60-GHz Millimeter-Wave Propagation Inside Bus: Measurement, Modeling, Simulation, and Performance Analysis

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ABSTRACT Millimeter-wave (mmWave) transmission over the unlicensed 60-GHz spectrum is a potential solution to realize high-speed internet access, even inside mass transit vehicles. The solution involves communication between users and a mmWave-band on-board unit that aggregates/disseminates data streams from/to commuters and maintains the connection with the nearest terrestrial network infrastructure node. In this paper, we provide a measurement-based channel model for the 60-GHz mmWave propagation inside a typical inter-city bus. The model characterizes power delay profile (PDP) of the wireless intra-vehicular channel, and it is derived from about 1000 data sets measured within the bus. The proposed analytical model is further translated into a simple simulation algorithm that generates in-vehicle channel PDPs. Different goodness-of-fit tests confirm that the simulated PDPs are in good agreement with the measured data. Finally, a tapped-delay-line (TDL) channel model is formulated from the proposed PDP model, and the TDL model is used to study the bit error rate (BER) performance of the mmWave link inside bus under varying data rates and link lengths.

INDEX TERMS Intra-vehicular communication, 60 GHz channel sounding, power delay profile, tapped delay line, bit error rate.

I. INTRODUCTION
Remaining connected while on-the-move is the mantra of the next generation commuters [1]. Over the last two decades, there had been a steady increase in long distance commuting as a result of urbanization of employment opportunities [2]. According to the European Union (EU) labor force survey conducted in 2015, 8% of the EU workforce commuted to work in a different region, and 1% commuted cross-border. The average commute time per day is about 1.1 hours [3], with around 10% extreme commuters spending more than 2 hours per day [4]. Driving personal cars for commuting is a unproductive task and leads to unnecessary stress. By letting someone else to do the driving, connected commuters can utilize the unpaid time for working, gathering information, or enjoying multimedia streaming [5].

For a daily commuter, public transport vehicles (e.g., buses, trains, and ferries) are cost-effective and
environmentally friendly alternatives, providing easy access to major urban joints and alleviating the user from parking worries. On top of that, policymakers are working hard to achieve car-free cities [6], [7] through subsidies and awareness campaigns. In spite of these, the shift in the commuting mode is sluggish [8]–[10] and the transportation sector is in desperate need of a game changer. A large section of stakeholders believe that equipping public vehicles with on-board gigabit wireless networks can accelerate the shift considerably. The idea is illustrated in Fig. 1, where the access network forms a two-hop system: base stations (BSs) or road side units (RSUs) serve the vehicular access point (AP) over vehicle-to-infrastructure (V2I) mobile channels, and the vehicular AP connects to the passengers inside over static intravehicular channels.¹ The network architecture helps in avoiding the penetration loss caused by metallic bodies and signaling overhead due to group handovers [1] experienced in direct outdoor channels. There are many other potential applications of intravehicular wireless signaling beyond user connectivity, which include counting the number of passengers in a public vehicle [11] or establishing a small-scale social networking platform between co-passengers [12]. However, wireless communication infrastructures inside public vehicles should be able to provide such on-demand real-time high-data-rate diverse services.

Millimetre wave (mmWave) is a promising new technology [13] which is able to provide an enormous bandwidth to support the aforementioned diverse services in current fifth generation (5G) [14] and beyond 5G [15] networks. There had been already several initiatives to implement mmWave in intelligent transportation systems [16], [17]. In the 60 GHz unlicensed band, mmWave networks can provide up to 100 Gbit/s [18] in short-range limited-mobility scenarios, i.e., inside offices and buildings [19]–[21], or in specialized environments such as inside vehicles.

¹The outdoor channel referred in Fig. 1 is basically an outdoor-to-indoor channel penetrating the vehicle and is used by a commuter to directly connect with a BS or RSU when there is no provision of connecting to an in-vehicle AP. Further, the AP needs to be equipped with two antennas, one outside the vehicle, which connects to the nearest BS/RSU and has a wired connection to the AP fitted inside the vehicle, and another inside the vehicle attached to the AP, which connects to the user devices.

