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# Generalized Time Slot Index Modulation for Optical Wireless Communications 

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#### Abstract

A novel modulation scheme for an indoor optical wireless communication (OWC) system referred to as generalized time slot index modulation (GTIM) is proposed for single-carrier frequency domain equalization (SC-FDE) systems. This scheme is further referred to as SC-GTIM. In SC-GTIM, information bits are flexibly encoded onto both the amplitudes and positions of pulses. A set partitioning algorithm is proposed to construct a codebook for the SC-GTIM system. In this paper, SC-GTIM will be compared with the conventional pulse amplitude modulationbased SC-FDE (SC-PAM) and time slot index modulation-based SC-FDE (SC-TIM). By considering a multipath optical wireless channel and the limited dynamic range of a light emitting diode (LED), a considerable bit error ratio (BER) gain can be achieved by SC-GTIM. It is concluded that SC-GTIM is a promising technique for low to medium spectral efficiency transmission and uplink schemes.


Index Terms-PAM, frequency domain equalization, index modulation, optical wireless communication.

## I. Introduction

THE recent forecasts in 2016 show that $60 \%$ of mobile data traffic is offloaded to a fixed network, e.g., WiFi or femtocell [2]. An exponentially growing number of mobile devices, followed by the ever-increasing mobile data traffic, is predicted to increase sevenfold in 2021. Moreover, cutting edge technologies, namely augmented/virtual reality (AR/VR), internet-of-things (IoT), massive machine-type communications (MMTC) and ultra high definition video/gaming applications create a data greedy ecosystem. This progressive demand will rapidly drain the capabilities of the existing radio communication technologies. To cope with this issue, higher frequencies (above 6 GHz ) spectra are being utilized, such as milimeter wave ( 30 GHz to 300 GHz ) [3], terahertz ( 300 GHz to 3 THz ) [4] and optical wireless communications (OWC) (10 THz to 1 PHz ) [5]. Specifically, OWC can enable a high area spectral efficiency that can further boost the network capacity [6], [7]. Furthermore, OWC offers an unlicensed and very

[^0]wide spectra which opens abundant opportunities for wireless systems to meet future mobile data traffic demands [7].

Inspired from spatial modulation (SM) [8], [9], which encodes both constellation symbols and transmit antenna index, the term index modulation (IM) is first introduced in [10]. Instead of the transmit antenna index, subcarriers are used in [10]. This method is referred to as subcarrier-IM orthogonal frequency division multiplexing (OFDM). These transmit entities, namely transmit index (space domain) and subcarrier index (frequency domain), can be extended to other domains such as time domain [11], [12] or combinations of them [13], [14]. In fact, IM is not only limited to these domains, and could, for example, include orbital angular momentum (OAM) [15]. Recent survey papers in [16]-[18] summarize that IM appears to be a potential candidate for the future wireless communication systems due to its spectral and energy efficiency as well as lower hardware complexity. These advantages can also be adopted to OWC, and this paper aims to focus on this topic.

In this paper, we specifically focus on the application of time-domain IM, namely time slot IM (TIM), to intensity modulation and direct detection (IM/DD)-based OWC. We base our work on the recent application of TIM to a singlecarrier frequency domain equalization (SCFDE)-based system as discussed in [11]. In [11], both time slots and constellation symbols are simultaneously encoded at the transmitter side. In the context of single carrier (SC) transmissions in IM/DDbased OWC, SCFDE systems have been studied in [19], [20]. It is concluded that the SCFDE systems offer lower computational complexity and lower peak-to-average power ratio (PAPR) compared to OFDM-based and carrierless amplitude and phase (CAP) systems. To the best of our knowledge, both limitations and advantages of TIM to SCFDE systems in OWC have not been investigated in detail.

Contribution: First, we show that the advantages of applying TIM to pulse-amplitude modulation (PAM)-SCFDE with linear dynamic range of light emitting diode (LED) are very limited. These limitations are mainly due to the constraint of choosing a fixed number of active time slots, which can cause the minimum Euclidean distance of the constellation symbols is smaller than that of the conventional system. Then, a novel TIM, namely generalized TIM (GTIM), which removes the constraint and allows arbitrary number of active time slots, is proposed. In addition, we also propose a set partitioning algorithm to design a good codebook which maximizes both Hamming and Euclidean distances. Consequently, the error performance of the proposed system shows a significant gain
compared to both TIM-based and conventional systems.

## A. Related Works

In the context of OWC with intensity modulation and direct detection, TIM can be interpreted as a hybrid PAM and pulseposition modulation (PPM) system from the perspective of IM due to the real and positivity constraints of the transmitted signal. This fact is also stated in [17]. In order to increase the spectral efficiency, PAM and PPM are merged and proposed in [21] for optical fibre communications. After that, several works have been published on the generalization of time domain pulses. In [22], an attempt to generalize conventional PPM is made. Unlike conventional PPM, two pulses are activated in a single block. A year after, a bandwidth efficient generalization of PPM, namely multiple PPM (MPPM) is proposed in [23] for optical communications. In 2004, the combination of PAM and PPM are used for ultra wideband systems [24], which is referred to as PPAM. The constructions of signal sets of PPAM are provided in [25], which is based on Euclidean distances and in [26], which uses the set partitioning algorithm for a trellis coded modulation [27]. Multiple PPAM (MPAPM) for OWC is briefly discussed in [28]. Combinations of TIM-based systems and other transmission entities, e.g., transmit antennas, are presented in [13], [14], [29]. In [11], single-carrier index modulation (SCIM) is proposed where only time slot entity is used without combining with other entities. In this paper, the advantages and limitations of the application of TIM in OWC systems are investigated. To the best of our knowledge, a similar study has not been presented yet.

The remainder of this paper is organized as follows. A system model for the PAM-SCFDE is provided in Section II. TIM and GTIM will be discussed in Section III. The proposed construction method for GTIM is given in Section IV. In Section V, simulation and analytical results are provided. Section VI concludes this paper.

Notation: A boldface, lower case letter denotes a vector, and a bold face, upper case letter denotes a matrix. The expected value and floor operations are denoted by $\mathbb{E}[\cdot]$ and $\lfloor\cdot\rfloor$, respectively. The superscripts T or H denote a transpose and a Hermitian transpose of a vector or a matrix, respectively. The tail distribution function of a standard normal distribution is denoted by $Q(\cdot)$.

## II. System Model

## A. Transmitter and Receiver Models

A block-based transmission system with a cyclic prefix (CP) and N -point discrete Fourier transform (DFT) is used in this paper as depicted in Fig. 1. The $B_{\mathrm{t}}$-bit binary vector $\mathbf{b}$ is partitioned into $L$ sub-blocks, i.e., $\mathbf{b}=\left[\mathbf{b}_{0}^{\mathrm{T}}, \mathbf{b}_{1}^{\mathrm{T}}, \cdots, \mathbf{b}_{L-1}^{\mathrm{T}}\right]^{\mathrm{T}}$, such that $J=N / L$, where $J$ is a time slot length. The length of each vector $\mathbf{b}_{l}$ becomes, $B=B_{\mathrm{t}} / L$ bits. Then, each $\mathbf{b}_{l}$ is mapped onto a $J$-length symbol vector $\mathbf{s}_{l}=\left[s_{l, 0}, s_{l, 1}, \cdots, s_{l, J-1}\right]^{\mathrm{T}}$ by a one-to-one mapping function $f_{\mathcal{S}}$ according to a set $\mathcal{S} \in\left\{\mathcal{S}^{\text {PAM }}, \mathcal{S}^{\text {TIM }}, \mathcal{S}^{\text {GTIM }}\right\}$ which denotes the codebooks of the unipolar SC-PAM, SC-TIM and SC-GTIM, respectively. Note that throughout this paper, SC-PAM refers to PAM-SCFDE [19], SC-TIM refers to PAM-based SCIM, and the proposed


Fig. 1. A system model of PAM with block transmission and CP.

