

# Performance analysis of optical extra-WBAN links based on realistic user mobility modeling

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**Abstract.** We address the features channel characterization and performance evaluation for wireless body-area networks (WBANs) in medical applications during a walk scenario using optical wireless transmission. More specifically, we focus on optical extra-WBAN uplink communication between a central coordinator node (CN) placed on the patient's body and an access point (AP) in a typical hospital room. To characterize the optical wireless channel, we use a Monte Carlo ray tracing-based method and take into account the effects of body shadowing and mobility based on realistic models, in contrast to the previous simplistic models considered in the literature. Using this approach, we derive the dynamic behavior of the channel DC gain for different configurations of the CNs and APs. Furthermore, based on the obtained results, we develop a statistical channel model based on kernel density estimation, which we use to investigate the impact of CN and AP placement on the communication link parameters. Also, based on the outage probability criterion, we discuss the link performance and further analyze the improvement in performance achieved through spatial diversity, i.e., by using multiple APs in the room, for different photodetector types, under different background noise conditions. The presented results show that CN placement and user's local and global mobility significantly impact the performance of extra-WBAN links, which can nevertheless be reduced using spatial diversity. Finally, the presented performance analysis shows that a single AP equipped with an avalanche photodiode photodetector allows an acceptable link performance for low-to-moderate background noise conditions, whereas multiple APs equipped with PIN photodetectors should be used in the case of moderate-to-strong background noise. © The Authors. Published by SPIE under a Creative Commons Attribution 4.0 International License. Distribution or reproduction of this work in whole or in part requires full attribution of the original publication, including its DOI. [DOI: [10.1117/1.OE.61.2.026113](https://doi.org/10.1117/1.OE.61.2.026113)]

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## 1 Introduction

E-health solutions are attracting increasing attention in many application scenarios as efficient systems for improving the quality of life and reducing health-care expenditures.<sup>1</sup> This popularity is driven by the advances made in different fields of engineering. The very large-scale integration of sensors and actuators in millimeter size chips<sup>2,3</sup> and the development of wireless communication technologies<sup>4</sup> allow relatively low-cost practical implementations of these solutions. Typical application examples include real-time remote monitoring of patients and elderly people in health-care centers or at home. Medical wireless body-area networks (WBANs) can be considered as the main building block of such systems by enabling the acquisition and transmission of timely data from several medical sensors. The most common architecture of WBANs is a star topology that connects these sensors to a coordinator node (CN), placed on the patient's body.

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The CN then transmits the collected data to a mobile device or a fixed access point (AP) through what is called an extra-WBAN link.<sup>5</sup>

Most previous works on extra-WBAN links have been based on the radio frequency (RF) technology, primarily using the unlicensed ISM (industrial, scientific, and medical) frequency bands. For instance, in Refs. 6 and 7, analytical models were proposed for describing body movements and postures in WBAN scenarios. In Ref. 8, a soldier-to-soldier communication link was characterized using an avatar derived from three-dimensional (3D) animation. Also, a motion capture-based model was used in Refs. 9 and 10 to study the extra-WBAN channel. However, RF-based solutions are subject to electromagnetic interference with the other existing networks that use the ISM band, in addition to data security issues that may arise. Also, the transmit power level is restricted to satisfy the specific absorption rate to avoid damaging the patients' tissues.<sup>11</sup>

These considerations have motivated the investigation of optical wireless communications (OWC) as an alternative or complementary solution to RF data transmission for medical WBANs. The main advantages of OWC are their immunity to external interference, their inherent security due to the confinement of light in indoor environments, and the negligible impact of multipath (small scale) fading.<sup>12</sup>

Note that the optical and RF WBAN channels are fundamentally different in terms of small-scale fading and sensitivity to shadowing. A few studies have considered OWC-based WBANs in the infrared band. For instance, Ref. 13 studied the optical intra-WBAN channel using a simplified model for the patient's body and mobility. In previous works,<sup>14,15</sup> we considered intra-WBAN channel characterization and performance evaluation using more sophisticated body mobility models. However, the fundamental differences between intra- and extra-WBAN links, in particular, in terms of shadowing and link blockage, necessitate a dedicated study for the latter case.

To study the OWC-based extra-WBAN links, a Monte Carlo ray tracing (MCRT)-based simulation tool is used in Ref. 16 to study a physical activity monitoring system for a moving patient while neglecting the effect of the patient's body. The authors in Ref. 17 studied downlink extra-WBAN transmission while neglecting local body movements and assuming that the receiver (Rx), i.e., the CN, is at a short distance from the body. In Refs. 18 and 19, user mobility was taken into account based on uniform-distributed random user positions and random orientations of sources and detectors. However, Ref. 18 neglected the body model, and Ref. 19 considered a 3D rectangular shape model, without any consideration for local mobility, and hence body parts' shadowing. Neglecting the effect of mobility, Ref. 20 proposed a site-specific ray tracing approach that considers the skin tissue for the channel modeling of WBAN communication links in hospital scenarios. Finally, neglecting the user's body modeling, Ref. 21 studied the performance of optical code-division multiple access (O-CDMA) for uplink extra-WBANs considering the orientation-based random waypoint (RWP) model for user mobility. A preliminary experimental study on the use of O-CDMA for extra-WBANs was further presented in Ref. 22. For a more comprehensive survey on channel modeling for optical WBANs, the reader is referred to Ref. 20. Note that, in addition to the simplified body and mobility models used, the considered models in the above-mentioned works do not take into consideration the notion of walk trajectories and, hence, neglect the correlation that exists between consecutive sequences of the movement.

This work investigates channel modeling and performance of OWC-based extra-WBANs in the uplink. In contrast to the previous works on the topic, this study is based on more accurate and realistic channel models with an acceptable level of details, in particular, concerning the body's 3D shape and user mobility. More precisely, a dynamic mobility model based on a 3D animation of a walk cycle is proposed to model body parts movement. Also, a modified RWP model is used for the trajectories taken by the patient within the room. This accuracy in modeling the body and its movements is a key requirement for accurate channel characterization since the change of body postures can cause variable shadowing that can significantly affect the link availability. Also, the time correlation between consecutive sequences of the walk trajectory needs to be taken into account in the system design. Our study is primarily based on the calculation of the channel impulse response (CIR) and its main related parameters. The link performance is quantified in terms of the outage probability considering the on-off keying (OOK) modulation. We further investigate the performance improvement achieved by using

multiple APs in the room. Two cases of PIN and avalanche photodiode (APD)-based RxS are considered with different background noise conditions.

