremely fast due to the inability of the electric field to propagate effectively, and the solutions are often mediocre at this time. Therefore, we further consider the case of a random array, that is, randomly removing 60 scatterers from the 10\*15 scatterer distribution. Besides, we clarify that the running time of our proposed algorithm mainly depends on the number of scatters. Therefore, the proposed EMIT-based model has higher computational efficiency than full-wave algorithms like FDTD, which provides great convenience for the description of complex space. But generally, the complexity of the real-world environment increases with the communication distance. In this

TABLE I  
COMPARISON OF CPU TIME AND RMS ERROR BETWEEN  
FDTD AND PROPOSED EMIT-BASED MODEL FOR THE  
CHARACTERISTIC OF COMPLEX SPACE

Scatterer Distribution (Total Number)	FDTD CPU Time (s)	EIT-Based CPU Time (s)	RMS Error
1*1 (1)	2.38	0.0104	0.86%
1*5 (5)	2.70	0.0449	0.52%
4*1 (4)	3.19	0.0268	0.18%
4*5 (20)	5.62	0.514	1.28%
10*15 (150)	8.74	6.53	0.07%
Radom (90)	27.6	4.72	1.02%

case, an efficient way to leverage the EMIT-based method is to build a common clustering model database. Compared with pilot-based methods, it also has good efficiency under the condition of a complete database. For example, a vehicle-to-vehicle channel is equivalent to a scattering cluster in the internet-of-vehicles channel modeling [16].

### C. Mode Analysis Step of the EMIT-Based Model

After efficient characterization of the complex space is verified, the EMIT-based model performs a mode analysis of the above characterization results to obtain theoretical interpretations to guide the design of wireless communications.

Consider an actual information transmission scenario where the RF front-end is a single-polarized dipole antenna array operating at 2.5 GHz (equivalent using an ideal line source operating at 2.5 GHz), with a scale of 10\*1 (designed to make the MIMO feature more obvious), and the complex space is simplified to a 4\*5 metallic cylindrical scatterer cluster. According to the quick algorithm in Section III-A, the electric field distribution on the validation plane is shown in Fig. 8. By substituting the solved coupling operator  $\bar{\mathbf{G}}_{EIT}$  into (6), the EM effective capacity  $C_{eff}$  of this model in wireless communication is known as 5.2, which means that the actual effective number of available channels is five. However, Fig. 3 shows that dyadic Green's function operators  $\bar{\mathbf{G}}_{TR}$  (coupling operators in free space) will bring channel gains far beyond 5.2. Therefore, the EMIT-based model provides a convenient tool to quantitatively explain the influence of the complex environment on wireless communication quality. The information at the receiving end in Fig. 8 can help us get the operator  $\bar{\mathbf{G}}_{EIT}$ . Through the SVD mentioned above, ten modes at the transmitting end can be decomposed, and the EM responses of these ten modes in the complex space are shown in Fig. 9. It is clearly found that the first five modes successfully send signals to the receiver effectively in different coupling paths. However, the coupling paths of higher-order modes bypass the receiver's acceptance range and become unavailable modes in wireless communication.

To define the concepts of “available” and “unavailable” more clearly, we show the distribution of modes’ singular values for the different number of channels in Fig. 10. A formal definition is given as follows: if all modes are numbered

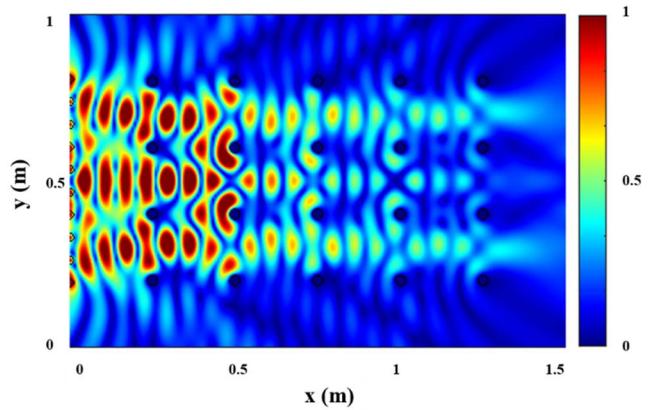


Fig. 8. Total normalized electric field distribution on the validation plane corresponding to the proposed complex space obtained by EMIT-based model.