GTIM will be referred to as SC-GTIM. Finally, the transmitted symbol vector is defined as $\mathbf{s}=\left[\mathbf{s}_{0}^{\mathrm{T}}, \mathbf{s}_{1}^{\mathrm{T}}, \cdots, \mathbf{s}_{L-1}^{\mathrm{T}}\right]^{\mathrm{T}}$ whose length is $N$ symbols.

For a unipolar $M$-PAM, the elements of the symbol vectors are chosen from a set of symbols, which is defined as:

$$
\mathcal{P}_{M}=\left\{\left.I_{\mathrm{a}}+\frac{I_{\mathrm{b}}-I_{\mathrm{a}}}{M-1} i \right\rvert\, i \in\{0,1, \cdots, M-1\},\left[I_{\mathrm{a}}, I_{\mathrm{b}}\right] \subset \mathbb{R}^{+}\right\} .
$$

Note that the constellation set $\mathcal{P}_{M}$ is defined such that the linear dynamic range of an LED, which is defined as $\left[I_{\mathrm{a}}, I_{\mathrm{b}}\right]$, is incorporated. Intuitively, the symbol $s_{l, j} \in \mathcal{P}_{M}$ models the driving current of an LED. Note that $\left[I_{\mathrm{a}}, I_{\mathrm{b}}\right]$ is assumed to be strictly positive. In practice, this is achieved by adding a DC bias.

Let $\eta$ be the normalized spectral efficiency in bits per symbol, and it is defined as:

$$
\begin{equation*}
\eta \triangleq \frac{B}{J} \quad(\mathrm{bits} / \text { symbol }) \tag{1}
\end{equation*}
$$

Note that the spectral efficiency of a block transmission in $\mathrm{bps} / \mathrm{Hz}$ with PAM and the CP length of $L_{\mathrm{cp}}$ is [30, p. 226]:

$$
\begin{equation*}
R=\frac{N}{N+L_{\mathrm{cp}}} 2 \eta \quad(\mathrm{bps} / \mathrm{Hz}) \tag{2}
\end{equation*}
$$

where $N /\left(N+L_{\mathrm{cp}}\right)$ is the bandwidth penalty due to the CP.
For a block transmission with SC-PAM, the transmitted symbol vectors are chosen from the set $\mathcal{S}^{\text {PAM }}$ which is defined as:

$$
\mathcal{S}^{\mathrm{PAM}}=\left\{\mathbf{s}_{l}=\left[s_{l, 0}, s_{l, 1}, \cdots, s_{l, J-1}\right]^{\mathrm{T}} \mid s_{l, j} \in \mathcal{P}_{M}\right\},
$$

where $\left|\mathcal{S}^{\mathrm{PAM}}\right|=2^{B}$ is the cardinality of the set $\mathcal{S}^{\mathrm{PAM}}$. Typically, SC-PAM is defined with $J=1$ and $s_{l} \in \mathcal{P}_{M}$ for $l=\{0,1,2, \cdots, L-1=N-1\}$. For example, with $M=2$ and $B=1$, if $b_{l, 0}=0$, then $s_{l, 0}$ is mapped to $s_{l, 0}=I_{\mathrm{a}}$. Otherwise, $s_{l, 0}$ is mapped to $s_{l, 0}=I_{\mathrm{b}}$. Equivalently with $J=2$, the set $\mathcal{S}^{\mathrm{PAM}}$ can also be expressed as:

$$
\mathcal{S}^{\text {PAM }}=\left\{\underset{\substack{I_{\mathrm{a}}  \tag{3}\\
I_{\mathrm{a}} \\
b_{l, 0}=[0] \\
b_{l, 1}=[0] \\
b_{l, 0}=[1] \\
b_{l, 1}=[0] \\
b_{l, 0}=[0] \\
b_{l, 1}=[1] \\
I_{l} \\
b_{l, 1}=[1]}}{ }\left[\begin{array}{c}
I_{\mathrm{b}} \\
I_{\mathrm{a}}
\end{array}\right],\left[\begin{array}{c}
I_{\mathrm{a}} \\
I_{\mathrm{b}}
\end{array}\right],\left[\begin{array}{c}
I_{\mathrm{b}} \\
I_{\mathrm{b}}
\end{array}\right]\right\} .
$$

Notice that the mapping remains the same as in the case with $J=1$. This representation will be used to relate SC-PAM to SC-TIM and SC-GTIM. For SC-PAM, the Gray code is always used to ensure the optimality in the sense of error probability


Fig. 2. (a) A continuous-time channel model after the parallel-to-serial conversion and before the serial-to-parallel conversion, and (b) a discrete-time equivalent channel model.
[31]. For a unit energy transmit pulse, the energy per symbol for the SC-PAM is denoted as:

$$
\begin{align*}
E_{\mathrm{s}, \mathrm{PAM}} & =\frac{1}{\left|\mathcal{P}_{M}\right|} \sum_{s \in \mathcal{P}_{M}}|s|^{2} \\
& =\frac{1}{6}\left(2\left(I_{\mathrm{a}}^{2}+I_{\mathrm{a}} I_{\mathrm{b}}+I_{\mathrm{b}}^{2}\right)+\frac{\left(I_{\mathrm{b}}-I_{\mathrm{a}}\right)^{2}}{M-1}\right) \tag{4}
\end{align*}
$$

where the second equality is obtained from sum of series formula. The energy per bit is denoted by $E_{\mathrm{b}, \mathrm{PAM}}=E_{\mathrm{s}, \text { PAM }} / \eta$.

A discrete-time channel $c(n)$ is formed by sampling a continuous-time channel impulse response $c_{\mathrm{c}}(t)$ with a symbol rate $R_{\mathrm{s}}$. The impulse response $c_{\mathrm{c}}(t)$ is the cascaded finite impulse response (FIR) filter that models a digital-to-analog converter (DAC), $h_{\mathrm{dac}}(t)$; LED, $h_{\mathrm{led}}(t)$; a frequency-selective wireless optical channel (WOC), $h_{\mathrm{oc}}(t)$; a photodiode (PD), $h_{\mathrm{pd}}(t)$; a transimpedance amplifier (TIA); and an analog-todigital converter (ADC), $h_{\text {adc }}(t)$, see Fig. 2 Note that $x(n)$ and $y(n)$ are the discrete-time transmitted and received symbols after the parallel-to-serial conversion and before the serial-to-parallel conversion, respectively. An additive white noise $w(n)$ with zero mean and variance $\sigma_{w}^{2}$ is used in this paper. This additive white noise represents both the thermal and shot noises that can be modeled as white, Gaussian, and signalindependent noise [32, p. 267].

## B. Channel Model

As for the calculation of WOC, a frequency-domain approach proposed by Schulze [33] is used in this paper. The advantage of the frequency-domain approach is that infinite reflections can be taken into account. In addition, an arbitrarilyoriented device that is being held by a person is assumed, and a human body is modeled by a rectangular prism in the calculation of the WOC as in [34], [35].

The block-based transmission with PAM is applied for an uplink case due to a lower computational complexity and lower PAPR requirements at the transmitter side. Furthermore, it is assumed that the transmission is carried out over the infrared (IR) spectrum to avoid any visual discomfort for the users. An attenuation factor between the access point (AP) and the user


Fig. 3. A geometric model for the optical channel.
equipment (UE), which is illustrated in Fig. 3 is defined as follows:

$$
\begin{equation*}
H=\frac{m+1}{2 \pi} \cos ^{m}(\phi) \frac{A_{\mathrm{AP}} \cos (\psi)}{d^{2}} v_{\mathrm{AP}, \mathrm{UE}}(\mathrm{FoV}) \tag{5}
\end{equation*}
$$

where $m$ is the Lambertian index, $A_{\mathrm{AP}}$ denotes the AP photodetector area and $\phi$ is the radiance angle between the normal vector $\mathbf{n}_{\mathrm{u}}$ and the position of the AP. The incidence angle between the normal vector $\mathbf{n}_{\mathrm{a}}$ and the position of the UE is denoted by $\psi$. The orientations of an LED in the UE and a PD in the AP are shown by the unit normal vector $\mathbf{n}_{u}$ and $\mathbf{n}_{\mathrm{a}}$, respectively. The distance between the UE and AP is denoted by $d$. A binary function $v_{\text {AP,UE }}$ is the visibility factor of the link between an AP and UE, i.e., $v_{\mathrm{AP}, \mathrm{UE}}$ is one if $-\pi / 2 \leq \phi \leq \pi / 2$ and zero otherwise. The visibility factor is also a function of the field-of-view (FoV) of a receiver. That is, $v_{\text {AP,UE }}(\mathrm{FoV})$ is one if $0 \leq \psi \leq \mathrm{FoV} \leq \pi / 2$. The diffuse link can be calculated using a sum of products of (5) and the delay factor of reflecting surfaces, see [33] for more detailed information.