The remainder of the paper is organized as follows. Section 2 presents the considered simulation approach and parameter specification. Then, numerical results are provided in Sec. 3 to study the channel DC gain for different configurations. The link outage probability is evaluated in Sec. 4, where the performance gain offered by using multiple APs is also studied. Section 5 concludes the paper.

## 2 Simulation Approach

This section describes our approach to optical channel characterization and performance analysis.

### 2.1 Channel Characterization

To estimate the channel CIR, we perform MCRT simulations using Opticstudio software from Zemax,<sup>23</sup> which can model complex 3D environments. The interest in using comprehensively sophisticated and extremely-time-consuming ray tracing simulations with OpticStudio is that the obtained results are close enough to reality as it was already demonstrated in Ref. 24. For an OWC link, the CIR  $h(t)$  is calculated as follows<sup>14</sup>:

$$h(t) = \frac{1}{N_{\text{ray}}} \sum_{i=1}^{N_{\text{ray}}} P_{ri} \delta(t - \eta_i), \quad (1)$$

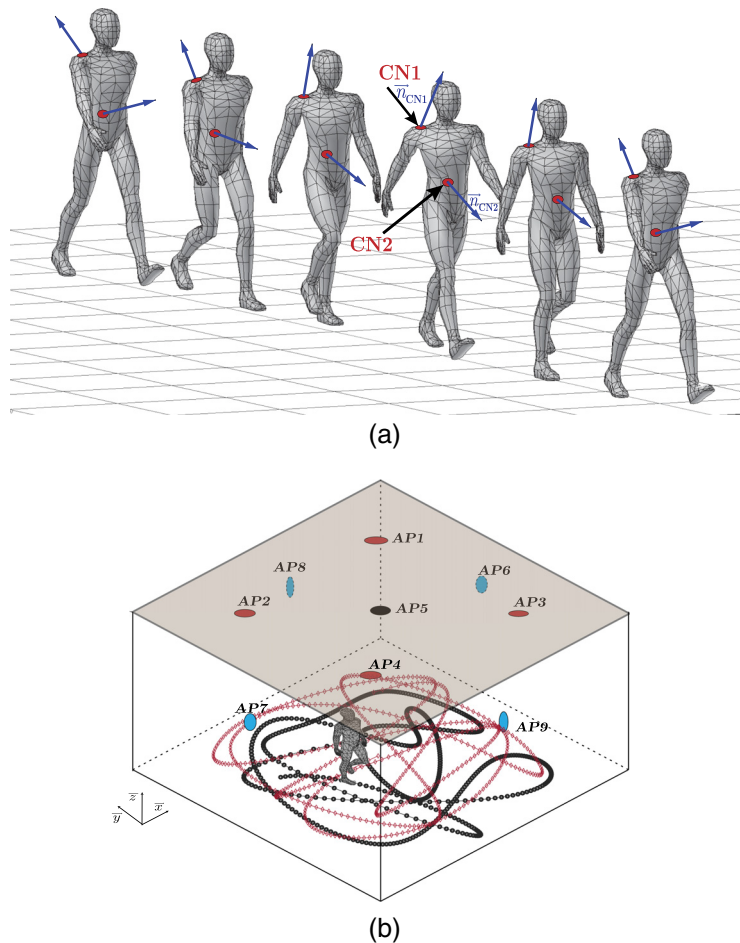
where  $N_{\text{ray}}$  is the number of rays that hit the detector,  $P_{ri}$  is the received power of the  $i$ 'th ray, and  $\eta_i = \sum_{r=0}^{N_{\text{ref},i}} \frac{d_{r,i}}{c}$  is the corresponding time delay. Also,  $\delta(\cdot)$  is the Dirac delta function,  $d_{r,i}$  stands for the optical path length between consecutive reflection surfaces hit by the ray,  $N_{\text{ref},i}$  is the reflection order of the  $i$ 'th ray, and  $c$  denotes the speed of light. Note that  $h(t)$  accounts for both the line-of-sight (LOS) and diffuse links that result from first- and higher-order reflections. A more detailed description of CIR calculation using Opticstudio can be found in Ref. 14.

Based on the calculated CIR, the main metric that determines the signal-to-noise ratio at the Rx is the channel DC gain, defined as  $H_0 = \int h(t) dt$ .<sup>12,25</sup> Other relevant channel metrics include the delay spread  $\tau$  and the coherence time  $T_c$ . The parameter  $\tau$  determines channel frequency selectivity and the need to channel equalization, whereas  $T_c$  describes channel time variations based on which are fixed the frame length and training sequences [used for channel estimation and adaptive power control at the transmitter (Tx)].<sup>26</sup> In Ref. 27, we showed that the optical extra-WBAN channels can be considered as frequency non-selective up to a data rate of 10 Mbps. Also,  $T_c$  varies between 0.9 and 4.2 s, which indicates slow channel time variations. Readers are referred to Ref. 27 for detailed numerical results. Consequently, in the sequel, we only focus on the channel DC gain as the main channel metric. Compared to Ref. 27, in this paper, we propose a more detailed investigation and interpretation of channel characterization with more accurate statistical models based on additional walk trajectories. Furthermore, we investigate the link performance for different Rx types, taking into account different noise sources. Also, we consider new configurations with different signal combination strategies in the case of using multiple APs.

### 2.2 User Mobility Modeling

To study the effect of user mobility, we take into account both the changes in the position of the user inside the room and the movement of the body parts, i.e., the arms, legs, etc., that we will refer to as global and local mobility, respectively.

