A General 3D Space-Time-Frequency Non-Stationary THz Channel Model for 6G Ultra-Massive MIMO Wireless Communication Systems

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Abstract—In this paper, a novel three-dimensional (3D) space-time-frequency (STF) non-stationary geometry-based stochastic model (GBSM) is proposed for the sixth generation (6G) terahertz (THz) wireless communication systems. The proposed THz channel model is very general having the capability to capture different channel characteristics in multiple THz application scenarios such as indoor scenarios, device-to-device (D2D) communications, ultra-massive multiple-input multiple-output (MIMO) communications, and long traveling paths of users. Also, the generality of the proposed channel model is demonstrated by the fact that it can easily be reduced to different simplified channel models to fit specific scenarios by properly adjusting model parameters. The proposed general channel model takes into consideration the non-stationarities in space, time, and frequency domains caused by ultra-massive MIMO, long traveling paths, and large bandwidths of THz communications, respectively. Statistical properties of the proposed general THz channel model are investigated. The accuracy and generality of the proposed channel model are verified by comparing the simulation results of the relative angle spread and root mean square (RMS) delay spread with corresponding channel measurements.

Index Terms—6G wireless communication systems, THz channel model, ultra-massive MIMO, long traveling path, space-time-frequency non-stationarity.

I. INTRODUCTION

The terahertz (THz) band (0.3–10 THz) is currently being explored for the sixth generation (6G) wireless communication systems [1]–[4]. THz band has huge bandwidth to provide ultra-high transmission data rate in numerous wireless applications. Some specific applications such as holographic video conferencing in the indoor room are supported by terabit wireless local area networks (T-WLAN). Data center [5] is a brand new scenario where thousands of devices are connected for cloud computing and storage. The interconnection of nanoscale machines in THz band strongly supports the Internet of nano things [6]. In addition, the THz band can provide ultra-high speed links for intercore communication in wireless on-chip networks [7] with nano on-chip antennas. In the THz communication, high transmission and reflection loss limit the transmission distance and massive multiple-input multiple-output (MIMO) technology is often utilized to compensate the transmission loss. In the aim of designing and evaluating THz communication systems more efficiently, an accurate and general channel model that can accurately capture THz propagation characteristics of different scenarios is essential. However, the existing THz channel models only focus on specific scenarios.

In THz communication systems, a significant challenge is the fact that the phenomena of scattering and diffraction are quite different from lower frequency bands. The THz propagation mechanisms were studied in [8]–[14]. THz measurements of multiple reflection effects on different materials were introduced in [8], [9]. Due to high reflection loss, high-order paths were very hard to detect resulting in the limited number of multipaths in THz band according to the measurement in [10]. The diffusely scattered propagation played an important role and was investigated in [11]–[14]. In [11], the detected signal powers in all directions for different materials were measured and simulated. Frequency-dependent scattering was also measured and simulated in [12], [13]. The wavelength of THz waves is in the same order with the roughness of some common materials. From these measurements and simulations, we found that The proportion of diffusely scattered rays gradually increased when it comes to higher frequency. In addition, each specular reflected path is surrounded with multiple diffusely scattered rays [14].

A number of channel models and measurements were investigated for indoor THz communications [15]–[21]. In [15], a multi-path ray tracing channel model for THz indoor communication was presented and validated with experiments. In


[16], a three-dimension (3D) time-variant THz ray tracing channel model was investigated for dynamic environments. In [17], a geometry-based stochastic THz indoor channel model considering the frequency dispersion was presented and verified by ray tracing. In [18], the authors investigated root mean square (RMS) delay spread and angular spread that were modeled by second order polynomial parameters for THz indoor communications. An indoor channel model based on ray tracing considering atmospheric attenuation was proposed in [19]. Antenna arrays were discussed for indoor THz communication systems in [20]. A wideband channel measurement for indoor THz wireless links was conducted between 260 GHz and 400 GHz [21]. Wireless data center networks in THz band were investigated in the literature [5], [22]–[25]. In addition, high speed transmission for THz wireless data center was promised with its large bandwidth and lower interference [26]–[31]. A stochastic channel model was proposed for THz data center and simulated in [32], [33].

In THz communication systems, ultra-massive MIMO technologies employing thousands of antennas are considered as one of the solutions to compensate the path loss and are expected to be utilized in 6G [3], [4]. Massive MIMO channel models and measurements were studied in [34]–[40]. In [34], a detailed survey of massive MIMO channel models and measurements were summarized. In [35]–[39], the effect of spherical wavefront and clusters evolution along the time and array axis are considered, showing the spatial non-stationarity in massive MIMO systems. Non-stationary massive MIMO channels by transformation of delay and angle of arrivals were studied in [40]. In [41]–[45], millimeter wave (mmWave) band channel measurements for massive MIMO channels in different scenarios were presented. A novel 3D geometry-based stochastic model (GBSM) based on homogeneous Poisson point process was proposed for mmWave channels [43]. A GBSM for mobile-to-mobile scenarios for mmWave bands was presented in [44].

In general, the existing massive MIMO channel models have only considered characteristics of sub-6 GHz and mmWave bands, and are not suitable for THz communication systems because propagation mechanisms in THz bands are quite different from lower frequencies. The ray tracing based deterministic channel models are so complex and not suitable for THz communication systems design. Different from deterministic channel models, the GBSMs are more flexible and widely used in the 5G standardized channel models [46]. However, the existing THz GBSMs do not support mobility, ultra-massive MIMO, and frequency non-stationarity simultaneously and only focus on one specific scenario. In order to design and evaluate 6G wireless communication systems more efficiently, in this paper, a general 3D THz GBSM considering space-time-frequency (STF) non-stationarity is proposed. The major contributions of this paper are listed as follows.

1) A general STF non-stationary THz channel model for 6G ultra-massive MIMO wireless communication systems is presented. The proposed model can support different scenarios whose transmission distances range from tens of meters to a few centimeters. The proposed THz channel is very general to different specific THz scenarios and applications with certain model parameters.

2) The STF non-stationarities in space, time, and frequency domains caused by ultra-massive MIMO, long traveling paths, and large bandwidths are considered. The evolution in STF domain based on birth-death process is presented.

3) The statistical properties such as STF correlation function (STFCF), the stationary interval, and power spectrum density (PSD) are derived. The simulation results show good agreements with the corresponding measurements, illustrating the validity and generality of the proposed THz simulation model.

The remainder of this paper is organized as follows. In Section II, the THz GBSM is described in detail. The channel impulse response (CIR) and the parameters are presented. The STF evolution based on birth-death process is introduced. In Section III, typical statistical properties of the proposed general THz channel model are derived. In Section IV, different statistical properties of the channel model are simulated and compared with measurements. Finally, conclusions are drawn in Section V.