## C. Equalization

With the CP length $L_{\mathrm{cp}}$ being equal or larger than the length of the channel taps, the discrete-time channel model can be denoted as:

$$
\begin{equation*}
C(z)=\sum_{n=0}^{L_{\mathrm{cp}}} c(n) z^{-n} \tag{6}
\end{equation*}
$$

The $N \times N$ channel matrix $\mathbf{H}$ is a circulant matrix whose first column is $\left[c(0), c(1), \ldots, c\left(L_{\mathrm{cp}}\right), \mathbf{0}\right]^{\mathrm{T}}$, where $N \geq L_{\mathrm{cp}}$. A circulant matrix can be diagonalized by using a normalized DFT matrix $\mathbf{W}$ as:

$$
\mathbf{H}=\mathbf{W}^{\mathrm{H}} \boldsymbol{\Gamma} \mathbf{W},
$$

where $\Gamma$ is a $N \times N$ diagonal matrix whose diagonal entry is the $N$-point DFT of the channel tap $c(n)$. At the receiver side, the received symbol vectors $\mathbf{r}$ are equalized by a feedforward equalization matrix F. Finally, the estimated symbol vectors $\hat{\mathbf{s}}$ are further processed to obtain an estimated bit stream of $\mathbf{b}$, which is denoted with $\hat{\mathbf{b}}$, by using the inverse of the mapping function $f_{\mathcal{S}}$, namely $f_{\mathcal{S}}^{-1}$.

A minimum mean square error (MMSE)-based filter is used in this paper. For a block transmission of PAM, it is shown in [36] that the optimal filter $\mathbf{F}$ can be written as:

$$
\begin{equation*}
\mathbf{F}=\mathbf{R}_{s} \mathbf{H}^{\mathrm{H}}\left(\mathbf{H} \mathbf{R}_{s} \mathbf{H}^{\mathrm{H}}+\mathbf{R}_{\mathbf{w}}\right)^{-1} \tag{7}
\end{equation*}
$$

where $\mathbf{R}_{\mathbf{s}}$ denotes the autocorrelation matrix of $\mathbf{s}$, i.e., $\mathbf{R}_{\mathbf{s}} \triangleq$ $\mathbb{E}\left[\mathbf{s s}^{\mathrm{H}}\right]$. The same definition also applies to the autocorrelation matrix of the noise vector $\mathbf{w}$, i.e., $\mathbf{R}_{\mathbf{w}}=\mathbb{E}\left[\mathbf{w w}^{\mathrm{H}}\right]=\sigma_{w}^{2} \mathbf{I}$. A brute-force search is used to estimate the transmitted symbol, i.e.,:

$$
\begin{equation*}
\hat{\mathbf{s}}_{l}=\underset{\mathbf{s}_{l} \in \mathcal{S}^{\mathrm{TIM}}}{\arg \min }\left\|\hat{\mathbf{s}}_{l}-\mathbf{s}_{l}\right\| . \tag{8}
\end{equation*}
$$

Then, $\mathbf{b}_{l}$ is estimated by $\hat{\mathbf{b}}_{l}=f_{\mathcal{S}^{\mathrm{TIM}}}^{-1}\left(\hat{\mathbf{s}}_{l}\right)$.

## III. Time Slot Index Modulation

For readability purposes, TIM for SC-FDE will be first explained based on [11]. Then, it is followed by the generalization of the SC-TIM, namely SC-GTIM. To apply an IM to SC-PAM, an additional symbol is required to denote that a transmission entity is inactive. In our case, that transmission entity is a time slot. The additional symbol is denoted as $I_{0}$. Therefore, by applying IM to SC-PAM, a symbol $s_{l, j}$ is now $s_{l, j} \in\left\{\mathcal{P}_{M},\left\{I_{0}\right\}\right\}$. The same argument is also used in [37, p. 736]. In the IM literature [16], [17], [37], $I_{0}$ is typically chosen to be 0 . In this paper, we assume a general case where $I_{0} \in\left\{0, I_{\mathrm{a}}-\epsilon\right\}$, for a small positive value $\epsilon$. A reason to choose $I_{0}=I_{\mathrm{a}}-\epsilon$ is to minimize the effect of the rapid switching problem, which decreases the lifetime of an LED, if $I_{0}$ is chosen to be 0 .

## A. Single-Carrier Time Slot Index Modulation (SC-TIM)

In SC-TIM, the $B$-bit vector $\mathbf{b}_{l}$ contains a $B_{1}$-bit and a $B_{2}$ bit binary vectors, i.e., $B=B_{1}+B_{2}$. A $B_{1}$-bit binary vector is used to modulate the index of the active time slots. Note that $B_{1}$ is defined such that $K$ out of $J$ are chosen from the set $\mathcal{P}_{M}$, and the others are set to be $I_{0}$. It is also worth noting that $K$ out of $J$ can be interpreted as choosing $K$ out of $J$ time slots to be active. A $B_{2}$-bit binary vector is used to modulate the $M$-PAM symbols from $\mathcal{P}_{M}$. Therefore, $B_{1}$ and $B_{2}$ can be calculated as:

$$
\begin{equation*}
B_{1}=\left\lfloor\log _{2}\binom{J}{K}\right\rfloor, \text { and } B_{2}=K \log _{2} M \tag{9}
\end{equation*}
$$

For example, with $K=1, J=2$ and $\mathcal{P}_{2}=\left\{I_{\mathrm{a}}, I_{\mathrm{b}}\right\}$, we have the possible transmitted symbol $\mathbf{s}_{l} \in \mathcal{S}^{\mathrm{TIM}}$, such that the mapping set $\mathcal{S}^{\text {TIM }}$ is:

$$
\mathcal{S}^{\mathrm{TIM}}=\left\{\begin{array}{c}
{\left[\begin{array}{c}
I_{0} \\
I_{\mathrm{a}}
\end{array}\right],\left[\begin{array}{c}
I_{0} \\
I_{\mathrm{b}}
\end{array}\right],\left[\begin{array}{c}
I_{\mathrm{a}} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
I_{\mathrm{b}} \\
I_{\mathrm{b}}
\end{array}\right]}  \tag{10}\\
\mathrm{I}_{l, 0}=[0] \\
b_{l, 1}=[0] \\
b_{l, 0}=[1]=[0] \\
b_{l, 0}=[0] \\
b_{l, 1}=[1] \\
b_{l, 0}=[1] \\
b_{l, 1}=[1]
\end{array}\right\} .
$$

In this case, $b_{l, 1}=0$ denotes the first time slot as inactive, while $b_{l, 1}=1$ denotes the second time slot as inactive. In addition, $b_{l, 0}=0$ denotes that the PAM constellation symbol $I_{\mathrm{a}}$ is chosen. Otherwise, the PAM constellation symbol $I_{\mathrm{b}}$


Fig. 4. The working principle of $f_{\mathcal{S}^{\text {SCTIM }}}$ with 10 .
is chosen. A working principle of (10) is shown in Fig. 4 Following the notation in Section II, suppose the binary input vector is $\mathbf{b}=[1,0,1,1,0,1,0,0,1,0]^{\mathrm{T}}$ and $L=5$, then we have $\mathbf{b}_{0}=\left[b_{0,0}, b_{0,1}\right]^{\mathrm{T}}=[1,0]^{\mathrm{T}}, \mathbf{b}_{1}=\left[b_{1,0}, b_{1,1}\right]^{\mathrm{T}}=[1,1]^{\mathrm{T}}$, etc. Based on $(10),[1,0]^{\mathrm{T}}$ maps to $\left[I_{0}, I_{\mathrm{b}}\right]^{\mathrm{T}}$, and $[1,1]^{\mathrm{T}}$ maps to $\left[I_{\mathrm{b}}, I_{0}\right]^{\mathrm{T}}$ consecutively.