In general, mmWave propagation is significantly distinct when compared to narrowband sub-6 GHz propagation in several respects [22]: sparse multipath, high path-loss, directional transmission, lesser penetration, diffuse scattering, etc. In a closed space environment such as interior of a vehicle, the effect of some of these characteristics is magnified. Thus, experimental study of intravehicle mmWave propagation channel is of fundamental interest.

The 60 GHz mmWave propagation was earlier investigated for links between cars [23]–[25], for links inside cars [26]–[29], and for links inside aircrafts [30], [31]. Recent works also compare the 60 GHz mmWave transmission to the traditional ultra-wideband (UWB) transmission in the 3-11 GHz band [32]–[34]. However, mmWave channel models inside public transport vehicles such as buses have rarely been studied in depth. Previous works include narrowband measurements at 2.4 GHz [35], at 5 GHz [36], [37], and UWB measurements in the range of 2.3-11 GHz [38]. The only exceptions are [39] or authors’ own works such as [40] and [41], in which some initial measurement data and preliminary channel modeling is presented for 60 GHz mmWave signal propagation inside a bus.

The electronic communications committee (ECC) recommendation on using 57-64 GHz band in 2009 [42] paved the way for 60 GHz mmWave-based pan-European cross-border experiments on connected and automated driving [43]. Standardization and regulation of such EU cooperative intelligent transport systems (C-ITSs) [44] is driven by car-2-car communication consortium (C2C-CC), European telecommunications standards institute (ETSI) and European committee for standardization (CEN). In this regard, the present work is the outcome of the collaboration between research groups in three different countries, namely Czech Republic, Austria, and Spain, in which we focus on intravehicular communications. The main contributions of this paper are detailed below:

- We provide measurement data for 60 GHz frequency-domain channel sounding inside a bus. Around a thousands data points have been recorded during the field trials. Instead of using a vector network analyzer (VNA) (which has range limitations due to the cable costs), we employed a signal analyzer, enabling us to cover the whole bus length.
- Unlike a VNA, a signal analyzer used as a receiver does not record the phase information. Instead, only the magnitude of the input signal for a defined frequency range is measured. Since the phase information is critical for a subsequent time-domain analysis, we propose a post-processing technique based on the Hilbert transform to reproduce the phase of the channel transfer function (CTF) from the recorded signal amplitude and the transmitter-receiver distance (d) measured in the 3D space. The data from a previous measurement campaign [29], in which a passenger car was considered, is employed to validate the accuracy of the proposed technique.
• Power delay profiles (PDPs) are obtained from the frequency-domain sounding data by means of the complex-valued inverse fast Fourier transform (IFFT). General PDP trends are identified, which helped to perform an analytical characterization and subsequently led to a simple PDP simulation algorithm which requires only the receiver-transmitter distance \( d \) as an input. Additionally, the performance of the algorithm has been evaluated and validated with several goodness-of-fit (GoF) tests.

• Discrete-time PDPs are derived from the continuous-time PDPs, hence the channel can be modeled as a tapped-delay-line (TDL) filter for link-level simulations. The entire simulation code along with the recorded data set is made available through IEEE Code Ocean platform (DOI: 10.24433/CO.1876676.v1). The TDL model is used to study the bit error rate (BER) performance of the mmWave link inside the bus with varying data rates and link lengths.

The rest of the paper is organized as follows. In Section II, we discuss the experimental setup employed for the channel measurements. The post processing method using the Hilbert transform is presented in Section III. In Section IV, we discuss our analytical PDP model and introduce the PDP simulation algorithm. Section V presents the TDL model and the results of the BER simulations. Finally, Section VI concludes the paper.

II. FIELD EXPERIMENTS
Field experiments were performed at Brno University of Technology campus in Czech Republic. The vehicle used is a 50-seater long-distance inter-city coach (model: Mercedes Benz Tourismo BlueTec4), parked in front of a covered garage. There were no other cars on the parking lot during measurements and the immediate surroundings have no buildings or concrete structures. The bus has a dimension of 12 m (length) by 2.5 m (width) by 3.3 m (height). Inside, the floor-to-ceiling distance is 2 m, the overhead luggage rack is at 1.7 m height and each seat is 1.25 m high. The half-rows are 0.95 m wide on each side with an aisle space of 0.5 m in-between, and the distance between successive seat rows is 0.75 m. The exterior and interior of the bus is shown in Fig. 2.