II. A General 3D Non-Stationary THz Channel Model

The diagram of THz wireless communication system is shown in Fig. 1 where uniform planar arrays (UPAs) are employed at the transmitter (Tx) and receiver (Rx) containing $M_T$ and $M_R$ elements at Tx and Rx, respectively. There are $M_T^V$ columns and $M_T^H$ rows in the Tx array. Similarly, $M_R^V$ columns and $M_R^H$ rows in the Rx array so that we have $M_T = M_T^T \times M_T^H$ and $M_R = M_R^T \times M_R^H$. Note that non-uniform antenna arrays are also supported in the model with the knowledge of the geometric relationship of antenna elements. As shown in Fig. 1, considering multi-bounce propagation, the $n$th path is represented by one-to-one pair clusters, i.e., $C_n^T$ and $C_n^R$ at the Tx and Rx side, respectively. In each path, The initial distance between the first element the Tx and Rx is denoted as $D$. The first element is the used as the reference to calculate other elements. It is a benchmark element in the array. It should be noticed that $(x_G, y_G, z_G)$ axes are established as the global coordinate system (GCS) whose $x$ axis is oriented to the first element of the receive array. It should be distinguished from the local coordinate systems (LCS) when calculating 3D antenna pattern in the LCS. The inter-element spacing in a row and in a column of Tx (Rx) are denoted as $\delta^T_H$ and $\delta^T_V$, respectively.

The elevation and azimuth angels of the row (column) for Tx are denoted as $\beta^T_H$ and $\beta^T_V$. Let $\vec{A}_p^T$ and $\vec{A}_q^R$ denote the position vectors of $A_p^T$ and $A_q^R$ from the first element of Tx and Rx array, respectively. The $\vec{A}_p^T$ can be expressed as

$$\vec{A}_p^T = \begin{bmatrix} \rho_H \delta_H^T \cos \beta_H^T \cos \beta_H^T & \rho_H \delta_H^T \cos \beta_H^T \sin \beta_H^T & \rho_H \delta_H^T \sin \beta_H^T \\
+pel_\delta_V^T \cos \beta_V^T & pel_\delta_V^T \cos \beta_V^T & pel_\delta_V^T \sin \beta_V^T \end{bmatrix} \begin{bmatrix} \beta_H^T \\
\beta_V^T \end{bmatrix}$$

(1)
Fig. 1. A 3D THz GBSM for massive MIMO communication systems.

| Parameters       | Definitions                                                                 |
|------------------|-----------------------------------------------------------------------------|
| \( \delta_{h}^{(R)}, \delta_{v}^{(R)} \) | Vertical and horizontal inter-element spacing plane of Tx (Rx), respectively |
| \( D \)         | Distance between the center of Tx and Rx at initial time                    |
| \( C_{A}^{1}, C_{E}^{p} \) | The first- and last-bounce clusters of the \( n \)th path, respectively     |
| \( d_{p,q}^{(t)}(t) \) | Distance from \( A_{p}^{n} \) to \( A_{q}^{n} \) at time instant \( t \)      |
| \( d_{p,q}^{(t)}(t) \) | The total distance from \( A_{p}^{n} \) to \( A_{q}^{n} \) via the \( m \)th ray of the \( n \)th cluster at initial time  |
| \( \beta_{A,L}^{(t)}, \beta_{E,L}^{(t)} \) | Azimuth and elevation angles of the Tx (Rx) array, respectively             |
| \( \alpha_{A,L}^{(t)}, \beta_{E,L}^{(t)} \) | Azimuth and elevation angles of departure (AAoD and EAoD) of the LOS path transmitted from \( A_{p}^{n} \) at time instant \( t \) |
| \( \phi_{A,L}^{(t)}, \phi_{E,L}^{(t)} \) | Azimuth and elevation angles of arrival (AAoA and EAoA) of the LOS path impinging on \( A_{p}^{n} \) at time instant \( t \) |
| \( v_{x}, v_{y} \) | Velocity of Tx and Rx                                                        |

where \( p_{v} \) and \( p_{H} \) mean that the \( p \)th element is located in the \( p_{H} \)th row and \( p_{v} \)th column, we have that \( p = (p_{H} - 1) \times M_{V} + p_{v} \). For Rx array the vector \( A_{q}^{R} \) can be calculated in a similar way.

A. Channel Impulse Response

Considering small-scale fading, path loss, shadowing, molecular absorption, and blockage effect, the complete channel matrix is given by

\[
H = [PL \cdot SH \cdot BL \cdot MA]^{\frac{1}{2}} \cdot H_{a}
\]  

(2)

where \( PL \) denotes the path loss. The path loss \( PL \) of the channel is modeled by close-in free space reference distance path loss model \([48]–[50]\) as

\[
PL(d) = PL_{0}(m) + 10\gamma \log_{10}(\frac{d}{m})
\]

(3)

where \( PL_{0}(m) \) is the free space path loss and can be calculated by Friis equation, \( \gamma \) means the propagation coefficient and \( m \) is the reference distance. \( SH \) denotes the shadowing and is modeled as a lognormal random variable. The blockage loss \( BL \) caused by human activities is taken into account \([42]\). The molecular absorption loss \( MA \) for THz communications can be found in HITRAN database \([51]\). All these parameters are calculated in power level in this THz channel.

Due to the large bandwidth of THz communication systems, it is not suitable to set the whole channel parameters at center frequency. To describe the channel more accurately, the communication band \( B \) is divided into \( N_{F} \) small sub-bands with bandwidth of \( B_{\text{sub}} = B/N_{F} \). The center frequency of the \( i \)th sub-band is \( f_{i} \). The channel matrix of the \( i \)th sub-band is denoted as \( H_{f_{i}} \). Considering the nonnegligible difference in different frequency sub-bands, both large-scale parameters and small-scale fading parameters are calculated for different sub-bands separately. The large-scale fading parameters in (2), \( i.e., PL, SH, BL \), and \( MA \) are updated for different frequency sub-bands. The small-scale channel matrix of a sub-band is denoted as \( H_{s,f_{i}} \). In traditional communication systems,
the received signal can be calculated by the convolution of the transmitted signal and the channel impulse response in the time domain or their product in the frequency domain. However, in the communication systems with large bandwidth, the transmitted signal can be also divided into the sub-band signals. After each sub-band experiencing its corresponding channel matrix, the received signal can be calculated by summing up all the band-pass signals as

$$R(t, \tau) = \sum_{i=0}^{N_p} \int_{-\infty}^{\infty} S(t, f) \cdot A_{f_i}(f) \cdot H_{f_i}(f, f) \cdot w(t) + n(t)$$  \hspace{1cm} (4)

where $S(t, f)$ is the spectrum of the transmitted signal, and $R(t, \tau)$ is the received signal. The filtering function $A_{f_i}(f)$ denotes the ideal bandwidth band-pass filter for the sub-band. The transfer function matrix $H_{f_i}(f, f)$ is the Fourier transform of $h_{f_i}(t, f)$, $w(t)$ is the time-variant beamforming matrix, and $n(t)$ is the normalized complex additive white Gaussian noise. The channel impulse response of the $i$th sub-band $H_{f_i, t} = [h_{p,q,f_i}(t, \tau)]_{M^R \times M^R}$ where $M^R$ and $M^T$ are the element numbers of Tx and Rx, respectively. $h_{p,q,f_i}(t, \tau)$ is the CIR from $A_t^p$ to $A_q^R$ and expressed as the summation of the line-of-sight (LOS) and non-LOS (NLOS) components, i.e.