Compared to the energy per symbol of the SC-PAM in (4), it is generally harder to obtain a compact expression for $\mathbf{R}_{\mathbf{s}}$. The naive way is (i) to list all possible combinations of alphabets $\mathbf{s s}^{\mathrm{H}}$; (ii) count the frequency of each element of the alphabet for given parameters $M, K$ and $J$; and (iii) calculate $\mathbf{R}_{\mathbf{s}}=\mathbb{E}\left[\mathbf{s s}^{\mathrm{H}}\right]$ by taking the expectation value of all possible combinations. However, for $K=1$ and $I_{0}=0, \mathbf{R}_{\mathbf{s}}{ }^{\mathrm{TIM}, K=1}$ can be found as shown in (11). The constants $p$ and $q$ can be calculated as:

$$
\begin{aligned}
& p=\frac{1}{2^{B}} \sum_{s \in \mathcal{P}_{M}}|s|^{2}=\frac{\left(\left(I_{\mathrm{a}}^{2}+I_{\mathrm{b}}^{2}\right)(2 M-1)+2 I_{\mathrm{a}} I_{\mathrm{b}}(M-2)\right)}{3\left(1-\frac{1}{M}\right) 2^{(B+1)}} \\
& q=\frac{1}{2^{J} 2^{M}} \sum_{x \in \mathcal{P}_{M}} \sum_{y \in \mathcal{P}_{M}} x y=M^{2}\left(I_{\mathrm{a}}+I_{\mathrm{b}}\right)^{2} 2^{-(J+M+2)},
\end{aligned}
$$

where sum of series formula is used to obtain these expressions. It is worth noting that $\mathbf{R}_{\mathbf{s}}{ }^{\mathrm{TIM}, K=1}$ is not a circulant matrix. This is an example of a scenario when the simple single-tap equalization cannot always be performed by a PAMbased SC-TIM. In this case, the PAM-based SC-TIM has a higher computational complexity.

Note that $p$ is also the average energy per symbol for the PAM-based SC-TIM which is lower than that without TIM, i.e., $E_{\mathrm{s}, \mathrm{TIM}}=p \leq E_{\mathrm{s}, \text { PAM }}$. Typically, IM has lower transmitted energy due to the value of the additional symbol, i.e., $I_{0}$, that is usually 0 . However, in terms of PAPR, the PAM-based SCTIM has a larger PAPR as the peak value of $\mathbf{s}$ is still $I_{\mathrm{b}}$, while the average energy is lower. It is worth noting that the increase in PAPR does not occur for other IM systems such as OFDMIM, where the PAPR is the same as the conventional OFDM system, see Fig. 4 in [17].

For an IM scheme, an average pairwise-error probability (PEP) over all possible pairs of symbols is typically used as an upper bound. The average BER is upper bounded by the average PEP (APEP), that is,:

$$
\begin{equation*}
P_{\mathrm{b}} \leq \frac{1}{B 2^{B}} \sum_{\mathbf{s}_{i} \in \mathcal{S}^{\mathrm{TIM}}} \sum_{\substack{\mathbf{s}_{j} \in \mathcal{S}^{\mathrm{TIM}} \\ \mathbf{s}_{j} \neq \mathbf{s}_{i}}} d_{\mathrm{H}}\left(\mathbf{s}_{i}, \mathbf{s}_{j}\right) P\left(\mathbf{s}_{i} \rightarrow \mathbf{s}_{j}\right)=\text { APEP. } \tag{12}
\end{equation*}
$$

A Hamming distance between the bits associated with both $\mathbf{s}_{i}$ and $\mathbf{s}_{j}$ is denoted by $d_{\mathrm{H}}\left(\mathbf{s}_{i}, \mathbf{s}_{j}\right)$. The PEP is defined as:

$$
\begin{equation*}
P\left(\mathbf{s}_{i} \rightarrow \mathbf{s}_{j}\right)=Q\left(\frac{d_{\mathrm{E}}\left(\mathbf{s}_{i}, \mathbf{s}_{j}\right)}{2 \sqrt{\sigma_{\mathbf{e}}^{2}}}\right) \tag{13}
\end{equation*}
$$

where $d_{\mathrm{E}}\left(\mathbf{s}_{i}, \mathbf{s}_{j}\right)=\left\|\mathbf{s}_{i}-\mathbf{s}_{j}\right\|$ denotes the Euclidean distance between $\mathbf{s}_{i}$ and $\mathbf{s}_{j}$. The autocorrelation matrix of the error vector $\mathbf{e}=\hat{\mathbf{s}}-\mathbf{s}$ is defined as:

$$
\begin{equation*}
\mathbf{R}_{\mathbf{e}} \triangleq \mathrm{E}\left\{\mathbf{e e}^{\mathrm{H}}\right\}=\left(\mathbf{R}_{\mathbf{s}}^{-1}+\mathbf{H}^{\mathrm{H}} \mathbf{R}_{\mathbf{w}}^{-1} \mathbf{H}\right)^{-1} \tag{14}
\end{equation*}
$$

where the Woodbury identity is used, see [36]. Let the average noise power of $\mathbf{e}$ be $\sigma_{\mathbf{e}}^{2}$, which is defined as:

$$
\sigma_{\mathbf{e}}^{2} \triangleq \operatorname{tr}\left(\mathbf{R}_{\mathbf{e}}\right) / N
$$

For the SC-PAM, instead of using the upper bound in (12), a good approximation can be obtained as follows:

$$
\begin{align*}
& P_{\mathrm{b}, \operatorname{PAM}} \approx \frac{1}{N} \sum_{i=0}^{N-1} \frac{2}{\log _{2} M}\left(1-\frac{1}{M}\right) Q\left(f\left(I_{\mathrm{a}}, I_{\mathrm{b}}, \beta_{i}\right)\right),  \tag{15}\\
& f\left(I_{\mathrm{a}}, I_{\mathrm{b}}, \beta_{i}\right)=\sqrt{\frac{1.5\left(I_{\mathrm{b}}-I_{\mathrm{a}}\right)^{2} \log _{2}(M) \beta_{i}}{(M-1)^{2}\left(2\left(I_{\mathrm{a}}^{2}+I_{\mathrm{a}} I_{\mathrm{b}}+I_{\mathrm{b}}^{2}\right)+\frac{\left(I_{\mathrm{b}}-I_{\mathrm{a}}\right)^{2}}{M-1}\right)}},
\end{align*}
$$

where the Gray code is assumed and $\beta_{i}=E_{\mathrm{b}, \operatorname{PAM}} /\left[\mathbf{R}_{\mathbf{e}}\right]_{i}$. It is assumed in (15) that the error vector $\mathbf{e}$ coming from both the noise $\mathbf{w}$ and the remaining intersymbol interference (ISI) is approximated by a Gaussian distribution [38].

## B. Single-Carrier Generalized Time Slot Index Modulation (SC-GTIM)

The main difference between SC-TIM and SC-GTIM is that $K$ does not have to be fixed in SC-GTIM. To compare it with (10), we have, for example,:

(a) SC-PAM (3)

(b) SC-TIM 10

(c) SC-GTIM 16

Fig. 5. Vector representations: (a) SC-PAM with $M=2$ and the Gray code and without IM $(\eta=1)$, (b) SC-TIM with $K=1$ and $J=2(\eta=1)$ and (c) SC-GTIM with $J=2(\eta=1)$. The circle markers show the symbols that can be chosen and labeled with binary vectors. The other markers show the chosen symbols.
where the number of bits in SC-GTIM is defined as $B=$ $\log _{2}\left|\mathcal{S}^{\text {GTIM }}\right|$. Notice that $K$ is not fixed for all elements of $\mathcal{S}^{\text {GTIM }}$ in the above example. In addition, SC-GTIM in 16 reduces to on-off keying (OOK) with return-to-zero pulse or PPM, which is known to achieve the Shannon capacity in low spectral efficiency regions [39]. It is worth mentioning that SCTIM cannot be reduced to OOK as SC-TIM is constrained by (9), which shows less flexibility compared to SC-GTIM.