For local mobility, we model the body by a 3D mesh<sup>28</sup> composed of 902 vertices, which are connected to form 1800 triangular faces. This mesh provides an acceptable level of detail to take into account shadowing effects from the body segments. The body movements are modeled



**Fig. 1** (a) Illustration of the mobility configuration with the considered placements for the CNs. (b) Considered placements of APs and the simulated walk trajectory (AP1 to AP5 are placed on the ceiling and AP6 to AP9 on the walls).

using a 3D walk cycle animation generated by the Blender software.<sup>29</sup> The walk cycle consists of  $N_f = 27$  frames representing body configurations at different instants of the walk. Guidelines for creating the walk cycle animation are available in Ref. 30. Figure 1(a) illustrates a few frames of the walk cycle.

For global mobility, a modified RWP mobility model<sup>14</sup> is used, where, instead of linear interpolation between the waypoints,<sup>31</sup> we use cubic splines to generate a curved trajectory.<sup>32</sup> This allows to consider a smooth path without sharp rotations, which is more realistic. The synchronization between the local and global mobility is done by calculating the optimal time step  $\Delta t$  for which we change the consecutive walk cycle frames, and it is given by

$$\Delta t = \frac{1}{N_f} \frac{d_{\text{cycle}}}{v_{\text{user}}}, \quad (2)$$

where  $N_f$  denotes the number of frames of the walk cycle,  $d_{\text{cycle}}$  is the distance traveled by the user during a walk cycle, and  $v_{\text{user}}$  is the user speed during that time. Considering a reasonable  $d_{\text{cycle}}$  of 2 m and a walk speed of 0.45 m/s,<sup>33,34</sup> we have  $\Delta t \approx 0.17$  s.

As suggested in Refs. 13 and 19, two positions are considered for the CN, i.e., on the shoulder and on the waist that we will refer to as CN1 and CN2, respectively. Both CN1 and CN2 are oriented perpendicular to their respective triangular faces [see Fig. 1(a)]. Note that as the body moves, the configuration of its triangular faces changes, thus, the positions and orientations of the CNs, placed on these faces will change accordingly. In fact, the reliability of the link

between the CN and the AP may depend on the CN placement. For example, the link corresponding to CN1 is less sensitive to shadowing from body parts than CN2, whereas the link corresponding to CN2 is more subject to diffuse links through first-order reflections.

We consider different placements for the APs on the ceiling and the walls, as shown in Fig. 1(b) together with the two generated trajectories shown in black and red using the modified RWP model. As can be seen, these trajectories cover most of the room area. This allows the derivation of accurately enough mobility statistics, as we will see later in Sec. 4. Also, AP1 to AP5 are symmetrically placed on the ceiling and point downward, whereas AP6 to AP9 are placed on the walls perpendicularly at half-width and 2-m height.

### 2.3 Signal Transmission Formulation

Justified by its implementation simplicity, intensity modulation with direct detection based on uncoded non-return to zero (NRZ) OOK modulation is considered. The received signal at the photodetector (PD) output is

$$r_n = \mathcal{R}GP_kH_0 + n_n, \quad (3)$$

where  $\mathcal{R}$  and  $G$  stand for the PD responsivity and gain, respectively. Also,  $P_k$  denotes the power of the transmitted symbol, which equals  $P_1$  for ON symbols or  $P_0 = \alpha P_1$  for OFF symbols, with  $\alpha$  being the extinction ratio. The average transmit power is  $P_t = \frac{(1+\alpha)P_1}{2}$ . Furthermore, in Eq. (3),  $n_n$  is the total photocurrent noise that accounts for thermal, background, dark, and shot noises. It is modeled as Gaussian with variance  $\sigma_{\text{tot}}^2$  and given by

$$\sigma_{\text{tot}}^2 = \sigma_0^2 + \sigma_{s,k}^2, \quad (4)$$

where  $\sigma_0^2 = \sigma_{\text{th}}^2 + \sigma_{\text{b}}^2$  with  $\sigma_{\text{th}}^2 = 4K_bT_eB/R_L$  and  $\sigma_{\text{b}}^2 = 2eG^2FBI_b$  being the variances of thermal and background noises, respectively.<sup>35</sup> Here,  $e$  is the electron charge,  $K_b$  is the Boltzman constant,  $T_e$  the equivalent noise temperature, and  $F$  the PD excess noise factor. Also,  $B$  denotes the Rx low-pass filter (LPF) bandwidth, typically approximated by half of the data rate  $B \approx R_b/2$ ,  $R_L$  is the load resistor of the trans-impedance amplifier (TIA), and  $I_b$  is the average photocurrent generated by a PIN PD due to the background noise.<sup>36</sup> Besides, in Eq. (4) the variance of the shot noise corresponding to OFF and ON symbols is denoted by  $\sigma_{s,k}^2 = 2eGFBI_k$ , with  $I_k = \mathcal{R}GP_kH_0$  being the received current of the corresponding symbol. We reasonably neglect the dark current noise compared with the other noise sources.<sup>35</sup>

#### 2.3.1 Statistical channel modeling

To avoid time-consuming MCRT simulations with Opticstudio, we propose a statistical model for the channel DC gain. The model is developed based on the nonparametric kernel density estimation (KDE),<sup>37</sup> which is applied to the obtained results for the simulated user trajectory (see Sec. 2.2). Note that, in contrast to the classical parametric approach, where the probability density function (PDF) is assumed to be known up to a finite number of parameters, the non-parametric methods assume that the PDF belongs to an infinite collection of functions and, hence, offer more modeling flexibility and are not affected by the choice of an inappropriate parametric model.<sup>38,39</sup>

The KDE model is defined for a set of  $N$  realizations  $x_i$  of a random variable  $x$  with the PDF  $f_X(x)$ . The KDE estimate of the PDF is given by<sup>40</sup>

$$\hat{f}_X(x) = \frac{1}{N\rho} \sum_{i=1}^N \mathcal{K}\left(\frac{x-x_i}{\rho}\right), \quad (5)$$

where  $\mathcal{K}$  is the kernel and  $\rho = 1.06\hat{\sigma}N^{-1/5}$  is the smoothing parameter. Here,  $\hat{\sigma}$  is the estimated standard variation of  $\log(x)$  related to the dataset of Eq. (5).<sup>41</sup> For the sake of simplicity, a Gaussian kernel is considered here.