according to the normalized singular value in descending order like Fig. 10; then the available modes are defined as those whose index is less than the EM effective capacity  $C_{eff}$ ; and the other modes are defined as the unavailable modes. Since the power resources in an actual wireless communication system are limited, the mode weight corresponding to each channel number is normalized here. Obviously, for an MIMO system, there will be an evident truncation of the modes’ singular value distribution, and modes below the truncation usually become “unavailable.” The number of “available” modes will be strictly determined by (8) after the coupling operator  $\bar{\mathbf{G}}_{EIT}$  is obtained by the EMIT-based model.

Obviously, the more channels available, the more information that can be transmitted, and the greater the channel capacity of the corresponding EM space. However, communication resources of the RF front-end are often limited, so it is important to allocate resources properly to achieve better information transmission efficiency. In the next section, power distribution is taken as the background problem to discuss the guidance significance of the EMIT-based model for real wireless communication in a complex environment.

## IV. EXPERIMENTAL ANALYSIS

It is worth noting that the above discussion on the application of the EMIT-based model is carried out by simulation. In order to fully explain the effectiveness of the EMIT-based model and the application method under the background of wireless communication, we carried out experimental exploration with the aid of a 3\*1 MIMO system.

The experiment construction is shown in Fig. 11, where the system is surrounded by the absorption boundary covered with absorbing materials, and cylindrical metal scatterers with a height of 0.25 m and a radius of 0.015 m are uniformly distributed in the EM space with 4\*5 arrays. The transmitting sources and the receiving fields were replaced by dipole antennas with a center frequency of 2.5 GHz and a gain of 2 dBi. The vector network analyzer (VNA) is used to measure the  $S_{21}$  between transmitting and receiving dipoles through the coaxial feed. Since the measurement of channel matrix elements is concerned with the single excitation properties of MIMO, we replace the actual MIMO system by changing the spatial position of the antenna in the transmitting aperture (shown as