To illustrate the set $\mathcal{S}^{\text {GTIM }}$ based on 16 and compare it with other examples in (3) and (10), the vector representations of $\left[s_{l, 0}, s_{l, 1}, \cdots, s_{l, J-1}\right]^{\mathrm{T}}$ are illustrated in Fig. 5. The point $o_{I_{0}}$ is the shifted origin with respect to $I_{0}$. e.g., in $\mathbb{R}^{2}, o_{I_{0}}$ is $\left(I_{0}, I_{0}\right)$. The index $l$ is omitted as it is not relevant. Note that a directed vector that starts from $o_{I_{0}}$ is used. Fig. 5] uses circles to show all possible $\mathbf{s}_{l}$ that can be chosen and the directed vectors with markers to show the chosen $\mathbf{s}_{l}$. In this way, we have a classic constellation diagram representation. Fig. 5(c) clearly illustrates a more flexible constellation design of SCGTIM compared to the others. If this can be cultivated fully, a significant performance gain can be achieved.

We now compare SC-GTIM with SC-TIM in (10). It is obvious to see that the BER of SC-TIM with $\mathcal{S}^{\text {TIM }}$ expressed in (10) will be limited by the vectors $\left[I_{0}, I_{\mathrm{a}}\right]^{\mathrm{T}}$ and $\left[I_{\mathrm{a}}, I_{0}\right]^{\mathrm{T}}$ if $\left|I_{\mathrm{a}}-I_{0}\right|<\left|I_{\mathrm{b}}-I_{\mathrm{a}}\right|$. Consequently, the BER of SC-TIM is worse than that of the conventional SC-PAM. In this case, a codebook
that has more possible constellation points is required. This is where SC-GTIM has the advantage by means of freeing the number of active time slots.

A collection of possible codewords $\mathcal{S}^{\mathrm{s}}$ is defined as:

$$
\begin{equation*}
\mathcal{S}^{\mathrm{S}}=\left\{\mathbf{s}_{l}=\left[s_{l, 0}, s_{l, 1}, \cdots, s_{l, J-1}\right]^{\mathrm{T}} \mid s_{l, j} \in\left\{\mathcal{P}_{M},\left\{I_{0}\right\}\right\}\right\} . \tag{17}
\end{equation*}
$$

The indexing and grouping of all possible $\left[s_{l, 0}, s_{l, 1}, \cdots, s_{l, J-1}\right]^{\mathrm{T}} \in \mathcal{S}^{\mathrm{GTIM}} \subseteq \mathcal{S}^{\mathrm{s}}$ will affect the BER. There are:

$$
\begin{equation*}
G=\frac{(M+1)^{J}!}{\left((M+1)^{J}-2^{B}\right)!} \tag{18}
\end{equation*}
$$

ways of indexing and grouping in the construction of the set $\mathcal{S}^{\text {GTIM }}$. Based on (13) and the upper bound in (12), the design problem can be formulated as:

$$
\begin{equation*}
\underset{\mathcal{S}^{\text {GTIM }}}{\arg \max } \quad C=\sum_{\mathbf{s}_{i} \in \mathcal{S}^{\text {GTIM }}} \sum_{\substack{\mathbf{s}_{j} \in \mathcal{S}^{\text {GTIM }} \\ \mathbf{s}_{j} \neq \mathbf{s}_{i}}} d_{\mathrm{H}}\left(\mathbf{s}_{i}, \mathbf{s}_{j}\right) d_{\mathrm{E}}\left(\mathbf{s}_{i}, \mathbf{s}_{j}\right) \tag{19}
\end{equation*}
$$

subject to: $\quad\left|\mathcal{S}^{\text {GTIM }}\right|=2^{B}$.
The problem in (19) can be seen as the problem of choosing $2^{B}$ active vectors in $\left[s_{l, 0}, s_{l, 1}, \cdots, s_{l, J-1}\right]^{\mathrm{T}}$ and labeling them with a $B$-bit vector. To be more specific, the goal is to label a vector such that it has both large Hamming and Euclidean distances. As 19 is a combinatoric problem which has $G$ many possible combinations, a meta-heuristic solution for a permutation-based combinatorial optimization problem can be used. However, it will be shown in the next section that a simpler way can be applied by using the knowledge of the Gray encoding.

## C. Spectral Efficiency Comparison for Different Time Slot Length J

In this subsection, the normalized spectral efficiencies in bits per symbol of the conventional SC-PAM, SC-TIM and SC-GTIM are compared, see (2) for a conversion to the unit $\mathrm{bps} / \mathrm{Hz}$. Specifically, we are interested in the maximum achievable $\eta$ given $M \geq 2$ and different $J$. Notice that it will be straightforward for both SC-PAM and SC-GTIM, but it is not the case for SC-TIM. For SC-TIM, the normalized spectral efficiency is defined as:

$$
\begin{aligned}
\eta_{\max }^{\mathrm{TIM}} & =\max _{1 \leq K \leq J} \frac{\left\lfloor\log _{2}\binom{J}{K}\right\rfloor+K \log _{2} M}{J} \text { and } \\
K^{*} & =\underset{1 \leq K \leq J}{\inf } \underset{\arg \max }{ } \frac{\left\lfloor\log _{2}\binom{J}{K}\right\rfloor+K \log _{2} M}{J}
\end{aligned}
$$

where inf denotes the infimum of a set. The infimum ensures that $K^{*}$ is not $J$ if there is another value except $J$. The closed form solutions of the above problem formulations are difficult to obtain due to the linear combination of the terms $\left[\log _{2}\binom{J}{K}\right]$ and $K \log _{2} M \square^{1}$ Therefore, $\eta_{\max }^{\mathrm{TIM}}$ and $K^{*}$ will be calculated numerically.

[^1]

Fig. 6. (a) Maximum achievable $\eta$ for SC-TIM, $\eta_{\max }^{\text {TIM }}$, and (b) $K$ optimal to achieve $\eta_{\text {max }}^{\text {TIM }}, K^{*}$.

As for the conventional SC-PAM, it can be readily obtained as $\eta_{\text {max }}^{\mathrm{PAM}}=\log _{2} M$. For SC-GTIM, given $M$ and $J$, there are $(M+1)^{J}$ possible choices of vector $\mathbf{s}_{l}$, see (17). Therefore, the number of bits that can be transmitted for SC-GTIM is upper bounded by:

$$
2^{B} \leq(M+1)^{J}
$$

Using the definition of $\eta$ in (1) and the fact that $B$ must be an integer, we have:

$$
\eta_{\max }^{\mathrm{GTIM}}=\left\lfloor J \log _{2}(M+1)\right\rfloor / J
$$

Hence, $\eta_{\max }^{\text {PAM }} \leq \eta_{\max }^{\text {GTIM }}$. The numerical results for $\eta_{\max }$ of all systems are shown in Fig. 6(a). Therefore, we have $\eta_{\max }^{\mathrm{PAM}} \leq \eta_{\max }^{\mathrm{TIM}} \leq \eta_{\max }^{\mathrm{GTIM}}$. This shows the advantages of SCGTIM in terms of spectral efficiency.

It is also interesting to discuss $K^{*}$ as shown in Fig. 6(b). Based on 99, SC-TIM with $K=J$ reduces to SC-PAM. It means that when $K^{*}=J$, then SC-TIM cannot have a better spectral efficiency than that of SC-PAM. This illustrates the disadvantage of having a fixed number of active time slots.

