A Comprehensive Spatial-Temporal Channel Propagation Model for the Ultra-Wideband Spectrum 2–8 GHz

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Abstract

Despite the potential for high-speed communications, stringent regulatory mandates on Ultra-Wideband (UWB) emission have hindered its commercial success. By combining resolvable UWB multipath from different directions, Multiple-Input Multiple-Output (MIMO) technology can drastically improve link robustness or range. In fact, a plethora of algorithms and coding schemes already exist for UWB-MIMO systems, however these papers use simplistic channel models in simulation and testing. While the temporal characteristics of the UWB propagation channel have been well documented, surprisingly there currently exists but a handful of spatial-temporal models to our knowledge, and only two for bandwidths in excess of 500 MHz. This paper proposes a comprehensive spatial-temporal channel propagation model for the frequency spectrum 2–8 GHz, featuring a host of novel parameters. In order to extract the parameters, we conduct an extensive measurement campaign using a vector network analyzer coupled to a virtual circular antenna array. The campaign includes 160 experiments up to a non line-of-sight range of 35 meters in four buildings with construction material varying from sheetrock to steel.

Index Terms
Uniform circular array, Multiple-Input Multiple-Output (MIMO)

I. INTRODUCTION

Ultra-Wideband (UWB) technology is characterized by a bandwidth greater than 500 MHz or exceeding 20% of the center frequency of radiation [1]. Despite the potential for high-speed communications, the FCC mask of -41.3 dBm/MHz EIRP in the spectrum 3.1–10.6 GHz translates to a maximum transmission power of -2.6 dBm. This limits applications to moderate data rates or short range. Multiple-Input Multiple-Output (MIMO) communication...
systems exploit spatial diversity by combining multipath arrivals from different directions to drastically improve link robustness or range [2]. Ultra-Wideband lends to MIMO by enabling multipath resolution through its fine time pulses and the fact that most UWB applications are geared towards indoor environments rich in scattering provides an ideal reception scenario for MIMO implementation; in addition, the GHz center frequency relaxes the mutual-coupling requirements on the spacing between antenna array elements. For these reasons UWB and MIMO fit hand-in-hand, making the best possible use of radiated power to promote the commercial success of Ultra-Wideband communication systems.

In fact, a plethora of algorithms and coding schemes already exist for UWB-MIMO systems, exploiting not only spatial diversity, but time and frequency diversity as well [3], [4], [5]. Yet these papers use simplistic channel models in simulation and testing. While the temporal characteristics of the UWB channel have been well documented in [1], [6], [7], [8], [9], [10], [11], [12], [13], [14], surprisingly there currently exists but a handful of spatial-temporal channel models to our knowledge [15], [16], [17], [18], and only two for UWB with bandwidths in excess of 500 MHz [19], [20]. Most concentrate on independently characterizing a few parameters of the channel, but none furnish a comprehensive model in multiple environments which allows total reconstruction of the spatial-temporal response, analogous to the pioneering work in the UWB temporal model of Molisch et al. [1]. In order to fill this void, we propose a detailed UWB spatial-temporal model. Specifically, the main contributions of this paper are:

- **a frequency-dependent pathloss model**: allows reconstructing the channel for any subband within $f = 2–8$ GHz, essential to test schemes using frequency diversity, and incorporates frequency-distance dependence previously modeled separately;

- **a spatial-temporal response model**: introduces the distinction between spatial clusters and temporal clusters, and incorporates spatial-temporal dependence previously modeled sepa-
rately;

- **diverse construction materials**: to model typical building construction materials varying as sheetrock, plaster, cinder block, and steel rather than with building layout (i.e. office, residential typically have the same wall materials);

- **high dynamic range**: the high dynamic range of our system allows up to 35 meters in non-line-of-sight (NLOS) range to capture the effect of interaction with up to 10 walls in the direct path between the transmitter and receiver.

The paper reads as follows: section II describes the frequency and spatial diversity techniques used to measure the spatial-temporal propagation channel. The subsequent section explains the design and specifications of our measurement system realized through a vector network analyzer coupled to a virtual circular antenna array, and outlines our suite of measurements. The main section IV features our proposed stochastic model characterizing the channel with parameters reported individually for eight different environments; given the wealth of accumulated data furnished through our measurement campaign, we attempt to reconcile the sometimes contradictory findings amongst other models due to limited measurements. The last section summarizes our conclusions.