$$h_{p,q,f_i}(t, \tau) = \sqrt{\frac{K^R}{K^R + 1}} h_{p,q,f_i}^L(t, \tau) + \frac{1}{K^R + 1} \sum_{n=1}^{N_{p,q}(t)} \sum_{m=1}^{M_{n,f_i}} h_{p,q,f_i}^N(t, \tau)$$  \hspace{1cm} (5)

where $K^R$ is the K-factor, $N_{p,q}(t)$ is the number of clusters from $A_t^p$ to $A_q^R$ at time instant $t$, and $M_{n,f_i}$ represents the time-space variant number of rays in the $n$th cluster of the $i$th sub-band. It is frequency dependent and assumed to follow a Poisson distribution Pois($\lambda$) [45] where $\lambda$ is the mean and variance of $M_{n,f_i}$.

For the LOS components in (5), the complex channel gain can be expressed as

$$h_{p,q,f_i}^L(t, \tau) = \begin{bmatrix} F_{q,V}(\phi_{E,L}(t), \phi_{A,L}(t)) \\ F_{q,H}(\phi_{E,L}(t), \phi_{A,L}(t)) \end{bmatrix}^T e^{j\theta_{LOS}} \begin{bmatrix} 0 \\ -e^{j\theta_{LOS}} \end{bmatrix}$$

$$= \begin{bmatrix} F_{p,V}(\phi_{E,L}(t), \phi_{A,L}(t)) \\ F_{p,H}(\phi_{E,L}(t), \phi_{A,L}(t)) \end{bmatrix} e^{j2\pi v_{p,q,i}^{LOS}(t)} \delta(\tau - \tau_{p,q}(t))$$  \hspace{1cm} (6)

where $\{\cdot\}^T$ denotes transposition operation, $\theta_{LOS}$ is uniformly distributed within $(0, 2\pi)$, $F_{p,q,V}$ and $F_{p,q,H}$ are the antenna patterns of $A_t^p$ and $A_q^R$ at vertical and horizontal planes, respectively. Note that in this paper, $\phi$ represents angles in the LCS and needs to be transformed from angles in the GCS. The angle in LCS is obtained by rotating from the corresponding angles in the GCS for three axes [46]. Similar to chapter 7.1 in [46], rotation of $x_G / y_G / z_G$ is denoted by $\gamma_x^T / \gamma_y^T / \gamma_z^T$, respectively. For example, $\phi_{A,L}^T$ and $\phi_{E,L}^T$ can be obtained by (7) and (8), respectively.

Then the antenna patterns in (6) can be calculated as

$$F_{q,V}(\phi_{E}, \phi_A) = G(\phi_{E}, \phi_A) \sin(\phi_{E})$$  \hspace{1cm} (9)

$$F_{q,H}(\phi_{E}, \phi_A) = G(\phi_{E}, \phi_A) \cos(\phi_{E})$$  \hspace{1cm} (10)

In the proposed channel model, antenna arrays at Tx and Rx sides can be any radiation pattern for each element. The radiation pattern can be substituted into (6) to generate the CIR of the LOS path. In this paper, we assume that the dipole antennas are utilized at both Tx and Rx sides. In this case [47], we have

$$G(\phi_{E}, \phi_A) = \sqrt{1.64} \frac{\cos(\frac{\phi_{E}}{2}) \cos(\frac{\phi_A}{2})}{\sin \phi_A}$$  \hspace{1cm} (11)

The delay $\tau_{p,q}(t)$ is calculated by $\tau_{p,q}^L(t) = \frac{d_{p,q}^L(t)}{c}$, where $\tau_{p,q}^L(t)$ denotes the distance from $A_t^p$ to $A_q^R$ at time instant $t$. The vector $\bar{d}_{p,q}^L(t)$ can be calculated as

$$\bar{d}_{p,q}^L(t) = \bar{D} + \bar{A}_P - \bar{A}_q + \int_0^T (\bar{v}^R(t) - \bar{v}^T(t)) dt$$  \hspace{1cm} (12)

In this model, we assume that the velocity vector of Tx or Rx is constant and can be expressed as

$$\bar{v}^T = v^T \cdot \begin{bmatrix} \cos \alpha_Z^T & \cos \alpha_A^T & \cos \alpha_E^T & \sin \alpha_A^T & \sin \alpha_E^T \end{bmatrix}$$

$$\bar{v}^R = v^R \cdot \begin{bmatrix} \cos \alpha_Z^R & \cos \alpha_A^R & \cos \alpha_E^R & \sin \alpha_A^R & \sin \alpha_E^R \end{bmatrix}$$  \hspace{1cm} (13, 14)

The Doppler frequency $\nu_{p,q,f_i}^L(t)$ between $A_t^p$ and $A_q^R$ is expressed as

$$\nu_{p,q,f_i}^L(t) = \frac{1}{\lambda(f_i)} \left\langle \bar{d}_{p,q}^L(t), \bar{v}^R - \bar{v}^T(t) \right\rangle$$  \hspace{1cm} (15)

where $\langle \cdot \rangle$ is the inner product operator and $\| \cdot \|$ calculates the Frobenious norm. The $\lambda(f_i)$ is the wavelength of the $i$th sub-band and calculated as

$$\lambda(f_i) = \frac{c}{T_i}$$  \hspace{1cm} (16)

$$\phi_{A,L}^T = \arccos \left( \cos(\gamma_x^T) \cos(\gamma_z^T) \cos(\phi_{A,L}^T) + \sin(\gamma_y^T) \cos(\phi_{E,L}^T - \gamma_z^T) - \sin(\gamma_z^T) \sin(\phi_{E,L}^T - \gamma_z^T) \right) \sin(\phi_{A,L}^T)$$

$$\phi_{E,L}^T = \text{arg} \left( \begin{bmatrix} \cos(\gamma_y^T) \sin(\phi_{A,L}^T) \cos(\phi_{E,L}^T - \gamma_z^T) + \\ \sin(\gamma_y^T) \sin(\phi_{A,L}^T) \end{bmatrix} \begin{bmatrix} \sin(\gamma_y^T) \cos(\phi_{A,L}^T) + \\ \sin(\gamma_y^T) \sin(\phi_{A,L}^T) \end{bmatrix} \right)$$  \hspace{1cm} (7, 8)
For NLOS components, the channel gain $h_{p,q,f,m,n}(t)$ can be written as