## IV. Construction of A Good $\operatorname{Set} \mathcal{S}^{\text {GTim }}$

In this section, the construction of a good set $\mathcal{S}^{\text {GTIM }}$ is provided. We start with a simple example in two dimensions (2-D), i.e., $J=2$, and this is followed by an introduction of several necessary notations. Then, a more detailed algorithm is given next with more relevant examples having higher dimensions, i.e., $J>2$.

## A. Simple Example in 2-D $(J=2)$

For a better enumeration and illustration, $\mathcal{P}_{M}$ is redefined as:
$\mathcal{P}_{M}=\left\{m_{0}, m_{1}, \cdots, m_{M-1} \mid I_{\mathrm{a}}=m_{0}<m_{1}<\cdots<m_{M-1}=I_{\mathrm{b}}\right\}$,

Now, assume that a uniform $M$-PAM constellation is used, i.e., $\left|m_{i}-m_{j}\right|=c, \forall i \neq j \in\{0,1, \cdots, M-1\}$ and a positive constant $c$. In this example, we use the following collection of possible codewords:

$$
\mathcal{S}^{\mathrm{s}}=\left\{\left[\begin{array}{c}
I_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
m_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
m_{1} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
I_{0} \\
m_{0}
\end{array}\right],\left[\begin{array}{c}
m_{0} \\
m_{0}
\end{array}\right],\left[\begin{array}{c}
m_{1} \\
m_{0}
\end{array}\right],\left[\begin{array}{c}
I_{0} \\
m_{1}
\end{array}\right],\left[\begin{array}{l}
m_{0} \\
m_{1}
\end{array}\right],\left[\begin{array}{l}
m_{1} \\
m_{1}
\end{array}\right]\right\}
$$

which represents the case where $M=2$ and $J=2$. Suppose that we define $B=3$ bits, then we should pick $2^{3}=8$ of $\left|\mathcal{S}^{\text {s }}\right|=$ 9. Now that we have defined $\mathcal{S}^{\mathrm{s}}$, our design problem can be summarized as follows:
given: $M, B, J$
choose $2^{B}$ of $\mathbf{s}_{l} \in \mathcal{S}^{\mathrm{s}}$ and label them with binary vector
such that the $2^{B}$ chosen $\mathbf{s}_{l}$ have as large $C$ (see 19p) as possible.

The basic operation of the construction algorithm is the set partitioning algorithm denoted by SP . The following are simple illustrations of the heuristic steps taken in the SP operation. Now, we discuss them in a high level. In the next subsection, these steps will be formalized and detailed.

1) The first step: Choose 2 sets of $\left\lfloor(M+1)^{J} / 2\right\rfloor=4$ vectors of $\mathbf{s}_{l}$ such that the Euclidean distance of the average or center of mass of two sets are maximum. Then, label each set with a binary value and denote the sets with $\mathbb{P}_{b_{0}}^{M=2, \tilde{J}=2}$, where $b_{0} \in\{0,1\}$ and $\tilde{J}$ is the number of elements of $\mathbf{s}_{l}$ which are taken into account. In this case, all elements of $\mathbf{s}_{l}$ are taken into account at once as $\tilde{J}=J$. A sample of the result is illustrated in Fig. 7(a). The chosen symbol vectors are shown by squares, and the cross marker shows the unchosen symbol vector. It is worth noting here that the elements in the set $\mathbb{P}_{b_{0}}^{2,2}$ will be labeled as $\left[b_{0} X X\right]$, where $X \in\{0,1\}$ denotes a binary digit that is not assigned yet. That is, elements in $\mathbb{P}_{1}^{2,2}$ will be labeled as [100], [101], [110] or [111] in the final step.
2) The second step: Partition each $\mathbb{P}_{b_{0}}^{2,2}$ as in the first step. The outputs are now denoted as $\mathbb{P}_{b_{0}, b_{1}}^{2,2}$, where $b_{0}$ is followed by $b_{1} \in\{0,1\}$ in the subscripts. Now, each set $\mathbb{P}_{b_{0}, b_{1}}^{2,2}$ contains two vectors. The binary mapping of $\mathbb{P}_{b_{0}, b_{1}}^{2,2}$ is carried out such that it forms the Gray mapping with its neighbors. For example:

$$
\mathbb{P}_{1,0}^{2,2}=\left\{\left[\begin{array}{c}
I_{0} \\
m_{0}
\end{array}\right],\left[\begin{array}{c}
I_{0} \\
m_{1}
\end{array}\right]\right\} \text { and } \mathbb{P}_{1,1}^{2,2}=\left\{\left[\begin{array}{l}
m_{0} \\
m_{1}
\end{array}\right],\left[\begin{array}{l}
m_{1} \\
m_{1}
\end{array}\right]\right\}
$$

are neighbors to each other, and [10] and [11] forms the Gray mapping. A complete mapping is shown in Fig. 7(b).
3) The third step: As only 1 remaining bit needs to be assigned, the final step is to make sure that each element forms the Gray mapping with its neighbors. As depicted in Fig. 7(c), the final chosen constellation symbols and their mappings are denoted as:

$$
\tilde{\mathbb{P}}^{2,2}=\left(\underset{000}{\left(\begin{array}{c}
I_{0}  \tag{20}\\
I_{0}
\end{array}\right]} \underset{001}{\left[\begin{array}{c}
m_{0} \\
I_{0}
\end{array}\right]}, \underset{010}{\left[\begin{array}{c}
m_{1} \\
m_{0}
\end{array}\right]}, \underset{011}{\left[\begin{array}{c}
m_{1} \\
I_{0}
\end{array}\right]}, \underset{100}{\left[\begin{array}{c}
I_{0} \\
m_{0}
\end{array}\right]}, \underset{101}{\left[\begin{array}{c}
I_{0} \\
m_{1}
\end{array}\right]}, \underset{110}{\left[\begin{array}{c}
m_{1} \\
m_{1}
\end{array}\right]}, \underset{111}{\left[\begin{array}{c}
m_{0} \\
m_{1}
\end{array}\right]}\right)
$$

which can be interpreted that each vector corresponds to the binary vectors written below it.

## B. Notations

The main goal of this subsection is to generalize the previous notations before formalizing the proposed algorithm for all $J$. We start by generalizing the notation $\mathbb{P}$. A set of $\tilde{J}$-length vectors whose elements are taken from $\left\{\mathcal{P}_{M},\left\{I_{0}\right\}\right\}$ is defined as:

$$
\mathbb{P}^{M, \tilde{J}}=\left\{\mathbf{p}=\left[p_{0}, p_{1}, \cdots, p_{\tilde{J}-1}\right]^{\mathrm{T}}\right\}, \text { where } p_{j} \in\left\{\mathcal{P}_{M},\left\{I_{0}\right\}\right\}
$$

The set $\mathbb{P}^{M, \tilde{J}}$ can be partitioned into $\mathbb{P}^{M, \tilde{J}}=\mathbb{P}_{x_{B-1}}^{M, \tilde{J}} \cup \mathbb{P}_{y_{B-1}}^{M, \tilde{J}} \cup$ $\mathbb{P}_{\mathrm{dc}}^{M, \tilde{J}}$, where $x_{B-1} \neq y_{B-1} \in\{0,1\}$ and they are disjointed from each other. The set $\mathbb{P}_{\mathrm{dc}}^{M, \tilde{J}}$ is a set that is not processed for the next step. With respect to our previous example, this set is labeled as a cross marker in Fig. 7(a). Now, a mean or 'center of mass' vector of a subset $\mathbb{P}_{x_{B-1}}^{M, J}$ is defined as:

$$
\begin{equation*}
\mathbf{d}_{\mathrm{S}}\left(\mathbb{P}_{x_{B-1}}^{M, \tilde{J}}\right)=\frac{1}{\left|\mathbb{P}_{x_{B-1}}^{M, \tilde{J}}\right|} \sum_{\mathbf{p} \in \mathbb{P}_{x_{B-1}}^{M, \tilde{J}}} \mathbf{p} \tag{21}
\end{equation*}
$$

which is used to measure the distance between two subsets of $\mathbb{P}^{M, \tilde{J}}$. The subset $\mathbb{P}_{x_{B}, \tilde{J}}^{M, \tilde{J}}$ can be further partitioned into $\mathbb{P}_{x_{B-1}}^{M, \tilde{J}} \supset \mathbb{P}_{x_{B-1}, x_{B-2}}^{M, \tilde{J}} \cup \mathbb{P}_{x_{B-1}, y_{B-2}}^{M_{B}, \tilde{J}}$. This partition can be done until the array $\left[x_{B-1}, x_{B-2}, \cdots, x_{0}\right]$ is obtained. The subset of $\mathbb{P}^{M, \tilde{J}}$ has a binary vector associated with it, for example with $B=3, \mathbb{P}_{x_{B-1}, x_{B-2}, x_{B-3}}^{M, \tilde{J}}$ associates with a binary vector $\left[x_{B-1}, x_{B-2}, x_{B-3}\right]$.