II. **Measuring the spatial-temporal response**

A. **Measuring the temporal response through frequency diversity**

The temporal response $h(t)$ of the indoor propagation channel is composed from an infinite number of multipath arrivals indexed through $k$

$$h(t) := \sum_{k=1}^{\infty} a_k e^{j\varphi_k} \delta(t - \tau_k),$$

(1)

where $\tau_k$ denotes the delay of the arrival in propagating the distance $d$ between the transmitter and receiver, and the complex-amplitude $a_k e^{j\varphi_k}$ accounts for both attenuation and phase change.
due to reflection, diffraction, and other specular effects introduced by walls (and other objects) on its path.

The temporal response $h(t)$ has a frequency response

$$H(f) = \sum_{k=1}^{\infty} a_k e^{j\phi_k} e^{-j2\pi f \tau_k},$$

suggesting that the channel can be characterized through frequency diversity: we sample $H(f) = \frac{Y(f)}{X(f)}$ at rate $\Delta f$ by transmitting tones $X(f)$ across the channel and then measuring $Y(f)$ at the receiver. Characterizing the channel in the frequency domain offers two important advantages over transmitting a UWB pulse and recording the temporal response directly: 1) it enables extracting the frequency parameter $\alpha$; 2) a subband with bandwidth $B$ and center frequency $f_c$ can be selected a posteriori in reconstructing the channel. The discrete frequency spectrum $X(f)$ transforms to a signal with period $\frac{1}{\Delta f}$ in the time domain [23], and so choosing $\Delta f = 1.25$ MHz allows for a maximum multipath spread of 800 ns which proves sufficient throughout all four buildings for the arrivals to subside within one period and avoid time aliasing.

B. Measuring the spatial response through spatial diversity

Replacing the single antenna at the receiver with an antenna array introduces spatial diversity into the system. This enables measuring both the temporal and spatial properties of the UWB channel. We chose to implement the uniform circular array (UCA) over the uniform linear array (ULA) in light of the following two important advantages: 1) the azimuth of the UCA covers $360^\circ$ in contrast to the $180^\circ$ of the ULA; 2) the beam pattern of the UCA is uniform around the azimuth angle while that of the ULA broadens as the beam is steered from the boresight.

Consider the diagram in Fig. 1 of the uniform circular array. The $P$ elements of the UCA are arranged uniformly around its perimeter of radius $r$, each at angle $\theta_p = \frac{2\pi p}{P}$, $p = 1 \ldots P$. The radius determines the half-power antenna aperture corresponding to $29.2^\circ \frac{r}{f_c}$ [24]. Let $H(f)$ be
the frequency response of the channel between the transmitter and reference center of the receiver array. Arrival $k$ approaching from angle $\phi_k$ hits element $p$ with a delay $\tau_{kp} = -\frac{\lambda}{c} \cos(\phi_k + \theta_p)$ with respect to the center [25], hence the element frequency response $H_p(f)$ is a phase-shifted version of $H(f)$ by the steering vector, or

$$H_p(f) = H(f) e^{-j2\pi f \tau_{kp}} = H(f) e^{j2\pi f \frac{\lambda}{c} \cos(\phi_k + \theta_p)}. \quad (3)$$

The array frequency response $H(f, \theta)$ is generated through beamforming by shifting the phase of each element frequency response $H_p(f)$ back into alignment at the reference [25]:

$$H(f, \theta) = \frac{1}{P} \sum_{p=1}^{P} H_p(f) e^{-j2\pi f \frac{\lambda}{c} \cos(\theta + \theta_p)} \quad (4)$$

The spatial-temporal response $h(t, \theta)$ can then be recovered through the Inverse Discrete Fourier Transform of its array frequency response by synthesizing all the frequencies in the subband:

$$h(t, \theta) = \frac{1}{P} \sum_{l=1}^{\frac{B}{\Delta f}} H(f, \theta) e^{j2\pi ft}, \quad (5)$$

where $f = f_c - \frac{B}{2} + l \cdot \Delta f$. 