$$h_{p,q,f,m,n}(t,\tau) = \begin{bmatrix} F_{p,V}(\theta_{E,m,n}^{R}(t),\theta_{A,m,n}^{R}(t)) \\ F_{p,H}(\theta_{E,m,n}^{R}(t),\theta_{A,m,n}^{R}(t)) \end{bmatrix}^{T} \begin{bmatrix} e^{j\theta_{V}^{m,n}} \\ e^{j\theta_{H}^{m,n}} \end{bmatrix} \sqrt{\frac{k_{m,n}e^{j\phi_{m}^{V,n}}}{\kappa_{m,n}+e^{j\phi_{m}^{H,n}}}} \begin{bmatrix} F_{p,V}(\theta_{E,m,n}^{T}(t),\theta_{A,m,n}^{T}(t)) \\ F_{p,H}(\theta_{E,m,n}^{T}(t),\theta_{A,m,n}^{T}(t)) \end{bmatrix} \sqrt{P_{p,q,f,m,n}(t)e^{2\pi j\nu_{p,q,m,n}(t)}e^{2\pi j\nu_{p,q,m,n}(t)}d(\delta-\tau_{p,q,m,n}(t))} \tag{17}$$

where $k_{m,n}$ stands for the cross polarization power ratio, $\theta_{V}^{m,n}$, $\theta_{H}^{m,n}$, and $\phi_{m}^{V,n}$, $\phi_{m}^{H,n}$ are initial phases with uniform distribution over $(0, 2\pi)$. $P_{p,q,f,m,n}$ is the powers of the $n$th ray in the $n$th cluster between $A_{p}^{T}$ and $A_{q}^{R}$ at time instant $t$. The delay of the $n$th in the $n$th cluster is denoted as $\tau_{p,q,m,n}(t)$ and calculated as $\tau_{p,q,m,n}(t) = d_{p,q,m,n}(t)/c$. The Doppler frequencies at Tx and Rx side are denoted as $\nu_{p,q,m,n}(t)$ and $\nu_{p,q,m,n}(t)\phi_{m}^{R,n}(t)$, respectively. The total number of paths in each cluster can be set as a typical number such as 50 or 100 in the channel simulation.

In order to generate the complete THz channel coefficient, we need firstly to generate a set of clusters at initial time and first element of Tx/Rx in the first band. Then the STF cluster evolution will be taken in the space-time-frequency domains and parameters will also be updated. In the same time, the new clusters have the probability of birth process.

For a new cluster, spherical wavefront needs to be considered due to the utilization of large antenna arrays. The time variant transmission distances caused by long traveling movement are also nonnegligible. In THz communication systems, the specular reflection accounts the main part in the received power and diffuse scattering is only detectable around the specular reflection points [10]. For a specular reflection path, the position of reflection point will move when the Tx or Rx moves. Different from other GBSSMs in [36] where the total distance is divided into three parts: from the Tx to the first cluster, form the Rx to the last cluster, and the virtual link between these two clusters. In this channel model, we calculate the total distance of the path directly, the total distance from the Tx to the Rx is considered as an integrated random variable. The total distance of cluster is described recursively with an interarrival distance relative to the previous cluster

$$d_{n} = d_{n-1} + \Delta d_{n} \tag{18}$$

where $\Delta d_{n}$ is an negative exponential distribution (NEXP) random variable [17] with parameter $\delta_{N}$. For the first cluster, $\Delta d_{1}$ means the distance difference relative to the LoS path. For the first cluster, $\Delta d_{1}$ means the distance difference relative to the direct distance between the Tx and Rx, which is the LOS path in the LOS scenario and is still valid in the NLOS scenario.

The angular parameters of each ray consist of two parts: the angle of the cluster center and the relative angle. The relative angles $\Delta \theta_{A,m,n}^{T}$, $\Delta \theta_{E,m,n}^{T}$, $\Delta \phi_{A,m,n}^{R}$, and $\Delta \phi_{R,m,n}^{R}$ calculate the angular offset compared to the center of clusters. The angles can be calculated as

$$[\Delta \theta_{A,m,n}^{T}, \Delta \theta_{E,m,n}^{T}, \Delta \phi_{A,m,n}^{R}, \Delta \phi_{R,m,n}^{R}] = [\phi_{A,m,n}^{T}, \phi_{E,m,n}^{T}, \phi_{A,m,n}^{R}, \phi_{R,m,n}^{R}] + [\Delta \phi_{A,m,n}^{T}, \Delta \phi_{E,m,n}^{T}, \Delta \phi_{A,m,n}^{R}, \Delta \phi_{R,m,n}^{R}], \tag{19}$$

The angular parameters $\phi_{A,m,n}^{T}, \phi_{E,m,n}^{T}$ of cluster are assumed to obey wrapped Gaussian distributions. Angles of arrival (AoAs) of the $n$th cluster are generated as

$$\phi_{A,m,n}^{T} = \text{std}(\phi_{A,m,n}^{T}) \nu_{A,m,n}^{T} + \psi_{A,m,n}^{T} \tag{20}$$

$$\phi_{E,m,n}^{T} = \text{std}(\phi_{E,m,n}^{T}) \nu_{E,m,n}^{T} + \psi_{E,m,n}^{T} \tag{21}$$

where $\nu_{E,m,n}^{T}$ and $\nu_{A,m,n}^{T}$ are the Gaussian distribution $N(0,1)$, and $\phi_{A,m,n}^{T}$ and $\phi_{E,m,n}^{T}$ are standard deviations of AoAs and need to be estimated.

In this model, we assume that all the diffusely scattered rays happen around the specular reflected path according to [14]. When modeling the relative angle, we assume that the center of the cluster is the specular point. All the scatterers around the specular point is Gaussian distributed. The standard variances of the relative angle $\Delta \phi_{A,m,n}^{T}$, $\Delta \phi_{E,m,n}^{T}$, $\Delta \phi_{A,m,n}^{R}$, and $\Delta \phi_{R,m,n}^{R}$ are denoted as $\sigma_{A,m,n}^{T}$, $\sigma_{E,m,n}^{T}$, $\sigma_{A,m,n}^{R}$, and $\sigma_{E,m,n}^{R}$ respectively.

The total distance of rays in each cluster $d_{p,q,m,n}$ can be calculated from the relative angles and distances of specular reflection $d_{p,q,m,n}$ by the geometric relationship.