An ordered set $\mathcal{B}_{B}$ is a collection of $B$-bit binary vectors that are arranged based on the Gray code, i.e.,:

$$
\begin{equation*}
\mathcal{B}_{B}=\left(\tilde{\mathbf{b}}_{0}, \tilde{\mathbf{b}}_{1}, \cdots, \tilde{\mathbf{b}}_{2^{B}-1}\right) \tag{22}
\end{equation*}
$$

where $\tilde{\mathbf{b}}_{i}=\left[\tilde{b}_{i, 0}, \tilde{b}_{i, 1}, \cdots, \tilde{b}_{i, B-1}\right]^{\mathrm{T}}, \tilde{b}_{i, j} \in\{0,1\}, \tilde{b}_{i, 0}$ and $\tilde{b}_{i, B-1}$ are the least significant bit (LSB) and the most significant bit (MSB), respectively. A common way to construct $\mathcal{B}_{B}$ is the iterative and mirroring method from $\mathcal{B}_{B-1}$ 31]. It follows that $\left[\tilde{b}_{i, 0}, \tilde{b}_{i, 1}, \cdots, \tilde{b}_{i, B-2}\right]^{\mathrm{T}}=$ $\left[\tilde{b}_{2^{B}-1-i, 0}, \tilde{b}_{2^{B}-1-i, 1}, \cdots, \tilde{b}_{2^{B}-1-i, B-2}\right]^{\mathrm{T}}$ for $0 \leq i \leq\left\lfloor\frac{2^{B}-1}{2}\right\rfloor$, see following examples:

$$
\begin{aligned}
& \mathcal{B}_{2}=\left(\left[\begin{array}{l}
0 \\
0
\end{array}\right],\left[\begin{array}{l}
1 \\
0
\end{array}\right],\left[\begin{array}{l}
1 \\
1
\end{array}\right],\left[\begin{array}{l}
0 \\
1
\end{array}\right]\right), \text { or } \\
& \mathcal{B}_{3}=\left(\left[\begin{array}{l}
0 \\
0 \\
0
\end{array}\right],\left[\begin{array}{l}
1 \\
0 \\
0
\end{array}\right],\left[\begin{array}{l}
1 \\
1 \\
0
\end{array}\right],\left[\begin{array}{l}
0 \\
1 \\
0
\end{array}\right],\left[\begin{array}{l}
0 \\
1 \\
1
\end{array}\right],\left[\begin{array}{l}
1 \\
1 \\
1
\end{array}\right],\left[\begin{array}{l}
1 \\
0 \\
1
\end{array}\right],\left[\begin{array}{l}
0 \\
0 \\
1
\end{array}\right]\right) .
\end{aligned}
$$

This inspires us to do the construction of the set $\mathcal{S}^{\text {GTIM }}$ by means of partitioning the set $\mathcal{B}_{B}^{2} \square^{2}$ That is, the MSB can be assigned first by following the Gray code without worrying about the other bits. In fact, it can be shown that the bit mapping as shown in Fig. 7 follows the iterative method discussed in [31].

Generally for $J>2$, we need a container set denoting the elements that have been chosen and labeled. Let the set $\mathcal{S}$ be

[^2]

Fig. 7. (a) The first, (b) the second and (c) the third steps of the set partitioning algorithm.
an ordered set whose elements are indexed by the binary digit $\mathbf{b}_{l}$. As an example, see (3), where $\mathbf{b}_{l}=\left[b_{l, 0}, b_{l, 1}\right]^{\mathrm{T}}=[0,0]^{\mathrm{T}}$ indexes the $0^{\text {th }}$ element, and $\left[b_{l, 0}, b_{l, 1}\right]^{\mathrm{T}}=[1,0]^{\mathrm{T}}$ indexes the $1^{\text {st }}$ element. An ordered set $\mathbb{S}_{J-j}^{M}$ is defined as:
$\mathbb{S}_{J-j}^{M}=\left(\left[\begin{array}{c}\mathbf{x} \\ s_{J-j} \\ \vdots \\ s_{J-2} \\ s_{J-1}\end{array}\right]\right), s_{J-j} \in\left\{\mathcal{P}_{M},\left\{I_{0}\right\}\right\}$, for $j \in\{0,1, \cdots, J-1\}$, where $\mathbf{x}$ is the $(J-j)$-length column vector denoting the elements that have not been defined yet. In other words, $\mathbb{S}_{J-j}^{M}$ defines an ordered set where the $(J-j)^{\text {th }}$ to the $(J-1)^{\text {th }}$ elements have been ordered and labeled. For example with $M=3, B=6, J=3, j=1$ and:

$$
\mathbb{S}_{3-1}^{3}=\left(\underset{0}{\left[\begin{array}{l}
\mathbf{x}  \tag{23}\\
I_{0}
\end{array}\right],} \underset{00 X X X X}{\left[\begin{array}{c}
\mathbf{x} \\
m_{0}
\end{array}\right],} \underset{01 X X X X}{\left[\begin{array}{c}
\mathbf{x} \\
m_{2}
\end{array}\right],} \underset{10 X X X X}{\left[\begin{array}{c}
\mathbf{x} \\
m_{1}
\end{array}\right]}\right)
$$

means that 2-length column vector $\mathbf{x}$ has not been defined; moreover, the $2^{\text {nd }}$ to the $3^{\text {rd }}$ elements have been ordered and labeled accordingly. The set $\mathcal{S}$ will be assigned as $\mathcal{S}=\mathbb{S}_{0}^{M}$ if all elements have been ordered and labeled. We use a convention that if no element has been assigned, then it is denoted by $\mathbb{S}_{J}^{M}$.

## C. The Proposed Set Partitioning Algorithm

A complete pseudocode for the proposed set partitioning algorithm is given in algorithm 1 for a given $M, B$ and $J$ (or $M, \eta$ and $J$ ). The main difference with the previous simple example is that for $J>2$, the operation SP can be invoked multiple times. In addition, $\tilde{J}$ of $J$ elements can be processed at once. The advantage of doing this is that it is more modular and more computationally efficient as the previously-defined mapping set can be reused.