The equipment used for the 60 GHz channel sounding inside the bus is also shown in Fig. 2. Channel sounding in the 60 GHz range is a non-trivial task: for the sub-40 GHz testing, 2.92 mm radio-frequency (RF) cables, connectors and off-the-shelf antennas are readily available, whereas for 60 GHz channel sounding, expensive 1.85 mm RF cables are needed and antennas have a waveguide coupling that requires waveguide to 1.85 mm cable adapters. Our measurements are performed in the frequency domain, with an experimental setup consisting of an analog signal generator (model: Agilent E8257D), a scalar signal analyzer (model: Rhode and Schwarz FSUP50), custom-built 60 GHz mmWave antennas, a power amplifier (PA) (model: Quinstar QPW 50662330) at the transmitter (Tx) end, a mixer (model: Rohde & Schwarz FS Z75) at the receiver (Rx) end, adapters, phase-stable coaxial cables (model: MegaPhase TM67), and a DC power supply (model: Diametral P230R51D). The generator output is set to 13 dBm, the PA has a gain of 31 dB and the resolution bandwidth of the analyzer is set to 10 kHz. The major advantage of the current setup over a VNA based setup is that the Tx and the Rx are connected to different hardware equipment. A general purpose interface bus (GPIB) cable connects the Tx and the Rx for synchronization, allowing for cascading two or more GPIB cables to extend the Tx-Rx distance. In contrast, for sounding with a VNA, both the Tx and the Rx must be
connected to the VNA, facing large cable losses and thus limiting the Tx-Rx distance.

A bandwidth (BW) of 10 GHz was covered, ranging from 55GHz to 65 GHz and 60 GHz being the center frequency. 1001 measurement points were recorded in each sweep. A complete sweep requires 4 minutes. The CTF amplitude, \(|H(f)|\), is recorded for each frequency point \((f_i; f_{i+1} - f_i = 10 \text{ MHz})\). The output of the signal generator is sent, by means of a cable, to the PA, which is powered by the DC power supply and exhibits a gain to compensate for the cable losses and also to boost the signal to be fed to the open waveguide type Tx antenna (OWGA). At the receiver end, the signal is captured by a substrate-integrated-waveguide slot antenna (SIW SA) and is sent to the signal analyzer. Given that the signal analyzer can only work up to 50 GHz, an external mixer down-converts the signal to an intermediate frequency (IF) of 404.4 MHz, mitigating also the high cable losses at mmWave frequencies.

The position of the Tx and the Rx antennas inside the bus is shown in Fig. 3. Since our goal is to evaluate the performance of the 60 GHz downlink channel, the Rx antenna is attached to a drop-down seat tray, imitating the typical position of a handheld personal wireless device. To cover the entire space inside the bus, the Rx antenna was placed at 15 different seats during the course of the measurements. On the other hand, the position of the Tx antenna is placed near the ceiling and the bust front window to emulate a rooftop access point. The close-up of the antenna fixtures is also shown in Fig. 3. The absorbers are used to limit reflections from metal parts of the fixtures. There are a number of WR 15 type waveguides used between the PA and the Tx antenna to point the antenna towards the desired direction.

As shown in Fig. 3, the radiation pattern of the Rx antenna is omni-directional (mobile user), whereas the Tx antenna (access point) is a directional waveguide, whose main beam is directed towards the seats to maximize directivity.

### III. POST PROCESSING

Measurement setups based on VNAs are favored to carry out frequency-domain static wireless channel measurements/sounding due to their robustness and high dynamic range. In spite of a static channel environment inside a bus, a VNA-based setup could not be employed for the channel measurements because the measurement distance would then have been restricted by the length of the coaxial cables. Contrarily, a signal analyzer cannot measure directly the phase of the captured signals, but the phase can be retrieved from the amplitude only measurement data to produce the complex-valued CTF. The complex-valued IFFT can then be applied to procure the channel response in the time domain.