Fig. 1. The uniform circular array antenna
III. THE MEASUREMENT SYSTEM AND CAMPAIGN

A. The measurement system

Fig. 2 displays the block diagram (a) and a photograph (b) of our measurement system. The transmitter antenna is mounted on a tripod while the UCA was realized virtually by mounting the receiver antenna on a positioning table. We sweep the $P = 97$ elements of the array by automatically repositioning the receiver at successive angles $\theta_p$ around its perimeter. At each element $p$, a vector network analyzer (VNA) in turn sweeps the discrete frequencies in the 2–8 GHz band. A total channel measurement, comprising the element sweep and the frequency sweep at each element, takes about 24 minutes. To eliminate disturbance due to the activity of personnel throughout the buildings and guarantee a static channel during the complete sweep, the measurements were conducted after working hours.

During the frequency sweep, the VNA emits a series of tones with frequency $f$ at Port 1 and measures the relative amplitude and phase $S^{21}(f)$ with respect to Port 2, providing automatic phase synchronization between the two ports. The long cable enables variable placement of the transmitter and receiver antennas from each other throughout the test area. Their height was set
to 1.7 m (average human height). The preamplifier and power amplifier on the transmit branch boost the signal such that it radiates at approximately 30 dBm from the antenna. After it passes through the channel, the low-noise amplifier (LNA) on the receiver branch boosts the signal above the noise floor of Port 2 before feeding it back.

The $S_{21}^0(f)$-parameter of the network in Fig. 2 can be expressed as a product of the $T_x$-branch, the $T_x$-antenna, the propagation channel, the $R_x$-antenna, and the $R_x$-branch

$$S_{21}^0(f) = H_{bra}^{T_x}(f) \cdot H_{ant}^{T_x}(f) \cdot H_{p}(f) \cdot H_{ant}^{R_x}(f) \cdot H_{bra}^{R_x}(f)$$

$$= H_{bra}^{T_x}(f) \cdot \left( H_{ant}^{T_x}(f) \cdot H_{ant}^{R_x}(f) \right) \cdot H_{p}(f) \cdot H_{bra}^{R_x}(f).$$  \hspace{1cm} (6)

The element frequency response $H_p$ is extracted by individually measuring the responses $H_{bra}^{T_x}$, $H_{bra}^{R_x}$, and $H_{ant}^{T}$ in advance and deembedding them from (6). In characterizing the antenna response, we separated the antennas by a distance of 1.5 m to avoid near-field operation and to reduce the effects of orientation, we spatially averaged the response by rotating the antennas with respect to each other every ten degrees. This was carried out on a flat open field with dimensions exceeding 100 m × 100 m to minimize ambient multipath to a single ground bounce which we subsequently removed by placing electromagnetic absorbers on the ground at their midpoint. Both antennas were set to a height of 1.7 m (average human height). Note in particular the following implementation considerations:

- to account for the frequency-dependent loss in the long cable when operating across such a large bandwidth, we ramped up the power at Port 1 with increasing frequency to equalize the radiated power from the transmitter across the whole band;
- we removed the LNA from the network in experiments with range below 10 m to protect it from overload and also avert its operation in the non-linear region;
- to extend the dynamic range of our system, we exploited the configurable test set option
of the VNA to reverse the signal path in the coupler of Port 2 and bypass the 12 dB loss associated with the coupler arm. The dynamic range of the propagation channel corresponds to 140 dB as computed through [26] for an IF bandwidth of 1 kHz and a SNR of 15 dB at the receiver.

B. The measurement campaign

The measurement campaign was conducted in four separate buildings on the NIST campus in Gaithersburg, Maryland, each constructed from a dominant wall material varying from sheet rock to steel. Table I summarizes the 40 experiments in each building (10 LOS and 30 NLOS), including as an element the maximum number of walls separating the transmitter and receiver. As an example, consider the floor plan of NIST North in Fig. 3: the experiments were drawn from two sets of 22 transmitter locations and 4 receiver locations (marked by the empty and solid circles respectively) to the end of achieving a uniform distribution in range in both LoS and NLoS conditions. The solid line identifies the experiment with the longest range traversing 9 walls between the transmitter and receiver. The ground-truth distance $d$ and ground-truth angle $\phi_0$ between the transmitter and receiver were calculated in each experiment by pinpointing their coordinates on site with a laser tape, and subsequently finding these values using a computer-aided design (CAD) model of each building floor plan.
IV. THE PROPOSED SPATIAL-TEMPORAL CHANNEL PROPAGATION MODEL