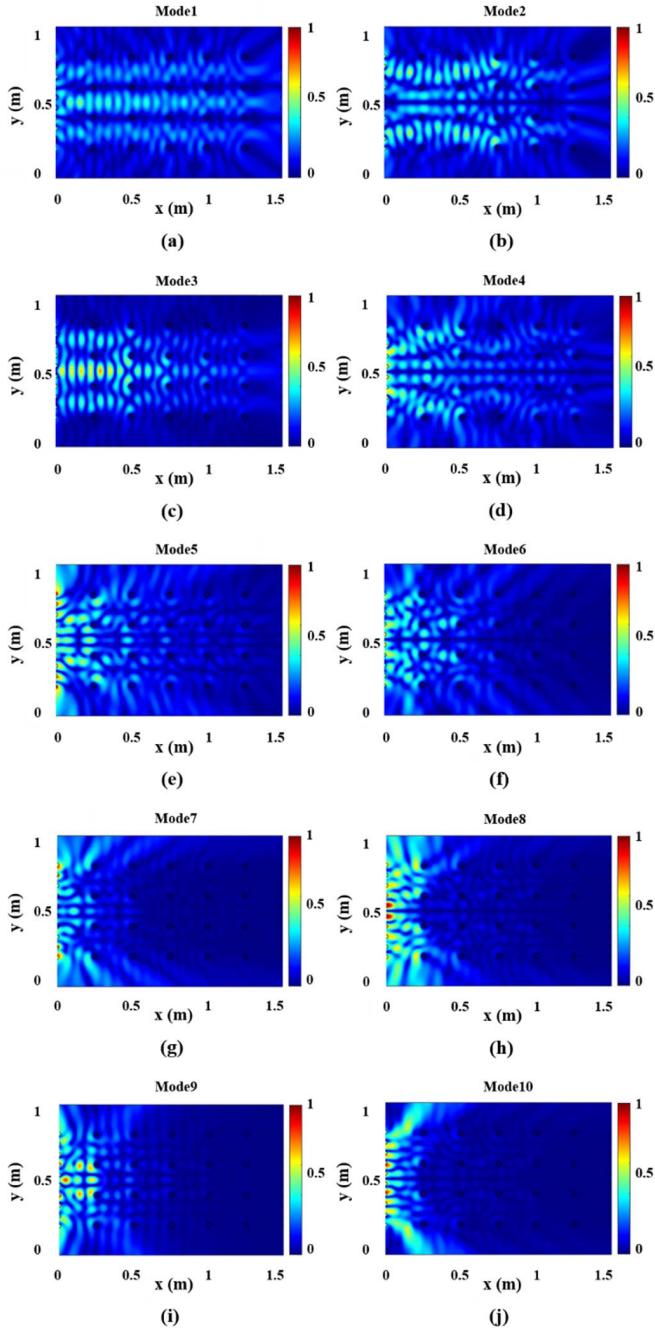


Fig. 9. Each mode's normalized electric field diagram on the validation plane obtained by EMIT-based model. (a)–(e) Available information transfer modes. (d)–(j) Unavailable higher-order information transfer modes.

the dotted white lines). Therefore, in the experimental design, we selected the weak coupling scenario with the antenna spacing as half-wavelength, and then measured the antenna excitation separately to avoid the impact of coupling on the verification of our EMIT-based model.

For the same scene, we performed effective characterization with the EMIT-based model, and the characterization results are shown in Fig. 12. Fig. 12(a) and (b) respectively represent the amplitude and phase comparison results between the simulation results of EMIT-based model and the experimental measurement values. The experimental results fully demonstrate the effectiveness of the EMIT-based model in complex

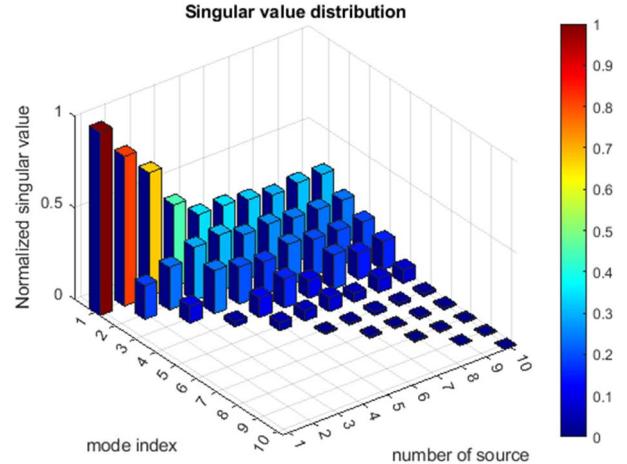


Fig. 10. Distribution of normalized singular values of different number of sources.

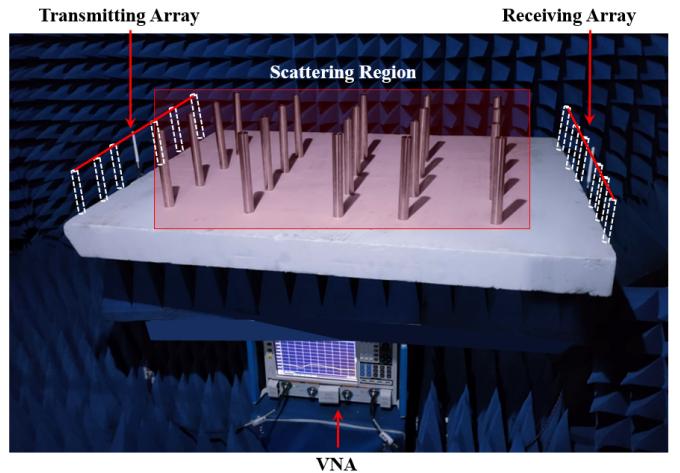


Fig. 11. Simple MIMO propagation system in complex space.