### 2.3.2 Bit-error rate calculation

Using the statistical models obtained in the Sec. 2.3.1, we randomly generate  $10^7$  values of  $H_0$  for every CN-AP extra-WBAN link. Then, for each channel realization, we calculate the bit-error rate (BER) for an uncoded NRZ OOK-based link for a given  $R_b$  and  $P_r$ . Assuming optimal maximum likelihood detection taking into account the signal-dependent noise, we have<sup>42</sup>

$$\text{BER} = \frac{1}{4} \operatorname{erfc} \left( \frac{\gamma_{\text{th}} - \alpha I_1}{\sqrt{2(\alpha \sigma_{s,1}^2 + \sigma_0^2)}} \right) + \frac{1}{4} \operatorname{erfc} \left( \frac{I_1 - \gamma_{\text{th}}}{\sqrt{2(\sigma_{s,1}^2 + \sigma_0^2)}} \right), \quad (6)$$

where  $\operatorname{erfc}(x) = \frac{2}{\sqrt{\pi}} \int_x^\infty \exp(-u^2) du$  is the complementary error function and  $\gamma_{\text{th}}$  is the optimal detection threshold. Also,  $I_1$  is the PD output photocurrent for ON symbols.

### 2.4 Parameter Specification

To ensure accurate enough CIR estimation,  $N_{\text{rays}} = 2.5 \times 10^7$  rays are considered for the MCRT simulation with at least three-order reflections for beam propagation. The Blender software is used to create the 3D scene for the simulator, similar to the approach used in Ref. 14. A Python script is then used to load a set of configurations specifying user movements and orientations for  $N_{\text{config}} = 1103$  sequential snapshots of a walk scenario. We consider two trajectories of 516 and 587 sequential snapshots as explained previously and illustrated in Fig. 1(a), which, overall, give an acceptable coverage of the room area and a detailed enough statistical description of the walk cycles. Note that generating these data has necessitated quite long Opticstudio simulations.

For the sake of simplicity, we consider an empty room, i.e., without any furniture or medical devices, with dimensions  $(5 \times 5 \times 3) \text{ m}^3$  with plaster walls and ceiling, and pinewood floor. The reflectance of materials is modeled by a wavelength-dependent classical Lambertian bidirectional reflectance distribution function.<sup>12</sup> The reflection properties of the different surfaces are fixed following Ref. 43. Given the low reflectivity of the skin compared to that of the walls,<sup>44</sup> the user body is assumed to be absorbent. Although cloths' reflectivity can range from 0.02 to 0.75,<sup>45</sup> we choose here the worst-case scenario that maximizes beam shadowing effects.

At the Tx (i.e., CN) side, an LED with wavelength  $\lambda_0 = 850 \text{ nm}$  is considered with a bandwidth of  $\Delta\lambda = 30 \text{ nm}$  and a Lambertian radiation pattern.<sup>12</sup> To ensure minimum beam blockage, a Lambertian order of  $m = 1$  is assumed, corresponding to a semi-angle of 60 deg. At the Rx (i.e., AP) side, the PD is assumed to have an  $1 \text{ cm}^2$  active area, without any lens and with a large field-of-view (FOV) of 60 deg (semi-angle). The relatively large PD surface is justified by the relatively low data rates, on the order of 1 Mbps at most, required in medical WBANs.<sup>46</sup> Two types of a PIN PD and an APD (with  $G = 50$  and  $F = 3$ ) are considered in this study. Table 1 summarizes the main system parameters.

**Table 1** Simulation parameters.

Tx (CN)	$m$	1	Room	Dimensions	$(5 \times 5 \times 3) \text{ m}^3$
	$\lambda_0$	850 nm		Walls reflectivity	0.83 (Plaster)
	$\Delta\lambda$	30 nm		Floor reflectivity	0.87 (Pinewood)
	OOK, $\alpha$	0.2		Ceiling reflectivity	0.83 (Plaster)
Rx (AP)	FOV	60 deg	Body	Dimensions	$(1.7 \times 0.3 \times 0.2) \text{ m}^3$
	PD active area	$1 \text{ cm}^2$		Reflectivity	0 (Absorbing)
	$\mathcal{R}$	0.65 A/W			
	TIA $R_L$	50 $\Omega$			



For the case of multiple APs, two approaches of equal-gain combining (EGC) and selection combining (SC) are considered. For the former approach, signals received on the APs are simply added together (assuming perfect time synchronization between them) before proceeding to demodulation, whereas the latter approach consists in performing demodulation on the strongest received signal.<sup>47</sup> To avoid any confusion, it is worth mentioning that we do not consider a multicell architecture here, and the use of multiple APs is only for improving the link reliability.

### 3 Channel Characterization

#### 3.1 Single AP Configurations

First, consider the case of a single AP at the center of the ceiling or on the wall, i.e., AP5 or AP8 in Fig. 1(b). Figure 2 shows the time evolution of  $H_0$ , for the two cases of CN1 and CN2 considering the first simulated walk trajectory (i.e.,  $N_{\text{config}} = 516$ ). The abscissa represents the simulation time, with a time step of  $\Delta t = 0.17$  s.

Notice first the larger  $H_0$  in average for CN1, compared to CN2. Interestingly, for a given CN position, the average  $H_0$  is more or less the same, regardless of the position of the AP. The mean  $H_0$  corresponding to CN1 and CN2 cases is around  $-53$  and  $-55$  dB, respectively. As for  $H_0$  variations, we notice for both AP5 and AP8, larger variations and longer mean time between fades (MTBFs) for CN1 case. For example, the range of  $H_0$  variations for AP5 is around 9.2 and 4.4 dB, for CN1 and CN2 cases, respectively. The corresponding standard deviation is about 3 dB for CN1, and it ranges between 1 to 2 dB for CN2. In addition, compared to AP5, larger  $H_0$  variations are observed for AP8, with ranges about 12.8 and 7.3 dB, for CN1 and CN2 cases, respectively. These relatively large variations can be explained by the fact that, compared to AP5, AP8 is more sensitive to body shadowing caused by body rotations during the walk trajectory.