According to the relative angle, the total distance of different scattering paths can be calculated

$$d_{p,q,m,n} = \sqrt{d_{p,q,m,n}^{N} + d_{p,q,m,n}^{H}} \tag{22}$$

where $d_{p,q,m,n}^{N}$ and $d_{p,q,m,n}^{H}$ represent the vertical and horizontal distances of the path length, respectively. They can be calculated by

$$d_{p,q,m,n}^{N} = d_{p,q,n} \sin(\phi_{A,m,n}^{R}) r_{c}^{R} / \cos(\Delta \phi_{E,m,n}^{T}) + d_{p,q,n} \sin(\phi_{A,m,n}^{T}) r_{c}^{T} / \cos(\Delta \phi_{E,m,n}^{T}) \tag{23}$$

$$d_{p,q,m,n}^{H} = d_{p,q,n} \cos(\phi_{A,m,n}^{R}) r_{c}^{R} / \cos(\Delta \phi_{A,m,n}^{R}) + d_{p,q,n} \cos(\phi_{A,m,n}^{T}) r_{c}^{T} / \cos(\Delta \phi_{A,m,n}^{T}) \tag{24}$$

where $r_{c}^{T,R}$ is the ratio of the distance between the cluster and the Tx(Rx) to the total distance. When the relative angles are zero, the path can be considered as specular reflection. Then, the relative distance can be obtained by $d_{p,q,m,n}$ and the specular path. For single bounce clusters, it is clear that $r_{c}^{T} + r_{c}^{R} = 1$. For multiple bounce rays, we have $r_{c}^{T} + r_{c}^{R} < 1$. The simulated probability density functions (PDF) for different variances are shown in Fig. 2.

**B. STF Evolution**

In this sub-section, we will generate the STF-variant parameters for the channel. The evolution in time axis and array axis is caused by the offset of geometry relationship. The evolution in the frequency domain is caused by frequency-dependent scattering.

birth-death process has been validated by channel measurements in vehicle-to-vehicle scenarios [35]. In massive MIMO
communication systems, similar properties can be observed on the array axis. In this paper, space-time cluster evolutions are modeled in a uniform manner. The flowchart of the STF cluster evolution based on the birth-death process is shown in Fig. 3. The generation (birth) and recombination (death) rates of clusters are denoted as \(\lambda_G\) and \(\lambda_R\), respectively. Firstly, initial parameters for clusters are generated at time \(t\) and first element of Tx and Rx. Then, the unified array-time cluster evolution is conducted and parameters at \(t + \Delta t\) and other elements are updated according to the geometric relationship.

For the Tx side, the probability of a cluster remains over the time interval \(\Delta t\) and antenna element spacing \(\delta_p\) can be calculated as

\[
P_{\text{remain}}^T(\Delta t, \delta_p) = \exp\left(-\lambda_R \left[ (\overline{\delta}_1) + (\overline{\epsilon}_2)^2 - 2\overline{\epsilon}_1\overline{\epsilon}_2 \cos \left(\alpha_A - \beta_A^T\right) \right]^2 \right)
\]

(25)

where

\[
\overline{\delta}_1 = \delta_p \cos \left(\beta_A^T\right) / D_c^A
\]

(26)

\[
\overline{\epsilon}_2 = v^T \Delta t / D_c^S.
\]

(27)

Different from the evolution along the uniform linear array (ULA) in [35], [36] the evolution along the UPA is not in one fixed direction. For the elements in the first row and first column, the \(\overline{\delta}_1 = \delta_p\), the \(\overline{\beta}_1^T\) equals to \(\beta_{H,E}^T\) and \(\beta_{V,E}^T\), respectively. For other elements in the array, we have \(\overline{\delta}_1 = \sqrt{2}\delta_p\) and \(\overline{\beta}_1^T = (\beta_{H,E}^T + \beta_{V,E}^T)/2\). Symbols \(D_c^A\) and \(D_c^S\) are scenario-dependent correlation factors in the array and time domains, respectively.

For the Rx side, the probability of a cluster existing over \(\Delta t\) and element spacing \(\delta_q\) is calculated as

\[
P_{\text{remain}}(\Delta t, \delta_p, \delta_q) = P_{\text{remain}}^T(\Delta t, \delta_p) \cdot P_{\text{remain}}^R(\Delta t, \delta_q).
\]

(28)

The mean number of newly generated clusters is obtained by

\[
\mathbb{E}\{N_{\text{new}}\} = \frac{\lambda_G}{\lambda_R} \left[ 1 - P_{\text{remain}}(\Delta t, \overline{\delta}_p, \overline{\delta}_q) \right].
\]

(29)

Since the center of all of the clusters are specular reflection, there exists a mirror point for \(A_{T,p}^T\) and \(A_{R}^T\). As we can see in the Fig. 4, the mirror point of \(A_{T,p}^T\) is symmetrical about the first reflection surface of the \(n\)th path at time instant \(t\) and denoted as \(A_{p,q,n}^T(t)\). Similarly, the mirror point of \(A_{R}^T\) is symmetric about the first reflection surface of the \(n\)th path at time instant \(t\) and denoted as \(A_{p,q,n}^R(t)\).

According to the geometrical relationships in Fig. 1, the distance vectors of the \(m\)th ray of the \(n\)th cluster \(d_{p,q,m,n}^T(t)\) and \(d_{p,q,m,n}^R(t)\) at the Tx side and Rx side are calculated as

\[
d_{p,q,m,n}^T(t) = d_{p,q,m,n}^T(t) \begin{bmatrix} \cos \phi_{E,m,n}^T \cos \phi_{A,m,n}^T \\ \cos \phi_{E,m,n}^T \sin \phi_{A,m,n}^T \\ \sin \phi_{E,m,n}^T \end{bmatrix}
\]

(30)
It is clear that the total distance of the path equals to the distance from the mirror point to the antenna in another side. We have that

$$\tilde{d}_{p,q,m}^T(t) = \|\tilde{d}_{p,q,m}^T(t)\| = \|\tilde{d}_{p,q,m}^R(t)\| = \left|\begin{array}{c}
\cos \phi_{E,m}^R \\
\cos \phi_{E,m}^R \sin \phi_{A,m}^R \\
\sin \phi_{E,m}^R
\end{array}\right|.$$  \hspace{1cm} (31)

For the array time evolution, the Tx and Rx are calculated separately. In the first step, we assume that Rx moving along time-array axis and Tx is static. We have that

$$\tilde{d}_{p,q,m}^R(t + \Delta t) = \left|\begin{array}{c}
\tilde{d}_{p,q,m}^R(t) \\
\tilde{d}_{p,q,m}^R(t) + \tilde{A}_R^q + \tilde{v}_R \cdot \Delta t
\end{array}\right|.$$  \hspace{1cm} (33)

where \(\tilde{d}_{p,q,m}^R(t + \Delta t)\) is a temporary distance after the first step. The \(\tilde{d}_{p,q,m}^R(t + \Delta t)\) is the vector from the mirror point of Tx to the \(A^R_q\). Then the Tx is moving and Rx keep static.
The new distance can be calculated as
\[ d_{p,q,m_n}(t + \Delta t) = \| \vec{d}_{p,q,m_n}(t + \Delta t) \| \]
\[ = \| \vec{d}_{p,q,m_n}^T(t + \Delta t) - \vec{A}_p \cdot \Delta t \| \] (34)
where \( \vec{d}_{p,q,m_n}(t + \Delta t) \) denotes the vector from the mirror point of Rx to the \( A_p^T \). The angle can also be calculated from the updated vector. For both single-bounce paths and multi-bounce paths, we only consider the total path lengths which contain the virtual link in multi-bounce paths.