Using the Hilbert Transform (HT), it is feasible to recover the phase information from the magnitude measurements of the received signal. One of the initial studies in this domain has been recorded in [45]. It considers two Hilbert-based techniques to derive the channel impulse response (CIR) in the sub-6 GHz frequency band, hence establishing the employability of a Hilbert-based estimation approach to recover phase information from amplitude only data. Similar HT-based techniques applied at higher frequencies have been presented in [46] and [47], but requiring that the CTF, \(H(f)\), be a minimum phase function. The basic idea presented in...
those papers can be demonstrated in the following manner. The complex-valued CTF in the frequency-domain can be written in polar form as

$$H(j\omega) = |H(j\omega)| \exp(j \arg(H(j\omega))).$$

(1)

and taking logarithms on both sides we have

$$\tilde{H}(j\omega) = \log_e(|H(j\omega)|) + j \arg(H(j\omega)),$$

(2)

where $\log_e(|H(j\omega)|) = \log_e(|H(j\omega)|) + j \arg(H(j\omega))$. Defining $\tilde{h}(n) = \mathcal{F}^{-1}[\tilde{H}(j\omega)]$ as the Fourier inverse of $\tilde{H}(j\omega)$, then the phase information can be obtained from the amplitude response [45] as follows

$$\arg(H(j\omega)) = \mathcal{H}[^{\log_e(|H(j\omega)|)}],$$

(3)

where $\mathcal{H}[\cdot]$ denotes the Hilbert transform assuming that $\tilde{h}(n)$ is causal, i.e., $\tilde{h}(n) = 0; n < 0$.

A. PROPOSED METHOD

In this subsection, we present our HT based approach to recover the phase information. The CTF need not be a minimum phase function in this case. However, it must be a causal function, a significantly weaker condition inflicted on the channel response.

The objective is to project the amplitude data into the real-valued component of the regenerated CTF employing the HT. Consequently, the phase delay is first derived as a function of the frequency. The phase difference ($\Delta \phi$), path difference ($\Delta x$), wavelength ($\lambda$), speed of light ($c$) and frequency ($f$) are related as $\Delta \phi = 2\pi \Delta x / \lambda$ and $c = \lambda f$. During our measurements, we recorded the Tx-Rx distance ($d$) for each pair of Tx-Rx combinations, which we use as the path difference. Combining the above mentioned equations, the phase delay function is then defined as:

$$\phi_d(f) = (2\pi d f) / c.$$  

(4)

The Tx-Rx propagation delay is compensated by the phase delay function, $\phi_d(f)$, hence the phase information appears right from the line-of-sight (LoS)/ strongest arrival path instant. The real-valued part is obtained as the cosine component of the CTF magnitude:

$$\text{Re}(\tilde{H}(f)) = |H(f)| \cos(\phi_d(f)).$$

(5)

This real-valued part is then fed as the argument of the hilbert operator:

$$\tilde{H}(f) = \mathcal{H}[\text{Re}(\tilde{H}(f))],$$

(6)

that generates the complex-valued CTF, $\tilde{H}(f)$. After applying the IFFT [48] on it, the corresponding PDP realization is obtained.

B. VALIDATION AND COMPARISON

To prove the validity/ accuracy of our method, we take advantage of the previous measurement campaign [29] in which 60 GHz intra-vehicular channel sounding was carried out in a sedan car employing a VNA. The phase data is first removed from the recorded measurements and then we try to regenerate the complex-valued CTF, $\tilde{H}(f)$, following the method proposed in the previous subsection.

Figs. 4 and 5 show the retrieved amplitude and phase plots, respectively, corresponding to a specific Rx-Tx position. This procedure is performed for all the data sets corresponding to the various Rx-Tx settings. The results have exhibited excellent agreement between both the estimated amplitude and phase plots and the true measurement data.

Next, IFFT is applied to the regenerated complex-valued CTFs which are then utilized to produce the PDPs. This is then compared with the PDPs produced directly by applying the IFFT on the VNA recorded data. Fig. 6 presents the PDP plots for four exemplary Tx-Rx settings. The generated PDPs closely resemble the corresponding measured PDP trails.