This section describes the proposed spatial-temporal channel propagation model. It is divided into two components: 1) the reference spatial-temporal response characterizes the shape of the two-dimensional multipath profile $h(t, \theta)$ and 2) the frequency-dependent pathloss scales its amplitude according to the distance between the transmitter and receiver and the frequency band of operation. The two corresponding subsections explain the extraction and modeling of the parameters of each component, following by a subsection that outlines in pseudocode how to implement the parameters to generate a stochastic channel response in the eight environments.

A. The frequency-dependent pathloss model

The frequency-dependent pathloss is defined as

$$PL(f) := |H(f)|^2$$

$$= \frac{1}{P} \sum_{p=1}^{P} |H_p(f)|^2$$

(7)
and so can equivalently be computed for each experiment through the measured element frequency response in (3). In order to generate a model for the pathloss, consider decomposing the arrival amplitude in (2)

\[ a_k = \tilde{a}_k \left( \frac{d}{d_0} \right)^{-\frac{n}{2}} \left( \frac{f}{f_c} \right)^{-\frac{\alpha}{2}} \] (8)

as a product of the reference amplitude \( \tilde{a}_k \) valid at reference point \((d_0, f_c)\) and the pathloss factor representing the distance and frequency dependences of the amplitude. Incorporating the frequency parameter \( \alpha \) into the model in addition to the conventional attenuation coefficient \( n \) [22] has been shown to improve channel reconstruction up to 40% for bandwidths in excess of 2 GHz [21]. Now by substituting (2) into (7) and expanding, the pathloss model can be written explicitly in terms of \( d \) to account for the distance of each experiment as

\[
(a) \quad PL(d, f) = PL(d_0, f_c) \left( \frac{d}{d_0} \right)^{-n} \left( \frac{f}{f_c} \right)^{-\alpha}; \\
(b) \quad PL(d_0, f_c) = \sum_{k=1}^{\infty} \tilde{a}_k^2.
\] (9)

The reference pathloss \( PL(d_0, f_c) \) for \( d_0 = 1 \) m and the attenuation coefficient \( n \) were extracted at the center frequency \( f_c = 5 \) GHz by fitting the model above to the data points of the experiments given with varying distance from (7). We actually found the breakpoint model [9] to represent the data much more accurately:

\[
PL(d, f_c) = \begin{cases} 
PL(d_0, f_c) \left( \frac{d}{d_0} \right)^{-n_0}, & d \leq d_1 \\
PL(d_1, f_c) \left( \frac{d}{d_1} \right)^{-n_1}, & d > d_1
\end{cases}
\] (10)
Next the frequency parameter $\alpha$ in (9a) was fit to the remaining data points by allowing the frequency to vary. Based on the Geometric Theory of Diffraction, in previous work [27] we noticed that wall interactions such as transmission, reflection, and diffraction increase $\alpha$ from the free space propagation value of zero. As the number of expected interactions increases with distance, a linear dependence of the frequency parameter on $d$ can be observed and modeled as

$$\alpha(d) = \alpha_0 + \alpha_1 \cdot d,$$

(11)

with positive slope $\alpha_1$ through all environments$^1$. Fig. 5(a) illustrates the frequency parameter versus the distance for the experiments in the *Child Care* in NLOS environment and Table II lists the parameters of the pathloss model for all eight environments.
B. The reference spatial-temporal response

Our model for the reference spatial-temporal response \( \tilde{h}(t, \theta) \) valid at \((d_0, f_c)\) essentially follows from (1) by augmenting \( h(t) \) in the \( \theta \) dimension as

\[
\tilde{h}(t, \theta) = \sum_{i=1}^{N} \sum_{j=1}^{\infty} \sum_{k=1}^{\infty} \tilde{a}_{ijk} e^{j \varphi_{ijk}} \delta(t - \tau_{ijk} - \Phi_{ijk})
\]  