The Doppler frequency \( \nu_{p,q,m_n}^T(t) \) at Tx and \( \nu_{p,q,m_n}^R(t) \) are calculated as
\[ \nu_{p,q,m_n}^T(t) = \frac{1}{\lambda(f_i)} \left\langle \frac{\vec{d}_{p,q,m_n}(t), \vec{v}_T}{\| \vec{d}_{p,q,m_n}(t) \|} \right\rangle \] (35)
\[ \nu_{p,q,m_n}^R(t) = \frac{1}{\lambda(f_i)} \left\langle \frac{\vec{d}_{p,q,m_n}(t), \vec{v}_R}{\| \vec{d}_{p,q,m_n}(t) \|} \right\rangle . \] (36)

For different sub-bands, positions of cluster and Tx/Rx remain unchanged so that the birth-death process of clusters are not considered. However, considering the scattering in different sub-bands, the birth-death process for rays in each clusters are considered along the frequency axis. The probability of a ray remains over two adjacent sub-channel is calculated as
\[ P_{\text{remain}}(B_{\text{sub}}) = \exp \left( -\frac{\lambda_R \cdot \rho_s \cdot B_{\text{sub}}}{B_c} \right) \] (37)
where \( \rho_s \) is a coefficient related to the reflection surface. Symbol \( B_c^T \) is scenario-dependent correlation factors in the frequency domain. Then the mean number of newly generated rays is obtained by
\[ E\{N_{\text{new}}\} = \frac{\lambda_G}{\lambda_R} [1 - P_{\text{remain}}(B_{\text{sub}})] . \] (38)

For the survived rays, the space-time-frequency varying ray power between \( A_p^T \) and \( A_q^R \) is expressed as
\[ P'_{p,q,f,m_n}(t) = \exp \left( -\tau_{p,q,m_n}(t) \frac{r - 1}{r_D S} \right) ^{10 \frac{Z_n}{10}} \cdot \xi_n(p,q) \cdot f_i \gamma_m \] (39)
where \( Z_n \) is the per cluster shadowing term in dB, \( DS \) is the RMS delay spread, \( r_D \) denotes the delay distribution proportionality factor and determined as the ratio of the standard deviation of the delays to the RMS delay spread \( [46] \). \( \gamma_m \) is an environment-dependent random variable and \( f_i \) is the center frequency of the communication system. The smooth power variations over the transmit and receive arrays can be simulated by a two-dimension (2D) spatial lognormal process \( \xi_n(p,q) \) and can be calculated as
\[ \xi_n(p,q) = 10^{\mu_n(p,q) + \sigma_n \cdot s_n(p,q)} \] (40)
along the large arrays, respectively. The 2D Gaussian process \( s_n(p,q) \) is given by
\[ s_n(p,q) = \sum_{k=1}^{K} c_k \cos(\theta - \theta_k + f_q, k + \theta_k) \] (41)
where \( K \) is the number of sinusoids, \( c_k \) is the amplitudes, \( f_q, k \) and \( f_k; k \) are the spatial frequencies on both ends. Phases \( \theta_k \) are uniformly distributed in \((0, 2\pi)\). The final ray powers are obtained by normalizing \( P_{p,q,f,m_n}(t) \)
\[ P_{p,q,f,m_n}(t) = \frac{P'_{p,q,f,m_n}(t)}{\sum_{n=1}^{N_{p,q}} \sum_{m=1}^{M_{p,q}} P'_{p,q,f,m_n}(t)} . \] (42)

In the scenarios with ultra-wideband MIMO, if the channel evolution is taken element by element, the amount of computation will be too large. In order reduce the computational costs without the loss of accuracy, the large array can be divided into sub-arrays. In each sub-array, the differences of total distance for all the elements can be neglected. As a result, the unified array time evolution can be updated only in adjacent sub-arrays. To determine the size of the sub-array, we calculate the Rayleigh distance \([54]\) of the sub-array and make sure that the Rayleigh distance is larger than the minimum path distance. The Rayleigh distance of the sub-array is calculated as \( R = 2D_t^2/\lambda \) where \( D_t \) is the maximum aperture of the sub-array. For example, for a typical indoor THz communication system working at 300 GHz with 1024 \times 1024 transmission array, the minimum transmission distance of paths is assumed as 4 m, then the maximum aperture of the sub-array is about 30 mm which supports maximum 60 \times 60 sub-array.

III. STATISTICAL PROPERTIES OF THE CHANNEL MODEL
A. STF-Variant Transfer Function
The time-variant transfer function \( H_{p,q,f}(t,f) \) can be calculated as the Fourier transform of the CIR with respect to delay
\[ H_{p,q,f}(t,f) = \int_{-\infty}^{\infty} h_{p,q,f}(t,f) e^{-2\pi f t} dt = \sqrt{\frac{K^R}{K^R + 1}} \rho_{p,q,f}(f) e^{-2\pi f t_{p,q,f}} + \frac{1}{K^R + 1} \sum_{n=1}^{N_{p,q}} \sum_{m=1}^{M_{p,q}} h_{p,q,f,m_n}(t) e^{-2\pi f t_{p,q,f,m_n}} . \] (43)