Table 1 summarizes the goodness-of-fit (GoF) results that gauge the similarity between the generated and the measured values of the PDP. We have considered three GoF metrics, correlation coefficient, root mean square error (RMSE) and Kolmogorov-Smirnov (K-S) test statistic. The correlation coefficient ($\hat{\rho}$) has been modeled according to [49]:

$$\hat{\rho} = \frac{\frac{1}{N} \sum_{n=1}^{N} |P(n)||P_g(n)|}{\sqrt{\frac{1}{N} \sum_{n=1}^{N} |P(n)|^2} \sqrt{\frac{1}{N} \sum_{n=1}^{N} |P_g(n)|^2}},$$

(7)

where $P(n)$ and $P_g(n)$ denote the data points for measured and generated PDPs, both of length $N$. Furthermore, the RMSE
has been evaluated as [50]

$$\text{RMSE} = \sqrt{\frac{1}{N} \sum_{n=1}^{N} |P(n) - P_g(n)|^2},$$  

(8)

and the two-sample K-S test has been evaluated according to [51]

$$K = \max[F(|P(n)|) - F(|P_g(n)|)],$$  

(9)

where $K$ denotes the statistic test of the K-S test and $F$ is the cumulative distribution function (CDF). The high values of the correlation coefficients demonstrate the existence of a strong association between the generated and the measured data. Low values of the RMSE between the two datasets and the two-sample K-S test statistic calculated with a 5% significance level corroborates the claim.

The efficacy of the proposed method is further proven as the time dispersion parameters are also estimated, i.e., the mean delay time ($\bar{\tau}$) and the root mean square (RMS) delay spread ($\tau_{\text{rms}}$). These are the first moment and the second central moment of the PDP, respectively, and they are defined as [52]

$$\bar{\tau} = \frac{\sum_{\tau=0}^{\tau_{\text{max}}} \tau P(\tau)}{\sum_{\tau=0}^{\tau_{\text{max}}} P(\tau)},$$

(10)

and

$$\tau_{\text{rms}} = \sqrt{\frac{\sum_{\tau=0}^{\tau_{\text{max}}} (\tau - \bar{\tau})^2 P(\tau)}{\sum_{\tau=0}^{\tau_{\text{max}}} P(\tau)}},$$

(11)
FIGURE 7. Generated PDP and CTF using the method proposed in [46]. The same car data is used to obtain the figures. Comparison of Fig. 7b with Fig. 4 and Fig. 7a with Fig. 6 shows that basic HT can give inaccurate results while the proposed modified HT gives accurate results.

FIGURE 8. General PDP trends inside the bus.

1) When viewed in log scale, the PDPs decay in a nonlinear manner.
2) Again, in the log scale, the gross decay slope across the PDPs are quite close for identical $d$ values.
3) There exists a power law relationship between the peak amplitude, say, $A(d)$, and $d$, across the PDPs. In the log-log scale, this is reflected as a linear decrease of the peak amplitude with distance.
4) The PDP values oscillate between a lower bound and an upper bound. The interval between the two boundaries does not change significantly with time delay.

A. ANALYTICAL PDP MODEL

From these observations, we may assume a PDP function of the form

$$f_d(\tau) = A(d) \exp[B(d) \tau],$$

where, as defined earlier, $A(d)$ is the peak amplitude while $B(d)$ denotes the decay rate, and both are functions of $d$. A visual inspection inspires us to set power-law relation for both the functions, i.e. $A(d) = \alpha d^m$ and $B(\tau) = \beta d^n$, which describes $f_d(\tau)$ as

$$f_d(\tau) = \alpha d^m \exp(\beta d^n \tau). \tag{13}$$

The empirical values of $\log(A(d))$ with $\log(d)$ are plotted across several experiments in Fig. 9. Consequently, the values $m = -1.3891$ and $\alpha = -5.434 \times 10^{-2}$ are obtained from the line of best-fit.