Note that the effects of mobility can be divided into two parts: effects of local mobility resulting from possible shadowing and blockage due to the periodic movements of body parts (legs and arms) and effects of global mobility resulting from the movements of the user inside the room. We can see the effects of local and global mobility in Fig. 2: For the case of CN1, placed on the shoulder and directed toward the ceiling, the link is less subject to shadowing from the body parts. Hence, the observed  $H_0$  variations are mainly due to global mobility. Depending on the position and the orientation of the user relative to the AP, we may observe a strong or a weak  $H_0$ , with a high range of variations. Also, the absence of periodic patterns in the time evolution of  $H_0$  for this case supports the argument of negligible local mobility effect on CN1.

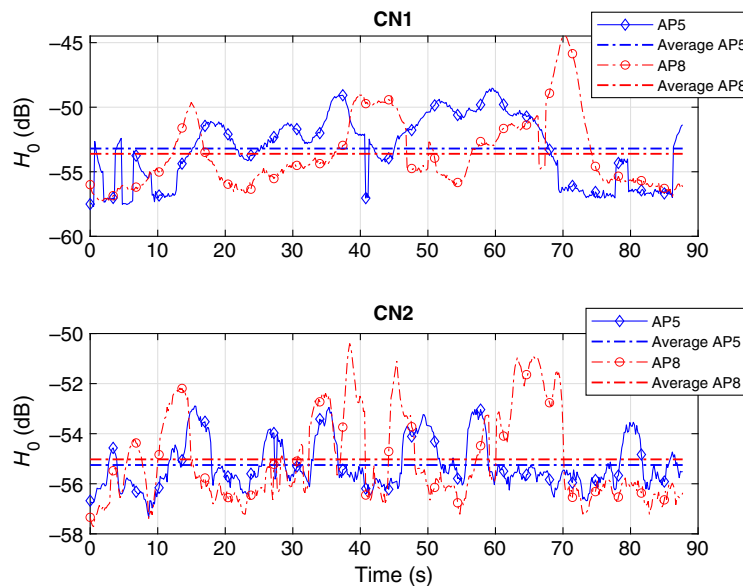


Fig. 2  $H_0$  variations for the cases of CN1 and CN2 for AP5 and AP8 configurations.

On the other hand, for the case of CN2, placed on the waist, the link is more affected by local mobility. For instance, the time evolution of  $H_0$  is periodic for both links to AP5 and AP8, with different magnitudes for each walk cycle. The latter is attributed to global mobility, whereas the former can be attributed to the shadowing resulting from the periodic swing of the body parts during a walk cycle. For instance, the right arm swinging back and forth might periodically obstruct the link to CN2.

It is worth mentioning that the observed values of  $H_0$  are not affected by the user walking speed as far as the walk pattern (i.e., the local mobility model) remains unchanged. This is because the effect of Doppler spread is negligible in OWC networks.<sup>12</sup> Meanwhile, an increased walking speed results in a slight decrease in the channel coherence time.

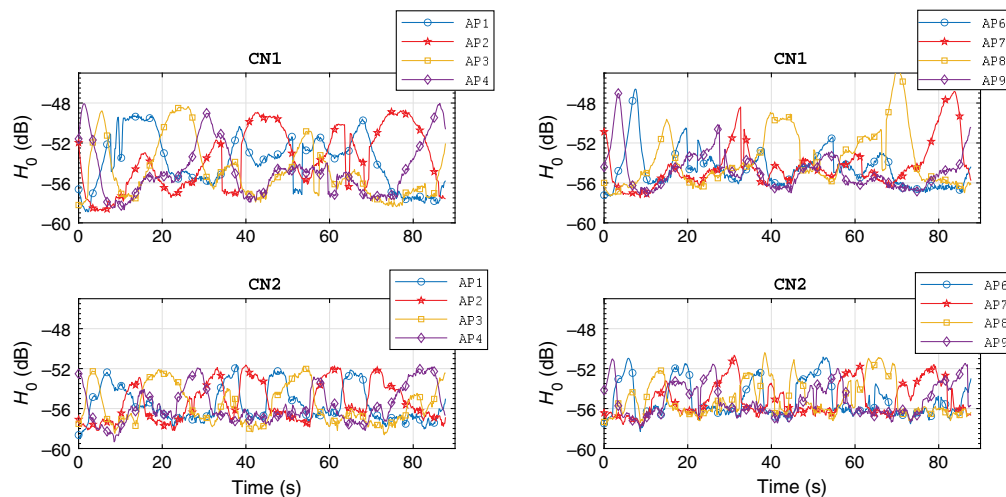
The above discussions testify to the significant impact of user mobility on the extra-WBAN optical channel. To reduce this impact, we investigate the use of spatial diversity in the following subsection.

### 3.2 Multiple AP Configurations

Consider the case of using four APs for two configurations of APs placed on the ceiling (i.e., AP1 to AP4) or on the walls (i.e., AP6 to AP9), see Fig. 1(b). Let us first consider the case of individual APs separately. In Fig. 3 the time evolution of  $H_0$  is shown for the cases of CN1 and CN2. Note that CN1 on average has a slightly larger  $H_0$  for the case of APs located on the ceiling (around 0.5 dB difference), whereas CN2 has a larger  $H_0$  for the case of APs located on the walls: the mean  $H_0$  for the case of CN1 and CN2 is  $-54$  and  $-55$  dB, respectively. Notice also higher  $H_0$  variations for the case of CN1 with the APs located on the ceiling, compared with APs on the walls: the corresponding standard deviation of  $H_0$  is about 3 and 2 dB, respectively. This can be explained by the fact that the links from CN1 to AP1 to AP4 experience frequent transitions between the availability or absence of the LOS path. On the other hand, CN2 has similar  $H_0$  variations for both configurations, where its standard deviations are around 1.7 dB.