space characterization. The purpose of conducting  $7 \times 7$  channel measurement in our experiment is to verify the EMIT-based model more convincingly. However, the following is mainly to illustrate how the EMIT-based model guides the RF front-end signal transmission. Therefore, to simplify the demonstration process, we select three groups of data evenly spaced to form a new  $3 \times 3$  MIMO system. In fact, the selection of  $3 \times 3$  channel positions is arbitrary. However, in this article, to make the mode orthogonality more significant and avoid the influence of mode crosstalk on the transmitting strategy, three positions with relatively small mode crosstalk are selected, and the crosstalk matrix  $\mathbf{CT}$  is as follows:

$$\mathbf{CT} = \begin{pmatrix} 1 & 0.1857 & 4.03 \times 10^{-15} \\ 0.1857 & 1 & 6.51 \times 10^{-15} \\ 4.03 \times 10^{-15} & 6.51 \times 10^{-15} & 1 \end{pmatrix}. \quad (22)$$

Consider the following problems in an actual wireless communication scenario:  $3 \times 3$  MIMO system needs to conduct data transmission in disorder EM space (with the standard deviation of noise  $\chi$ ), the maximum transmitted power of RF front-end is  $P_0$ . Under this constraint, since it is a very important subject to consider the optimal power distribution, which is related to whether the upper bound for capacity can be achieved, the coupling operator  $\overline{\mathbf{G}}_{EIT}$  obtained by the EMIT-based model

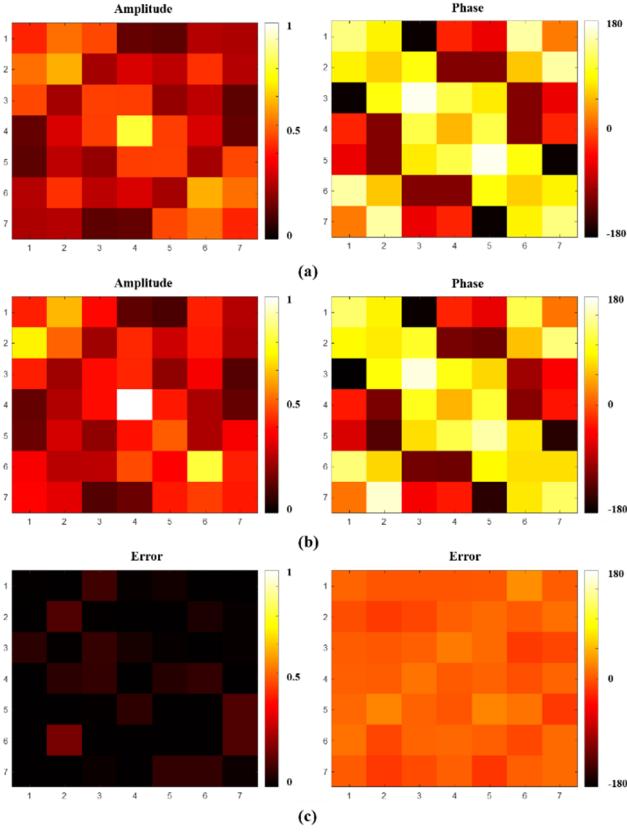


Fig. 12. Normalized amplitude and phase of  $S_{21}$  in  $7 \times 7$  MIMO system. (a) Simulation results in EMIT-based model. (b) Measurement results. (c) Error of the two above.