From (13) we can write

$$\log_e \left[ \frac{f_d(\tau)}{(\alpha d^m)} \right] / \tau = \beta d^n. \tag{14}$$

From the regression plot in Fig. 9, it is possible to obtain the values for $\alpha$ and $m$. Next, with the help of (14), we can find the maximum likelihood values of $\beta$ and $n$ in a similar manner for all experiments (the corresponding fittings are omitted for brevity). After analyzing and calibrating the values of $\beta$ and $n$ across all the measured experimental data sets, we set them to $n = -0.4$ and $\beta = -1 \times 10^{-12}$.

Next, we define an adjustable boundary function, $b(\tau)$, to account for the variations between the upper and the lower bounds,

$$b(\tau) = \left( 1 + \rho \log \left| \frac{\tau}{\tau_{\text{scale}}} + 1 \right| \right) f_d(\tau), \tag{15}$$

where $\rho$ and $\tau_{\text{scale}}$ govern the rate of decay of the small-scale PDP variation in (15) and are set to $\rho = 0.09$ and $\tau_{\text{scale}} = 10$. 

FIGURE 9. $\log(A(d))$ versus $\log(d)$ regression plot.
This boundary function helps us define the upper limit ($b_U$) and the lower limit ($b_L$) of the PDP as

$$b_U(\tau) = (1 - B/100 + \varepsilon)b(\tau), \quad (16)$$

and

$$b_L(\tau) = (1 + \varepsilon)b(\tau), \quad (17)$$

where $B$ is the range of the PDP values expressed in dBm units and $\varepsilon$ lowers the PDP curve by 100 dBm. The following set of values, $\varepsilon = 0.005, B = 10$, was found to be optimum.

The simulated PDP, $P_S(\tau)$, for every time delay $\tau$, can be finally obtained by generating a uniformly distributed random number,

$$P_S(\tau) = \mathcal{U}(b_L(\tau), b_U(\tau)),$$

where, $\mathcal{U}(\cdot, \cdot)$ returns a uniform random number ranging between its arguments.

### B. SIMULATION OF PDP

The method of generating PDP, which is discussed in Section IV-A in detail, is summed up in this subsection in the form of a pseudocode (see Algorithm 1). Our simulation algorithm takes Tx-Rx distance ($d$) as input and returns simulated channel PDP ($P_S(\tau)$) as output. The algorithm produces a continuous-time PDP starting from zero to a maximum time delay $\tau_{\text{max}}$ with a sampling rate $\tau_{\text{res}} = 1/BW$. Considering the dimensions of the bus, the parameter $\tau_{\text{max}}$ is set to 50 ns.

**Algorithm 1** Channel PDP Simulation

**Input:** $d$ [Tx-Rx distance]

**Assignments:** [Parameters]

- $\alpha \leftarrow -5.434 \times 10^{-2}$
- $\beta \leftarrow -1 \times 10^{-12}$
- $m \leftarrow -1.3891$
- $n \leftarrow -0.4$
- $B \leftarrow 10$
- $\rho \leftarrow 0.09$
- $\varepsilon \leftarrow 0.005$
- $\tau \leftarrow 0$
- $\tau_{\text{scale}} \leftarrow 10$
- $\tau_{\text{res}} \leftarrow 10^{-10}$
- $\tau_{\text{max}} \leftarrow 50 \times 10^{-9}$

1. while $\tau \leq \tau_{\text{max}}$
2. $f_d(\tau) := \alpha d^m \exp(\beta d^\tau)$
3. $b(\tau) := \left(1 + \rho \log \left(\frac{1}{\tau_{\text{scale}}} + 1\right)ight)f_d(\tau)$
4. $b_U(\tau) := (1 - B/100 + \varepsilon)b(\tau)$
5. $b_L(\tau) := (1 + \varepsilon)b(\tau)$
6. $P_S(\tau) := \mathcal{U}(b_L(\tau), b_U(\tau))$
7. $\tau \leftarrow \tau + \tau_{\text{res}}$
8. end while

**Output:** $P_S(\tau)$ [Simulated channel PDP]

A sample simulated PDP is shown in Fig. 10 along with the PDP obtained from the measurement data. The generated and measured PDPs compare pretty well with a strong correlation, the coefficient being greater than 0.7.