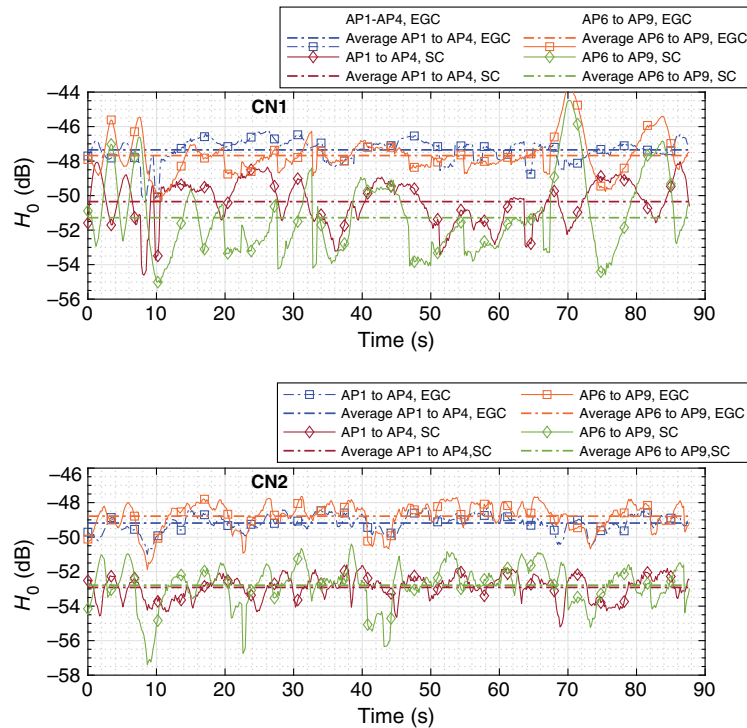
Interestingly, in the case of CN2, less correlation is observed between the signals received on different APs. In other words, a higher effective diversity is obtained in this case. This can be explained by the short distance between CN1 and the Aps, which gives a strong LOS, and the coverage area of CN1 is smaller than CN2, which makes it subject highly to user's movements.

Figure 4 shows the variations of  $H_0$  for the cases of EGC and SC techniques. As expected, the use of multiple APs, in particular with EGC, increases the mean  $H_0$  and especially reduces its standard deviation due to spatial diversity. Compared to the single AP configuration, the increase in the mean  $H_0$  is around 5.8 to 6.2 dB with EGC, and around 2.2 to 2.8 dB with SC, for both (AP1 to AP4) and (AP6 to AP9) cases. Meanwhile, compared with CN2, we notice a higher average  $H_0$  (about 1.1 to 2.6 dB) and larger  $H_0$  variations (about 1.4 to 3.6 dB) for the case of



**Fig. 3**  $H_0$  variations for the cases of CN1 and CN2 considering different APs on the ceiling (AP1 to AP4) and on the walls (AP6 to AP9).





**Fig. 4**  $H_0$  variations for the cases of CN1 and CN2 for the multiple-AP configurations using EGC and SC detection methods.

CN1, similar to the single AP configurations. Also, larger variations and shorter MTBFs are observed for the case of (AP6 to AP9), compared with (AP1 to AP4). This would suggest that the (AP1 to AP4) configuration is more robust against user mobility effects. Based on this, in the sequel, only the case of (AP1 to AP4) configuration is considered.

## 4 Link Performance Analysis

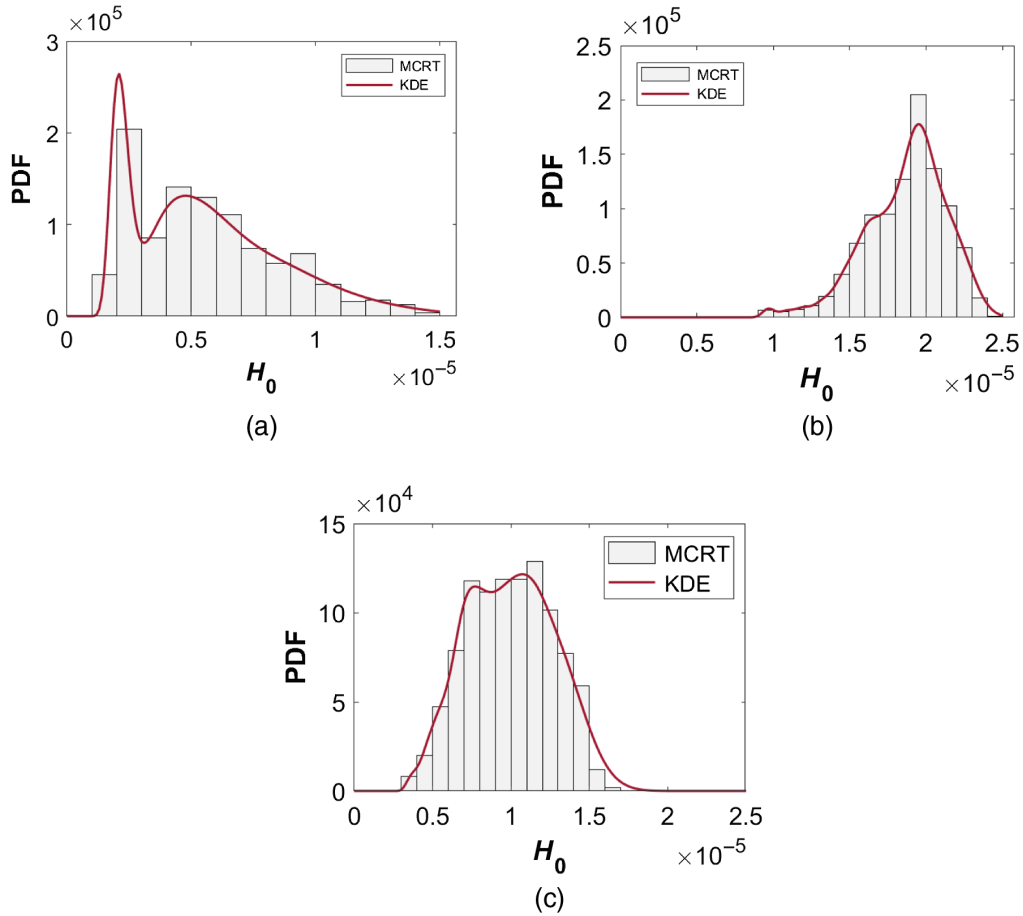
### 4.1 Statistical Model of Channel Gain

As explained in Sec. 2.3, to calculate the link BER,  $H_0$  is generated randomly based on Gaussian-kernel KDE for the considered scenarios. For this, we reasonably combine the data of the two generated trajectories to increase the accuracy of the statistical estimation and also to have a better representation of the movement area. For instance, in Fig. 5, the 1103 samples histograms of the simulated  $H_0$  for the case of CN1 via Opticstudio are presented together with the KDE-based PDF fits. The corresponding KDE parameters are specified in Table 2.