B. STF Correlation Function
The STF correlation function (STFCF) measures the correlation properties between two adjacent channel coefficients \( \rho_{p,q,p',q'}(t,f,\Delta t, \Delta f) \) can be calculated as
\[ \rho_{p,q,p',q'}(t,f,\Delta t, \Delta f) = \left[ H_{p,q,f}(t,f) H^*_{p',q,f}(t) e^{+\Delta f, t + \Delta f} \right] . \] (44)
(44) can be calculated as the superposition of the the LOS and the NLOS components, i.e.,
\[ \rho_{p,q,p',q'}(t,f,\Delta t, \Delta f) = \rho_{p,q,p',q'}^{\text{LOS}}(t,f,\Delta t, \Delta f) + \rho_{p,q,p',q'}^{\text{NLOS}}(t,f,\Delta t, \Delta f) . \] (45)
For the LOS component, the correlation is expressed as
\[ \rho_{p,q,p',q'}^{LOS}(t,f,\Delta t,\Delta f) = \frac{K^R}{K^R + 1} h_{p,q,f,i}(t) h_{p',q',f',i'}^* \left( t + \Delta t \right) e^{j2\pi\sigma_1} \] (46)
where \( \sigma_1 = f[\tau_{p,q,n}(t) - \tau_{p',q',n}(t + \Delta t)] + \Delta f \tau_{p',q',n}(t + \Delta t) \). For the NLOS component, the correlation function is calculated similarly as
\[ \rho_{NLOS}^{NLOS}(t,f,\Delta t,\Delta f) = \frac{P_{\text{remain}}(\Delta t,\delta_p,\delta_q)}{K^R + 1} \]
\[ \sum_{n=1}^{N_{p,q}(t)} \sum_{n'=1}^{N_{p',q'}(t)} P_{\text{remain}}(\Delta f) \]
\[ h_{p,q,f,i,n}(t) h_{p',q',f',i',n'}^* \left( t + \Delta t \right) e^{j2\pi\sigma_2} \]
where \( \sigma_2 = f[\tau_{p,q,n}(t) - \tau_{p',q',n}(t + \Delta t)] + \Delta f \tau_{p',q',n}(t + \Delta t) \). It is hard to present the STFCF directly because of the high dimension of the STFCF. However, by setting \( \Delta f = 0 \), \( p = p' \), and \( q = q' \), the STFCF is reduced to the time-frequency-variant ACF
\[ \rho_{ACF}^{ACF}(t,f,\Delta f) = \rho_{p,q}(t,f,\Delta t,0) \]
\[ = E \left[ H_{p,q,f,i}(t) H_{p,q,f,i}(t + \Delta f) \right] \] by setting \( \Delta f = 0 \) and \( \Delta t = 0 \), STFCF is reduced to the space-cross-correlation function (SCCF)
\[ \rho_{SCCF}^{SCCF}(f,t) = \rho_{p,q,p',q'}(t,f,0,0) \]
\[ = E \left[ H_{p,q,f,i}(f,t) H_{p,q,f,i}(f,t) \right] \] .
Similarly, by setting \( \Delta t = 0 \), \( p = p' \), and \( q = q' \), the STFCF is reduced to the time-frequency-variant frequency correlation function (FCF)
\[ \rho_{FCF}^{FCF}(t,f,\Delta f) = \rho_{p,q}(t,f,0,\Delta f) \]
\[ = E \left[ H_{p,q,f,i}(f,t) H_{p,q,f,i}(f + \Delta f,t) \right] \] .

C. Delay PSD and Delay Spread

The time-frequency variant delay PSD for the channel between the \( p \)th Tx antenna and the \( q \)th Rx antenna can be expressed as
\[ \Upsilon_{p,q}(t,f_i,\tau) = \sum_{n=1}^{N_{p,q}(t)} \sum_{m=1}^{M_{p,q,m}} \left| h_{p,q,f,i,m}(t) \right|^2 \delta(\tau - \tau_{p,q,m}(t)). \] (51)
The time-variant delay PSD is influenced by the time-dependent mean powers of rays with clusters and delays of rays parameters.

The delay spread is an important parameter in determining whether or not an adaptive equalizer is required at the receiver. If the delay spread exceeds 10 percent to 20 percent of the symbol duration, then an adaptive equalizer is required. The delay spread of the channel model can be written as
\[ \sigma_{\Delta t}(f_i) = \sqrt{\frac{\sum_{n=1}^{N_{p,q}(t)} \sum_{m=1}^{M_{p,q,m}} (\tau_{p,q,m}(t) - \mu_{\tau})^2 \Upsilon_{p,q}(t,f_i,\tau_{p,q,m}(t))}{\sum_{n=1}^{N_{p,q}(t)} \sum_{m=1}^{M_{p,q,m}} \Upsilon_{p,q}(t,f_i,\tau_{p,q,m}(t))}} \] (52)
where \( \mu_{\tau} \) is the average delay of the channel and is calculated as
\[ \mu_{\tau} = \frac{\sum_{n=1}^{N_{p,q}(t)} \sum_{m=1}^{M_{p,q,m}} (\tau_{p,q,m}(t)) \Upsilon_{p,q}(t,f_i,\tau_{p,q,m}(t))}{\sum_{n=1}^{N_{p,q}(t)} \sum_{m=1}^{M_{p,q,m}} \Upsilon_{p,q}(t,f_i,\tau_{p,q,m}(t))} \] (53)

D. Local Doppler PSD

The local Doppler PSD measures the average power as a function of the Doppler frequency. It can be calculated by the Fourier transform of the temporal ACF with respect to \( \Delta t \) and can be written as
\[ S_{p,q}(f,t) = \int_{-\infty}^{\infty} \rho_{ACF}^{ACF}(t,f,\Delta t) e^{-j2\pi f \Delta t} d\Delta t. \] (54)

E. Angular PSD and Angular Spread

The angular PSD of the channel model measures the power distribution in the angular domain. It can be calculated from the CIR of the channel model and written as
\[ \Lambda^T(t,T_\Omega,\Omega_A) = \left[ \frac{K^R}{K^R + 1} \right]^{1/2} \left[ \delta(\Omega_E - \phi_{E,L}(t)) \delta(\Omega_A - \phi_{A,L}(t)) \right] \]
\[ + \left[ \frac{K^R}{K^R + 1} \right]^{1/2} \left[ \delta(\Omega_E - \phi_{E,m}(t)) \delta(\Omega_A - \phi_{A,m}(t)) \right] \]
(55)
where \( \Omega_E \) and \( \Omega_A \) are the EAoD and AAoD at the Tx side. The angular PSD at the Rx side can be calculated in the similar method.

The cluster level angle spread reflects degree of the diffusely scattering and can also be obtained from the CIR. The cluster level angle spread \( \psi^T_{A,n} \) is expressed as
\[ \psi^T_{A,n}(t) = \sqrt{\frac{\sum_{m=1}^{M_{A,n}} \left( \Delta \phi^T_{A,m,n} \right)^2 \Lambda^T(t,\phi^T_{A,m,n})}{\sum_{m=1}^{M_{A,n}} \Lambda(t,\phi^T_{A,m,n})}}. \] (56)

F. Stationary Regions in STF Domains

For non-wide sense stationary channel, stationary interval is utilized in [52] to measure the period within which the channel amplitude response can be seen as stationary. In this model, non-stationarities in space, time, frequency domains are considered. Therefore, an improved definition for the stationary regions is extended to space and frequency domains. The stationary regions in space, time, and frequency domains are denoted as stationary distance, stationary interval, and stationary bandwidth and can be calculated as maximum distance/interval/frequency bandwidth within which the autocorrelation function of time-frequency dependent delay PSD exceeds the threshold.