### C. VALIDATION OF THE SIMULATION METHOD

To assess the performance of the simulation model, the same set of GoF tests are employed as done in Section III-B, with replacing $P_S(n)$ with $P_S(n)$, where $P_S(n)$ denote the data points for simulated PDPs. The values correlation coefficient, RMS error and K-S test are enlisted for 9 cases in Table 3.

![Simulated PDP](image)

**Figure 10.** Simulated PDP [$d = 972 \text{ cm, } \hat{\rho} = 0.7230$].

**Table 3.** GoF between measured and simulated PDPs.

| Expt. No. | $d$ [cm] | Correlation coefficient | RMSE ($\times 10^{-10}$) | K-S test statistics |
|-----------|----------|------------------------|--------------------------|---------------------|
| 1         | 224      | 0.74                   | 4.16                     | 0.2072              |
| 2         | 366      | 0.71                   | 2.48                     | 0.2837              |
| 3         | 370      | 0.78                   | 2.43                     | 0.3924              |
| 4         | 512      | 0.71                   | 1.38                     | 0.2032              |
| 5         | 516      | 0.71                   | 2.22                     | 0.2334              |
| 6         | 666      | 0.72                   | 0.96                     | 0.1694              |
| 7         | 818      | 0.63                   | 1.86                     | 0.1855              |
| 8         | 970      | 0.71                   | 2.11                     | 0.2051              |
| 9         | 972      | 0.73                   | 2.11                     | 0.1515              |

The comparison is performed for all the measured test points, and PDPs calculated from the measured CTFs are compared to the corresponding PDPs simulated with the measured $d$ value. In order to validate the proposed intravehicle channel model simulation, we have visualized the CDFs from the two-sample K-S test shown in Fig. 11 for a typical measurement scenario with a Tx-Rx separation of 5.16 m.

Finally, in Fig. 12 we show a comparison of the time dispersion parameters for the measured and the generated data sets using a comparative bar graph, showing how closely the simulated channel time dispersion parameters resemble the measured channel parameters across all the 9 experimental data sets.

**Figure 11.** CDFs of both measured and simulated PDPs.

### V. BER PERFORMANCE ANALYSIS

A direct application of the proposed PDP simulation algorithm is the performance evaluation of the mmWave link in

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terms of the bit error rate (BER), which requires to discretize the PDP and derive an equivalent TDL model.

### A. TDL MODEL

In a multipath fading wireless channel, if there exists a small number of distinctive multipath components (MPCs) or if the multipath components can be grouped into clusters where the path delays between clusters can be resolved but the MPC delays within a cluster are non-resolvable, then the channel can be modeled in the time-domain as a tapped-delay-line (TDL) filter. In the TDL model, delays between taps differentiate one scatterer group from the other and each tap represents the contribution of an individual scatterer group. The impulse response for the channel in this case can be conveniently expressed in the form

\[
h(t, \tau) = \sum_{i=0}^{N-1} G_i \gamma_i(t) \delta(t - \tau_i),
\]

where \(N\) is the total number of taps, and \(G_i, \gamma_i(t)\) and \(\tau_i\) are the gain, small-scale fading and delay of the \((i+1)\)th tap, respectively.

For our model we considered a uniform delay of \(\tau_{i+1} - \tau_i = 5\) ns between taps and for a \(\tau_{\text{max}} = 50\) ns, the TDL model consisted of \(N = 10\) taps. The first tap is normalized \((G_0 = 0\, \text{dB}, \, \gamma_0 = 1)\), denoting the LoS/strongest path, whereas the remaining tap gains are derived by sampling the simulated PDP and the small-scale fading coefficients following a Rayleigh process, \(\gamma_i \sim \mathcal{CN}(0, 1)\), where \(\mathcal{CN}(\mu, \sigma^2)\) denotes circularly symmetric complex-valued Gaussian random variable with mean \(\mu\) and variance \(\sigma^2\). Fig. 13 shows a typical channel realization obtained following our TDL model.