First, notice a good fit between the histograms of the simulated data and the estimated PDFs, in particular for the multiple AP cases with EGC. Also, the reduction in the histogram spread (in particular, for EGC) shows the benefit of spatial diversity in reducing shadowing and link blockage effects.

### 4.2 Performance Comparison of Single and Multiple AP Configurations

Using the obtained approximate PDFs, we randomly generate more than  $10^7$  channel realizations. For each generated  $H_0$ , the BER is calculated from Eq. (6). Given the slow time-varying nature of the channel, the outage probability  $P_{\text{out}}$  is considered for the link performance analysis, which is defined as the probability that the instantaneous link BER goes more than a given threshold  $\text{BER}_{\text{th}} = 10^{-3}$ . For the case of PIN-based Rx, Fig. 6 shows plots of  $P_{\text{out}}$  versus  $P_t$  for  $R_b$  ranging between 10 kbps to 1 Mbps, for the cases of CN1 and CN2, and single or multiple APs on the ceiling, i.e., AP5 or (AP1 to AP4). Considering first the single AP case, notice the degradation of

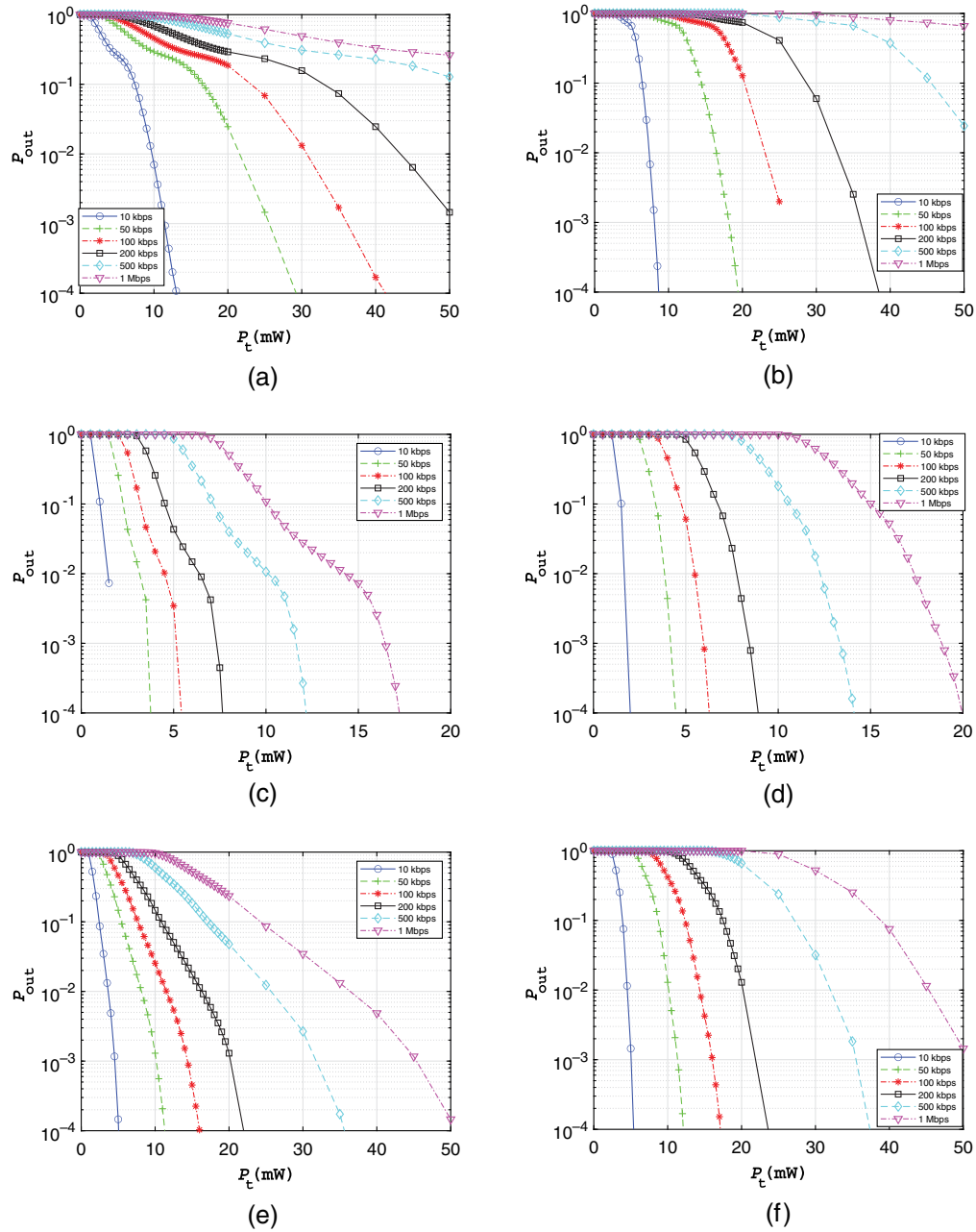


**Fig. 5** Histograms of  $H_0$  (in linear scale) for 1103 channel realizations and estimated PDF by KDE with Gaussian kernel: (a) CN1 and AP5, (b) CN1 and AP1 to AP4 with EGC, and (c) with SC.