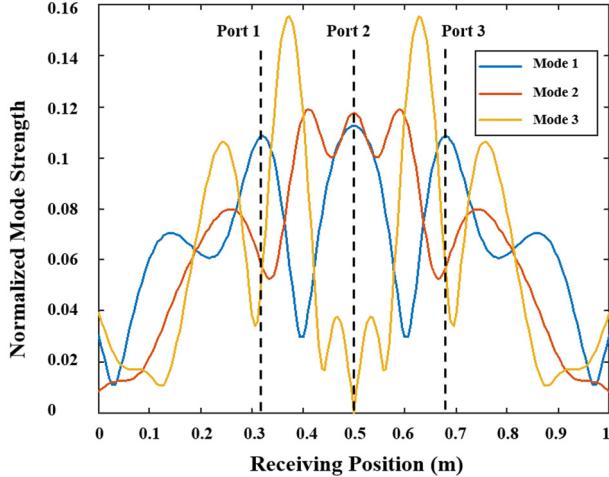


Fig. 13. Field distribution of three orthogonal modes at the receiver aperture. The locations of the three sources are indicated by black dotted lines on the diagram.

is used to endow the shape of the transmitted signal to obtain the best quality of information transmission.

Rewrite (5) and (6) to obtain a new coupling equation based on the scenario we are considering

$$\mathbf{U}_{EIT}^\dagger \mathbf{Y} = \mathbf{S} \mathbf{V}_{EIT}^\dagger \mathbf{X} \quad (23)$$

where  $\mathbf{U}_{EIT}$  and  $\mathbf{V}_{EIT}$  are determined by  $\overline{\mathbf{G}}_{EIT}$  to guide the signal processing of the transmitter and receiver, respectively.  $\mathbf{X}$  and  $\mathbf{Y}$  represent the signal form of the transmitter and receiver, respectively. The singular value matrix  $\mathbf{S}$  is

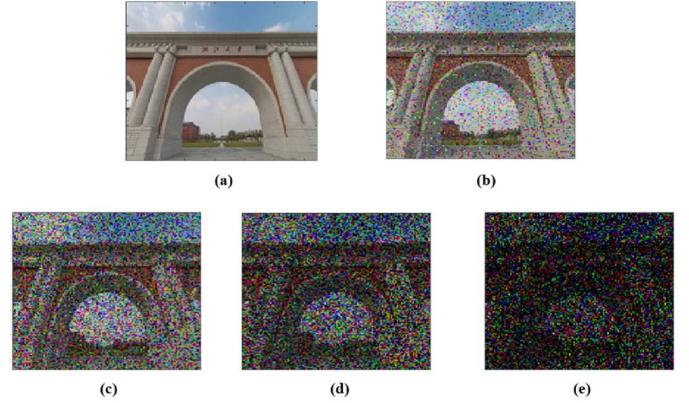


Fig. 14. Image transmission case based on EMIT-based model. (a) Original image with  $128 \times 128$  pixels. (b) Optimized power distribution. (c)–(e) Conducting single-mode transmission using modes 1, 2, and 3 respectively.

disassembled to obtain the received signal evaluation function  $f$

$$f = \sum_{m=1}^3 \langle \sigma_m V_m | \alpha \rangle. \quad (24)$$

In (24),  $\mathbf{X}$  is decomposed as a bitstream of information  $\mathbf{X}$  (here we use simple binary phase shift keying (BPSK) modulation) multiplied by the excitation coefficient  $|\alpha\rangle$ , and the unitary matrix  $\mathbf{V}_{EIT}$  is decomposed into three-mode vectors  $|V_m\rangle (m = 1, 2, 3)$ . Since there are three sources in our MIMO system, both  $|V_m\rangle$  and  $|\alpha\rangle$  here have three elements.  $\sigma_m (m = 1, 2, 3)$  are the diagonal elements of the singular value matrix  $\mathbf{S}$ , representing the influence of each mode on the receiving end. To have a clearer understanding of  $\sigma_m$ , we depicted the electric field distribution at the receiving end with the help of the EMIT-based model, as shown in Fig. 13. The calculated proportions of the three modes are 67.55%, 23.19%, and 9.25%, respectively. Mode 1 with the strongest proportion just contributes its crest to the receiving end, while mode 3 with the weakest proportion just contributes its trough to the receiving end. This provides a clear perspective for signal waveform design from the EM point of view, and reveals that EM space is not the only factor determining mode contribution, and EM space characteristics and RF front-end characteristics should be considered together.