and exchanging \( a_{ijk} \) with \( \tilde{a}_{ijk} \). In order to extract the parameters of the model, \( \tilde{h}(t, \theta) \) can be computed for each experiment by normalizing the measured array frequency response in (4) by the pathloss factor and so replacing it in (5) with \( \tilde{H}(f, \theta) = \frac{H(f, \theta)}{(\frac{4\pi}{\lambda})^{\frac{3}{2}} (\frac{f}{f_c})^{\nu}} \) instead. Note that the parameters of the pathloss model in the previous subsection are necessary to generate \( \tilde{h}(t, \theta) \) and so must be extracted a priori. Once generated, the arrival data points \((\tilde{a}_{ijk}, \varphi_{ijk}, \tau_{ijk}, \Phi_{ijk})\) are extracted from the responses through the CLEAN algorithm in [16]. Only the most significant arrivals, as determined by an average power threshold of 27 dB from the maximum peak in the response, were used to fit the model parameters in the sequel.

The reference spatial-temporal response partitions the arrivals indexed through \( k \) into \( N \) spatial clusters, or superclusters indexed through \( i \), and subordinate temporal clusters, or simply clusters indexed through \( j \). It reflects our measured responses composed consistently from 1) one direct supercluster arriving first from the direction of the transmitter and 2) one or more wave-guided superclusters arriving later from the door(s) (when placing the receiver in a room) or from the hallway(s) (when placing it in a hallway); the doors and hallways effectively guide the arrivals through, creating “corridors” in the response. Consider as an example the measured response in Fig. 4(a) taken in Child Care with three distinct superclusters highlighted in different colors. The partial floor plan in Fig. 4(b) shows the three corresponding paths colored accordingly and the coordinate \((\tau_i, \Phi_i)\) of each path appears as a dot on the response. The direct supercluster

\(^1\)Only Child Care in LOS exhibited a small negative slope due to lack of data where the building structure limited the longest LOS distance to only 15.3 m.
arrives first along the direct path and the later two along the wave-guided paths from the opposite
directions of the hallway. We model $N$ for NLOS through the Poisson distribution\(^2\) as

$$N \sim \mathcal{P}(\eta)$$

and set $N = 1$ for LOS\(^3\).

The notion of clusters harks back to the well-known phenomenon witnessed in temporal
channel modeling [1], [9], [29] caused by larger scatterers in the environment which induce a
delay with respect to the first cluster within a supercluster. Notice the two distinct clusters of
each wave-guided supercluster in Fig. 4(a).

1) The delay $\tau_{ijk}$:

The equations in (14) govern the arrival delays. The delay $\tau_1$ of the direct supercluster coincides
with that of the first arrival. In LOS conditions, $\tau_1$ equals the ground-truth delay $\tau_0 = \frac{d}{c}$, i.e. the
time elapsed for the signal to travel the distance $d$ at the speed of light $c$. However our previous
work [30] confirms that the signal travels through walls at a speed slower than in free space,
incuring an additional delay $(\tau_1 - \tau_0)$. As illustrated in Fig. 5(b), the additional delay scales
with $\tau_0$ according to $\Omega$ in (14a) since the expected number of walls in the direct path increases
with ground-truth delay. Based on the well-known Saleh-Valenzuela (S-V) model [29], the delay
between wave-guided superclusters $(\tau_i - \tau_{i-1})$, $i > 2$ depends on the randomly located doors
or hallways and so obeys the exponential distribution\(^4\) in (14a); so does the delay $(\tau_{ij} - \tau_{i,j-1})$
between clusters within supercluster $i$ in (14b) and the delay $(\tau_{ijk} - \tau_{i,j,k-1})$ between arrivals

\(^2\mathcal{P}(\eta) = \frac{e^{-\eta} \eta^N}{N!}\\
\(^3\)We actually observed two superclusters in all our LOS experiments, however the second arriving with an offset of 180°
relative to the first was clearly due to the reflections off the opposite walls attributed to our testing configuration in the hallways
rather than to the channel.\\
\(^4\mathcal{E}(L) = \frac{1}{L} e^{-\frac{(\tau_1 - \tau_0)}{L}}\)
within cluster $ij$ in (14c) due to randomly located larger and smaller scatterers respectively.