**Table 2** Calculated smoothing parameter  $\rho$ .

Configuration		CN1	CN2	
Single AP	Ceiling (AP5)	0.1579	0.0402	
	Wall (AP8)	0.1115	0.0731	
Multiple AP	Ceiling	EGC	0.0339	0.0314
		SC	0.0813	0.0462
	Walls	EGC	0.0453	0.0417
		SC	0.1138	0.0608

$P_{out}$  with an increase in  $R_b$ , as expected. In fact, the Rx LPF bandwidth is changed according to  $R_b$ . Consequently, for increased  $R_b$ , the Rx noise variance increases, resulting in a BER degradation. For the multiple AP case, from Figs. 6(c) and 6(d), for a given  $R_b$ , we reasonably note a considerably lower requirement of  $P_t$  with EGC to attain a target  $P_{out}$ . For instance, for  $P_{out} = 10^{-3}$  and  $R_b = 100$  kbps, the required  $P_t$  is around 5.5 and 18 mW for the cases of EGC and SC signal detection, respectively, compared with 40.5 mW for the case of single AP from Fig. 6(a). This elucidates the tangible enhancement of the link reliability achieved by spatial diversity. Another interesting result is that the CN2 case offers a better performance than CN1 for the single AP configuration. This can be explained by the fact that the link to CN1 depends highly on the availability of the LOS, whereas the link using CN2 depends rather on first-order reflections.

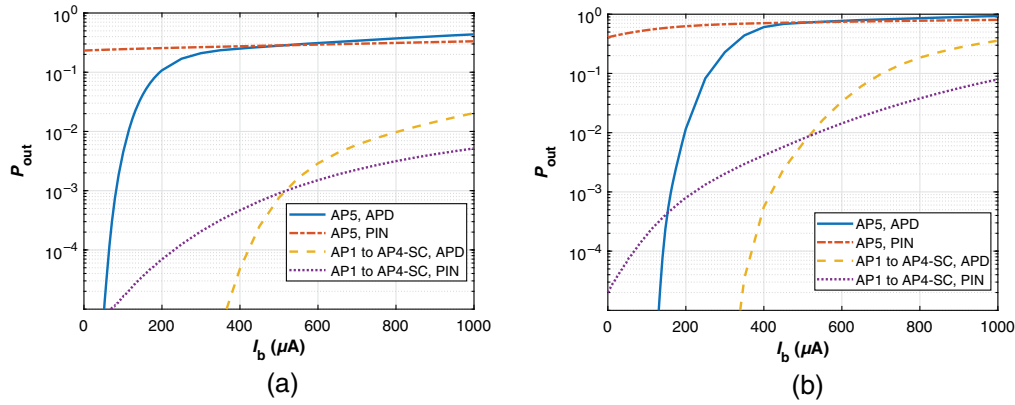


**Fig. 6**  $P_{out}$  versus the transmit power for different data rates for the cases of (a) CN1 and AP5, (b) CN2 and AP5, (c) CN1 and AP1 to AP4 with EGC, (d) CN2 and AP1 to AP4 with EGC, (e) CN1 and AP1 to AP4 with SC, and (f) CN2 and AP1 to AP4 with SC. Uncoded OOK modulation.

Given that the availability of LOS is less likely than that of first- and higher-order reflections, the link to CN2 can be considered to be more stable. However, for the multiple AP configuration, the links to CN1 and CN2 have similar performance for both EGC and SC strategies. This similarity is again due to the spatial diversity that increases the likelihood of LOS and diffuse link availability, with the result of reduced shadowing and link blockage effect.

### 4.3 Performance Comparison of PIN- and APD-based Rx's

Figure 7 shows the performance of PIN and APD versus  $I_b$  with the two combination strategies for the case of CN1 and CN2,  $P_t = 15$  mW and  $R_b = 200$  kbps (recall that  $I_b$  denotes the average photocurrent due to background noise).



**Fig. 7**  $P_{\text{out}}$  versus background noise current  $I_b$  for PIN- and APD-based Rxs with different AP configurations: (a) CN1, (b) CN2. Uncoded OOK modulation.  $P_t = 15$  mW,  $R_b = 200$  kbps.

First, as it could be expected, the multiple AP configuration with EGC offers the best performance for both cases of CN1 and CN2. The corresponding  $P_{\text{out}}$  is too low for the considered  $I_b$ , this is why they are not shown in the figure. From Fig. 7(a), considering a target  $P_{\text{out}} = 10^{-3}$  and the APD-based Rx, the link can operate up to an  $I_b$  of 80 and 510  $\mu\text{A}$ , for the cases of AP5 and AP1 to AP4 with SC detection, respectively. Also, as was expected,<sup>35</sup> the performance of the APD-based Rx is more affected by  $I_b$ , compared to the PIN-based Rx. For example, from Fig. 7(b), considering AP1 to AP4 with SC detection, for  $I_b$  up to 525  $\mu\text{A}$ , the APD-based Rx has a better performance, whereas the inverse holds for  $I_b > 525$   $\mu\text{A}$ . Interestingly, the case of single AP with APD-based Rx outperforms the PIN-based multiple AP with SC detection for  $I_b \lesssim 155$   $\mu\text{A}$ . This  $I_b$  corresponds to environments with filtered ambient indirect sunlight.<sup>36,48</sup>

## 5 Conclusions

This work investigated channel characterization and performance study for medical optical extra-WBAN links, taking into account user mobility using a dynamic model based on a 3D animation of the user walk cycle and a modified RWP model. Based on an accurate MCRT approach, we calculated the CIR and the channel DC gain and studied the impact of the CN position on the user body and AP placement at different locations of the room. The presented results and analysis are quite insightful in the design of medical WBANs by taking into account practical and realistic system parameters and channel characteristics.

In particular, we demonstrated the benefit of using multiple-APs with EGC and SC combining strategies in improving link reliability due to the reduction of beam shadowing during user movements, which justifies the increased network implementation complexity. Placing the APs on the ceiling was found to provide a better performance, given the lower DC gain variations it offers. Also, our study revealed that the placement of the CN on the waist is more advantageous, as compared to the position on the shoulder, especially for the single AP configuration. Finally, we investigated the use of PIN- and APD-based Rxs and showed that except under high background noise conditions, using one APD-based AP provides acceptable link performance. However, when working under moderate-to-strong background noise conditions, we recommend using multiple PIN-based APs.

Future work can consider the case of multiple users in a room and the development of efficient multiple-access techniques, as in Ref. 21. Also, to use in large indoor spaces, appropriate multicell architectures need to be designed with the possibility of hand-over, supporting user mobility.

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