According to the crosstalk matrix calculated in (22), we treat these three modes as orthogonal. Therefore,  $|V_m\rangle$  becomes a set of orthogonal basis in a Hilbert space, meeting  $\langle V_m^\dagger | V_n \rangle = 0$  and  $\langle V_m^\dagger | V_m \rangle = 1$ . Hence,  $|\alpha\rangle$  can be written as an orthogonal basis expansion

$$|\alpha\rangle = \sum_{n=1}^3 \lambda_n |V_n\rangle \quad (25)$$

where  $\lambda_n$  represent the corresponding weight of each basis vector, which determines the power distribution on the transmitting source. Therefore, the constraint of constant total power  $P_0$  can be equivalent to that the excitation vector is located on a fixed circle in the Hilbert space, and the received signal evaluation function  $f$  is the sum of the weighted

projections of the excitation vector on the three basis functions

$$\begin{cases} \text{find : } \lambda_m (m = 1, 2, 3) \\ \text{max : } f \\ \text{s.t. : } \sum \lambda_m = P_0. \end{cases} \quad (26)$$

The optimization problem can be easily solved by using Cauchy inequality. By substituting (8), the information transfer function can reach the maximum value only when  $\lambda_m/\sigma_m$  is a constant for different  $m$ . This is similar to the “water-filling” algorithm in channel estimation, while the core difference is that the key informatics parameters in this article are deduced by an effective EM algorithm. In addition,  $\bar{\mathbf{G}}_{TR}\bar{\mathbf{G}}_{TR}^\dagger$  or  $\bar{\mathbf{G}}_{EIT}\bar{\mathbf{G}}_{EIT}^\dagger$  have the same physical meaning as the transmit signal covariance matrix, and the main difference is that  $\bar{\mathbf{G}}_{TR}\bar{\mathbf{G}}_{TR}^\dagger$  or  $\bar{\mathbf{G}}_{EIT}\bar{\mathbf{G}}_{EIT}^\dagger$  is calculated by the EM methods based on dyadic Green’s function.

It is seen from the mode analysis based on the EMIT-based model that only under the guidance of a specific power allocation strategy, MIMO information transmission can achieve the effect of receiving power equal to transmitting power times path loss. To fully illustrate the guiding significance of the EMIT-based model for power distribution (essentially waveform design), Fig. 14 shows an image transmission case. BPSK is used to discretize every pixel in the picture into an 8-bit data stream for transmission, and noise  $\chi$  is joined to EM space. Obviously, under the premise of not processing channel noise, the RF front-end working strategy based on the EMIT-based model is much better than other transmission modes. Therefore, the EMIT-based model can not only efficiently represent the complex space but also make more valuable guidance for wireless communication.

## V. CONCLUSION

In this article, an EMIT-based model is presented to simulate the performance of MIMO systems in complex EM complex space effectively. First, the EM expression of the information coupling operator is given in the free space, and two key informatics parameters, EM effective capacity and path loss, are extracted from the EM perspective. It is proven that the MIMO antennas’ aperture is the critical factor in the EM effective capacity of the MIMO system. Second, the basic principle of the EM representation method in complex space is given, and several typical scenarios are analyzed, which proves the accuracy and efficiency of the EMIT-based model proposed. The results show that the EMIT-based model can reliably analyze the EM space about 10% of the time compared to the full-wave simulation. Finally, the MIMO performance in real propagation scenarios is calculated using the EMIT-based model. The experimental results verify that the channel matrix calculated is in good agreement with the measured ones. Based on this, it is pointed out how the EMIT-based model can effectively guide the MIMO design and feeding in a power distribution question.

The ultimate goal of the proposed EMIT-based model is to advance the development of EMIT and demonstrate a new idea of extracting information parameters to the antenna and propagation community using the basis of computational

electromagnetism. Currently, it is suitable for cluster models with arbitrary distribution, size, and material, providing an efficient and reliable method for guiding the power and phase allocation of antenna units in scattering complex space. The proposed model can also be easily extended to the guidance of MIMO antenna design in complex spaces by numerical discrete and optimization methods.

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