\[(a) \quad \tau_1 - \tau_0 = \Omega \cdot \tau_0, \quad \tau_0 = \frac{d}{c}; \]
\[(\tau_i - \tau_{i-1}) \sim \mathcal{E}(L), \quad i > 2 \]

\[(b) \quad \tau_{ij} - \tau_{ij-1} \sim \mathcal{E}(\Lambda), \quad \tau_{n1} = \tau_n \]
\[(c) \quad \tau_{ijk} - \tau_{ijk-1} \sim \mathcal{E}(\lambda), \quad \tau_{ij1} = \tau_{ij} \]

(14)

2) The angle $\phi_{ijk}$:

As the walls retard the delay of the direct supercluster $\tau_1$, they also deflect its angle $\phi_1$ from the ground-truth angle $\phi_0$ through refraction and diffraction. Our previous work [30] reveals that the degree of deflection also scales with $\tau_0$ according to $\omega$ in (15a). Concerning the angle of the wave-guided superclusters $\phi_i, i > 2$, our experiments confirm the uniform distribution in (15a) supported by the notion that the doors and hallways could fall at any angle with respect to the orientation of the receiver. The cluster angle $\phi_{ij}$ in (15b) approaches from the same angle as the supercluster due to the guiding effect of the doors and hallways, and in agreement with [15], [16], [17] the Laplacian distribution\(^5\) models the intra-cluster angle ($\phi_{ijk} - \phi_{ij}$), i.e. the deviation of the arrival angle from the cluster angle in (15c).

\[(a) \quad |\phi_1 - \phi_0| = \omega \cdot \tau_0; \]
\[\phi_i \sim \mathcal{U}[0, 2\pi), \quad i > 2 \]

\[(b) \quad \phi_{ij} = \phi_i \]
\[(c) \quad (\phi_{ijk} - \phi_{ij}) \sim \mathcal{L}(\sigma) \]

\[^5\mathcal{L}(\sigma) = \frac{1}{\frac{2\sigma}{\pi}} e^{-\frac{1}{\sigma^2} \cdot |\phi_{ijk} - \phi_{ij}|} \]
3) The complex reference amplitude $\tilde{a}_{ijk} e^{j\tilde{\nu}_{ijk}}$:

Like in the S-V model, the cluster amplitude $\tilde{a}_{ij}$ fades exponentially versus the cluster delay $\tau_{ij}$ according to $\Gamma$ in (16a) and as illustrated in Fig. 5(c); the arrival amplitude $\tilde{a}_{ijk}$ also fades exponentially versus the intra-cluster delay $(\tau_{ijk} - \tau_{ij})$ according to $\gamma(\tau_{ij})$ in (16b). Our experiments suggest a linear dependence of $\gamma$ on $\tau_{ij}$ in some buildings confirmed by other researchers [1], [9]. The parameter $s$ drawn from a Normal distribution $\mathcal{N}(0, \sigma_s)$ quantifies the deviation between our model and the measured data and in that capacity represents the stochastic
nature of the amplitude, of particular use when simulating time diversity systems [5]. The arrival phase $\varphi_{ijk}$ in (16c) is well-established in literature as uniformly distributed [23].

\[
\begin{align}
(a) \quad \tilde{a}_{ij} &= \tilde{a}_{11} \cdot e^{-\frac{1}{2} \left( \frac{\tau_{ij} - \tau_{11}}{\Gamma} \right)} \\
(b) \quad \tilde{a}_{ijk} &= \tilde{a}_{ij} \cdot e^{-\frac{1}{2} \left[ \frac{(\tau_{ijk} - \tau_{ij})}{\gamma(\tau_{ij})} + s \right]} \cdot e^{-\frac{1}{2} \frac{|\phi_{ijk} - \phi_{ij}|}{\beta(\tau_{ij})}} \\
\gamma(\tau_{ij}) &= \gamma_0 + \gamma_1 \cdot \tau_{ij}, \\
\beta(\tau_{ij}) &= \beta_0 + \beta_1 \cdot \tau_{ij}, \\
s &\sim \mathcal{N}(0, \sigma_s)
\end{align}
\]