In Table 4, we enlist the tap gains obtained from the simulated PDPs for all the taps and for four different Tx-Rx separations. The distance values are chosen to represent different sections of the bus; a \(d\) value of 1.66 m (2.35 m) denotes the second (third) row from the front, the distance is 5.76 m to the seat near the middle door, and in the back seat \(d\) goes up to 9.72 m. As one can notice, the tap gains decay sharply at the front rows, whereas at the back seat, the tap gain remains constant for the last four to five taps. The slow decrease of the tap gain indicates that the effect of multipath will be much more prominent at the back seats.

### B. SIMULATION OF BER

The TDL model derived in the previous subsection is utilized to simulate the bit error rate performance. A general Monte Carlo simulation approach is followed and equivalent baseband modulation is employed to speed up the simulation. The BER simulations were performed in MATLAB and the block diagram for the BER evaluation is shown in Fig. 14.
During the simulation, a bit-stream of $B_N = 10^6$ data bits were transmitted through the channel after modulation. The modulated signal is defined as

$$x(t) = \sum_{k=0}^{B_N} \hat{x}(k) \delta(t - k \tau_p),$$  \hspace{1cm} (20)

where $\tau_p$ is the sampling period of the input which, assuming a simple binary phase shift keying (BPSK) modulation, i.e. $\hat{x}(k) \in \{+1, -1\}$, renders $R = 1/\tau_p$ to be the transmitted bit rate and $E_b = E[\hat{x}^2(k)] \tau_p$ to be the energy transmitted per bit.

The modulated signal is fed as the input to the multipath fading channel. The tap coefficients for various distances are taken from Table 4. The received signal is,\footnote{Each time an averaging over $10^3$ runs were performed to test BER up to the range $10^{-7}$.} the received signal is, $y(t) = x(t) * h(t) + n(t)$, where $*$ denotes the convolution operator and $n(t)$ is additive Gaussian white noise (AWGN) with double-sided power spectral density $N_0$. The output $y(t)$ is truncated from $t = 0$ to $t = B_N \tau_p$ and demodulated to recover the input signal. The demodulated bit-stream, $\hat{y}(t)$, is deduced as

$$\hat{y}(k) = \text{sign}[y(t - k \tau_p)],$$  \hspace{1cm} (21)

and is compared to $\hat{x}(k)$. The number of mismatches are divided by $B_N$ to compute the BER. During the simulations, BER is determined as a function of the signal-to-noise ratio (SNR), $E_b/N_0$.

Fig. 15 shows the BER variation with different data rates for a fixed $d$ value of 1.66 m. BER curves are shown for three different data rates, 50 Mbit/s, 100 Mbit/s and 200 Mbit/s. These data rates are comparable to data rates available with wired cable modem or 4G cellular LTE networks. Fig. 15 shows that the SNR requirement may go up by 10 dB when the required data rate doubles. For lower bit rates, the effect of inter-symbol interference is reduced, as the information bits are spaced out. The AWGN BER curve is also shown as a reference.

Next, we plot the BER curves for varying Tx-Rx distances in Fig. 16 keeping the bit rate fixed at 100 Mbit/s. As the distance increases, the multipath environment degrades the signal. In our case, the access point is placed in front, so the back seat passenger experiences an additional SNR penalty of about only 10 dB compared to the front row passengers. This observation nullifies the myth that mmWave cannot propagate over larger distances in confined environments. The BER curves indicate that if a proper link margin is maintained, it is possible to use 60 GHz hardware to provide broadband access with desired reliability.

VI. CONCLUSION

The article presents measurement-based modeling and performance evaluation of 60 GHz mmWave wireless link inside a bus. Comprehensive measurements are conducted to study the PDP behavior which led to the development of an analytical framework followed by an algorithmic flowchart to simulate PDPs. Simulated PDP values are used to derive a TDL equivalent link model which forms the basis for link performance evaluation. It is observed that the distance of the user from the access point and the specified data rate has a large impact on the BER performance of the intravehicular mmWave link. Although the modeling and analysis is carried out for a bus, the characterization is applicable for other public transport vehicles (e.g. subway coaches, trams and trolleybuses) which have similar internal structures and comparable dimensions.

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