The stationarity regions in three domains can be written as
\[ I(\Delta p, \Delta q, \Delta t, \Delta f) = \inf \left\{ \Delta p, \Delta q, \Delta t, \Delta f \mid R_{p,q,p',q'}(t,f,\Delta p,\Delta q,\Delta t,\Delta f) < c_{th} \right\} \] (57)
where $\inf\{\cdot\}$ calculates the infimum of a function. It should be noticed that the threshold $c_{th}$ can be adjusted for different requirements. The normalized ACF of the delay PSD $R_{\tau_{p,q;p',q'}(t,f,\Delta p,\Delta q,\Delta t,\Delta f)}$ is defined by

$$R_{\tau_{p,q;p',q'}(t,f,\Delta p,\Delta q,\Delta t,\Delta f)} = \frac{\int \Psi_{p,q}(t,f,\tau)\Psi_{p',q'}(t+\Delta t,f+\Delta f,\tau)d\tau}{\max\left\{\int \Psi_{p,q}(t,f,\tau)d\tau, \int \Psi_{p',q'}(t+\Delta t,f+\Delta f,\tau)d\tau\right\}}$$

where $\Psi_{p,q}(t,f,\tau)$ is the delay PSD of the model and $p' = p + \Delta p, q' = q + \Delta q$.

For the time domain, by setting the difference $\Delta p = 0, \Delta q = 0, \Delta f = 0$ in (58), the stationary interval is written as

$$I(\Delta t) = \inf\left\{\Delta t \mid R_{\tau_{p,q}(t,f,\Delta t)} < c_{th}\right\}.$$  \hspace{1cm} (59)

For the space domain, the stationary distance can be obtained by setting $\Delta t = 0, \Delta f = 0$ in (58) as

$$I(\Delta p, \Delta q) = \inf\left\{\Delta p, \Delta q \mid R_{\tau_{p,q;p',q'}(t,\Delta t)} < c_{th}\right\}.$$ \hspace{1cm} (60)

For the frequency domain, the stationary bandwidth can be calculated by setting the difference $\Delta p = 0, \Delta q = 0, \Delta t = 0$ as

$$I(\Delta f) = \inf\left\{\Delta f \mid R_{\tau_{p,q}(t,f,\Delta t)} < c_{th}\right\}.$$ \hspace{1cm} (61)

**IV. RESULTS AND ANALYSIS**

In this section, the statistical properties of the THz channel models are simulated and discussed. The frequency of the system is chosen from 300 GHz to 350 GHz. The TX array is set as $128 \times 128$ and The RX array is set as $16 \times 16$. Both $\delta_T$ and $\delta_R$ equal to half of wavelength at 325 GHz. To validate the proposed THz channel model, basic simulation parameters for a typical THz indoor scenario are set as follows. The elevation and azimuth angles of TX array $\beta_{V,E}, \beta_{V,\Delta A}, \beta_{H,E}, \beta_{H,\Delta A}$ are set as $-\pi/16, \pi/16$, and $-\pi/16$, respectively. The velocity of RX is $v_R = 0.6$ m/s with $\alpha_E = 0$ and $\alpha_A = \pi/2$. The TX is assumed as fixed. The bandwidth of sub-band is 0.1 GHz.

**A. STF Correlation Function**

The simulations for correlation functions are assumed to imitate indoor scenarios. The analytic results are obtained by ideal cases where the paths in each cluster are trend to infinity. However, it is unpracticable in the real hardware. The simulated results are calculated after the channel generation. It can be utilized in the real communication hardware. The generation rate $\lambda_G$ is assumed as 0.8, and the recombination rate $\lambda_R = 0.04$, the mean number of clusters is 20. The time ACF of the model can be calculated by setting $\Delta p, \Delta q, \Delta f$ as 0. The comparison of channel model at $t_0 = 0$ s and $t_1 = 5$ s at 300 GHz and 350 GHz $(D = 3 \text{ m}, \sigma_{p,q,n} = 5 \text{ m}, p = 1, q = 1, v_R = 0.6 \text{ m/s}, v^3 = 0 \text{ m/s}, \sigma_{A,n} = 2.8^\circ, \sigma_{E,n} = 1.4^\circ)$. The absolute values of FCFs for NLOS paths at different frequencies $f_1 = 300 \text{ GHz}, f_2 = 325 \text{ GHz}$, and $f_3 = 350 \text{ GHz}$ for THz massive MIMO channel model are shown in Fig. 7. We can notice that differences among FCFs at different frequencies are small but still observable which means that the non-stationarity in the frequency domain needs to be considered.

**B. Stationary Bandwidth**

The stationary bandwidth is simulated with parameters of a typical indoor scenario according to [17], [55]. The initial distance between the first elements of TX and RX is 3 m,
Cluster level angle spread reflects the degree of diffusely scattering in a cluster. In this simulation, the parameter $\sigma_{E,n}^R$ is optimized according to minimum mean square error (MMSE) criterion to fit the data from the channel measurements [10] or ray tracing [18]. In the optimization process, an iterative process is conducted to determine the minimum mean square error and its corresponding parameters. The CDFs of relative azimuth angles of arrival with different $\sigma_{E,n}^R$ are simulated and compared with the measurement data [10] in Fig. 9. The measurements in [10] were conducted in a small static office scenario with some furniture. The measured frequency was from 275 GHz to 325 GHz. The distance between Tx and Rx is 2.8 m. We can observe the good agreement when $\sigma_{E,n}^R = 1.4^\circ$. Then we simulate the scattering in two materials with different roughness. The cluster level angle spread is greatly affected by the roughness of the material. The height standard deviation of plaster1 is $\sigma_{plaster1} = 0.5 \text{ mm}$ and $\sigma_{plaster2} = 1.5 \text{ mm}$ in [18]. Fig. 10 demonstrates the comparison of cluster level angle spread with different materials in this model and simulation by ray tracing in [18]. The results show that the proposed model can approximate well with the ray tracing for different materials.
D. RMS Delay Spread

The RMS delay spread is simulated and fitted with the measurement data [56] by minimum mean square error criterion in Fig. 11. In [56], channel measurement on the desktop where the transmission distance is less than 1 m at 300 GHz was conducted. In the simulation, the transmission distance is from 0.1 m to 0.75 m and a single antenna for both the Tx and Rx is applied. The parameters $\lambda_Q = 0.2$ and $\lambda_R = 0.04$. $d^n$ are linearly increase with the LOS distance and the parameters at $D = 0.15$ m are estimated. We can observe good agreement between the simulation results and the measurement data. It means that the proposed model is suitable for this scenario.

V. CONCLUSIONS

A novel 3D general THz GBSM for 6G wireless communication systems has been proposed in this paper. The non-stationarities and cluster evolution in space, time, and frequency domains based on birth-death processes have been demonstrated. The novel general THz channel model can be reduced to specific channel models to fit various scenarios by adjusting model parameters. Statistical properties of the proposed general THz channel model such as the correlation functions and stationarity regions have been derived and simulated. The good agreements between simulated statistical properties and corresponding channel measurements have verified the accuracy and generality of the proposed THz model.

The proposed general THz channel model can be utilized in 6G THz communication systems design for different scenarios with good accuracy.

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