We found the arrival amplitude $\tilde{a}_{ijk}$ also to fade exponentially versus the the intra-cluster angle $|\phi_{ijk} - \phi_{ij}|$ according to $\beta(\tau_{ij})$ in (16b). In NLOS the walls spread the arrival amplitude in angle with each interaction on the path to the receiver. The number of expected interactions increases with the cluster delay, justifying a linear dependence of $\beta$ on $\tau_{ij}$ as modeled through (16b). Fig. 5(d) illustrates this phenomenon for the Child Care in NLOS environment. However we exercise caution in generalizing this phenomenon as indeed it depends on the construction material: in NIST North, even though sheetrock walls are the most favorable of the four buildings in terms of signal penetration, the aluminum studs inside the walls spaced every 40 cm act as “spatial filters”, reflecting back those arrivals most deviant from the cluster angle and hence sharpening the clusters in angle with increasing cluster delay, as indicated through negative $\beta_1$; this dependence is less noticeable in Sound and Plant where we record $\beta_1$ an order of magnitude less in comparison to the other two buildings since the signal propagates poorly through cinder block and steel respectively, and so wave guidance defaults as the chief propagation mechanism. In the past, Spencer [15], Cramer [16], and Chong [17] have claimed spatial-temporal independence: for Spencer, these conclusions were drawn from experiments conducted in buildings with concrete
and steel walls similar to *Sound* and *Plant* respectively, where we too notice scarce dependence; for Chong and Cramer, the dependence was less observable because the experiments were conducted at a maximum distance of 14 m.

**C. Reconstructing the spatial-temporal response**

A stochastic spatial-temporal response can be reconstructed from our model through the following steps:

1) Select $d$ (and in turn $\tau_0 = \frac{d}{c}$), $\phi_0$, and the parameters from one of the eight environments in Table II;

2) Generate the stochastic variables $N$ and $(\bar{a}_{ijk}, \varphi_{ijk}, \tau_{ijk}, \phi_{ijk})$ of the arrivals from the reference spatial-temporal model in section IV-B: set $\bar{a}_{11} = 1$ in (16a) and then normalize the amplitudes to satisfy (9b), keeping only those clusters and arrivals with amplitude above some threshold;

3) Choose a subband in $f = 2$–8 GHz with bandwidth $B$ and center frequency $f_c$, and sample interval $\Delta f$; compute $H(f)$ in (2) for each sample frequency from the pathloss model in section IV-A and the generated arrivals;

4) Select $P$ and compute $H_p(f)$ in (3) from $H(f)$ for each element in the circular antenna array (note that any array shape can be used by applying the appropriate steering vector);

5) Compute $H(f, \theta)$ in (4) from $H_p(f)$ which yields the sought spatial-temporal response $h(t, \theta)$ through (5).

**V. Conclusions**

In this paper, we have proposed a detailed spatial-temporal channel propagation model with 19 parameters for the UWB spectrum 2–8 GHz in eight different environments. The parameters were fit through an extensive measurement campaign including 160 experiments using a vector
TABLE II
THE PARAMETERS OF THE PROPOSED SPATIAL-TEMPORAL CHANNEL MODEL FOR THE EIGHT ENVIRONMENTS

<table>
<thead>
<tr>
<th>environment</th>
<th>pathloss $PL(d_0;f_c)$ (dB)</th>
<th>$n_0$</th>
<th>$n_1$</th>
<th>$d_1$ (m)</th>
<th>$\alpha_0$</th>
<th>$\alpha_1$ (m$^{-1}$)</th>
<th>$\eta$</th>
<th>$\Omega$</th>
<th>$L$ (ns)</th>
<th>$\Lambda$ (ns)</th>
<th>$\omega$ (°/ns)</th>
<th>$\sigma$ (°)</th>
<th>$\gamma_1$ (ns)</th>
<th>$\gamma_2$ (°)</th>
<th>$\beta_0$ (°/ns)</th>
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<td>6.0</td>
<td>11</td>
<td>1.7</td>
<td>.015</td>
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<td>6.3</td>
<td>10</td>
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<td>.100</td>
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<td>Plant</td>
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<td>1.4</td>
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<td>NA</td>
<td>1.9</td>
<td>.030</td>
<td>1.6</td>
<td>0.52</td>
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A network analyzer coupled to a virtual circular antenna array. The novelty of the model captures the dependence on the signal propagation delay of the frequency parameter, the delay and angle of the first arrival, and the cluster shape. Most importantly for UWB-MIMO systems, our model discriminates between clusters arriving from the direct path along the direction of the transmitter and those guided through doors and hallways.

REFERENCES