In line $3, J_{\mathrm{b}}$ denotes a parameter to control the number of steps such that they are upper bounded by $J-J_{\mathrm{b}}+1$. The statement $J_{\mathrm{b}} \eta=\left\lfloor J_{\mathrm{b}} \eta\right\rfloor$ is used to avoid rounding the number of allocated bits to label $\mathbb{S}_{0}^{M}$. The variable $J_{\mathrm{it}}$ is used as a step size for the loop process in line 6 . In line $6(\mathrm{~h}), B_{\text {sp }}$ denotes the number of bits required for labeling the set $\mathbb{S}_{J-(j-\tilde{J})}^{M}$ during the partitioning and labeling process, which is named as $\operatorname{SP}(\cdot)$ in line $6(\mathrm{i})$. Specifically, $\mathrm{SP}\left(\tilde{J}, B_{\mathrm{sp}}, \mathbb{S}_{J-(j-\tilde{J})}^{M}\right)$ reads so that the $\tilde{J}$

```
Algorithm 1 The proposed set partitioning algorithm param-
eterized by \(\left(M, B, J, J_{\mathrm{b}}\right)\)
    1. Set \(M, B\) and \(J\).
    2. Calculate \(\eta=B / J\).
    3. Set \(J_{\mathrm{b}}\) such that \(J_{\mathrm{b}} \eta=\left\lfloor J_{\mathrm{b}} \eta\right\rfloor\) and \(J_{\mathrm{b}} \in\{1,2, \cdots, J\}\).
    4. Calculate \(J_{\text {it }}=\inf (\{k \mid k \in\{1,2, \cdots, J\}, k \eta=\lfloor k \eta\rfloor\})\).
    5. Initialize \(\mathbb{S}_{J}^{M}\) and assign \(j=0\).
    6.(a) do
    (b) if \(J-j \neq J_{\mathrm{b}}\)
    (c) \(\tilde{J}=J_{\mathrm{it}}\);
    (d) else
        (e) \(\tilde{J}=J_{\mathrm{b}}\);
        (f) end
        (g) \(j=j+\tilde{J}\);
        (h) \(\quad B_{\mathrm{sp}}=\tilde{J} \eta\);
        \(\mathbb{S}_{J-j}^{M}=\operatorname{SP}\left(\tilde{J}, B_{\mathrm{sp}}, \mathbb{S}_{J-(j-\tilde{J})}^{M}\right) ; / /\) partition \(\mathbb{S}_{J-(j-\tilde{J})}^{M}\)
        (j) while \(j<J\)
    7. Output \(\mathcal{S}^{\text {GTIM }}=\mathbb{S}_{0}^{M}\).
```

elements of the vector in the set $\mathbb{S}_{J-(j-\tilde{J})}^{M}$ will be labeled with $B_{\text {sp }}$ bits, and the output is denoted as $\mathbb{S}_{J-j}^{M}$. The function $S P$ will be formalized in the next subsection (at the first step of the first example). The steps detailed in lines 6(b-i) are repeatedly carried out until $j=J$. Therefore, if $J_{\mathrm{b}}=J$, then only 1 step is required. Generally, $J_{\mathrm{b}}$ can be interpreted as the number of elements of a vector in $\mathbb{S}^{M}$ that will be labeled at once in one step. Finally, the algorithm outputs $\mathcal{S}^{\text {GTIM }}$ as $\mathcal{S}^{\text {GTIM }}=\mathbb{S}_{0}^{M}$.

## D. More Examples

1) The algorithm with ( $M=2, B=3, J=2, J_{b}=2$ ): This is the example that is previously discussed and illustrated in Fig. 7. Here, we want to connect that example with the definitions in the previous subsection. In this example, the number of loops is one as $J_{\mathrm{b}}=2$. The algorithm 1 reduces to $\mathcal{S}^{\mathrm{GTIM}}=\mathbb{S}_{0}^{2}=\operatorname{SP}\left(2,3, \mathbb{S}_{2}^{2}\right)$. The steps taken in Figs. 7 (a-c) are formalized as following.

The $1^{\text {st }}$ step (Fig. 7 (a)): The first step is to partition $\mathbb{P}^{2,2}$ into $\mathbb{P}_{0}^{2,2}$ and $\mathbb{P}_{1}^{2,2}$ such that:

$$
\begin{equation*}
\left(\mathbb{P}_{0}^{2,2}, \mathbb{P}_{1}^{2,2}\right)=\underset{\mathbb{P}_{0}^{2,2}, \mathbb{P}_{1}^{2,2}}{\arg \max }\left\|\mathbf{d}_{\mathrm{S}}\left(\mathbb{P}_{0}^{2,2}\right)-\mathbf{d}_{\mathrm{S}}\left(\mathbb{P}_{1}^{2,2}\right)\right\| \tag{24}
\end{equation*}
$$

subject to:

$$
\begin{equation*}
\left|\mathbb{P}_{0}^{2,2}\right|=\left|\mathbb{P}_{1}^{2,2}\right|=\left\lfloor(M+1)^{J} / 2\right\rfloor \tag{25}
\end{equation*}
$$

$$
\begin{equation*}
\mathbb{P}_{0}^{2,2} \neq \mathbb{P}_{1}^{2,2} \neq \emptyset, \mathbb{P}_{0}^{2,2}, \mathbb{P}_{1}^{2,2} \subset \mathbb{P}^{2,2} \text { and } \mathbb{P}_{0}^{2,2} \cap \mathbb{P}_{0}^{2,2}=\emptyset \tag{26}
\end{equation*}
$$

The expression in 24 guarantees that $\mathbb{P}_{0}^{2,2}$ and $\mathbb{P}_{1}^{2,2}$ have the farthest distance with the constraints 25 and 26 . The expression in 25 indicates that both set $\mathbb{P}_{0}^{2,2}$ and $\mathbb{P}_{1}^{2,2}$ must contain $\left\lfloor(M+1)^{J} / 2\right\rfloor$ elements. This is equivalent to:

$$
\left\lfloor\log _{2}\left|\mathbb{P}_{0}^{2,2}\right|\right\rfloor=\left\lfloor\log _{2}\left|\mathbb{P}_{1}^{2,2}\right|\right\rfloor \geq B_{\mathrm{sp}}-1
$$

which means that the partitioned sets $\mathbb{P}_{0}^{2,2}$ and $\mathbb{P}_{1}^{2,2}$ can be at least represented by $B_{\mathrm{sp}}-1$ bits. The vectors in $\mathbb{P}_{0}^{2,2}$ and $\mathbb{P}_{1}^{2,2}$ will be assigned with the same $2^{B_{\mathrm{sp}}-1}$ binary vectors, which will be each appended by a new binary bit as the MSB. It is intuitive that 24 can be found in the outer layer of $\mathbb{P}^{2,2}$, see Fig. 8 for more examples and observe the positions of a collection of binary vectors that differs only at the MSB. This can be seen in the red and blue markers.

The $2^{\text {nd }}$ step (Fig. 7 b)): At the $2^{\text {nd }}$ step, the sets $\mathbb{P}_{0}^{2,2}$ and $\mathbb{P}_{1}^{2,2}$ can be further partitioned into $\mathbb{P}_{0,0}^{2,2}, \mathbb{P}_{0,1}^{2,2}, \mathbb{P}_{1,0}^{2,2}$ and $\mathbb{P}_{1,1}^{2,2}$. The arrangement of all the sets is done such that $\mathbb{P}_{0,1}^{2,2}$ is the neighbor of $\mathbb{P}_{1,1}^{2,2}$ as the Hamming distance of the bits associated with both sets is 1 . Doing this ensures that the MSB and the next bit position of binary vectors of $\mathbf{s}_{l} \in \mathbb{P}_{0,1}^{2,2}$ and $\mathbf{s}_{l} \in \mathbb{P}_{1,1}^{2,2}$ have the Hamming distance of 1 .

The $3^{\text {rd }}$ step (Fig. 7 (c)): The same procedures in the $2^{\text {nd }}$ step can be repeated in the $3^{\text {rd }}$ step. Alternatively, we can assign the LSB of the elements of $\mathcal{S}^{\mathrm{s}}$ and arrange it such that it differs only at 1 -bit position with their neighbors. The output of the third step is the ordered set $\tilde{\mathbb{P}}^{2,2}$ as shown in Fig. 7(c). Notice that the Hamming distance between two neighboring binary vectors is 1 ; moreover, the Hamming distance with the next neighbors is 2 and so on. Translating these processes with respect to $(19)$, the function $S P$ ensures that the farther constellation point has a larger Hamming distance. Other examples with $J=2$ and different values of $M$ and $B$ can be found in Fig. 8
2) The algorithm with $\left(M=2, B=6, J=4, J_{b}=2\right)$ : Instead of $J=2$, now we extend the previous example to $J=4$. In this example, it can be shown that the alreadydefined codebook can be used to design another codebook with higher dimensions. Furthermore, the number of iterations is 2 , and at each iteration $\operatorname{SP}\left(2,3, \mathbb{S}_{J-(j-2)}^{M}\right)$ is performed, which is already illustrated in Fig. 7 The output of the first iteration, i.e., $\mathbb{S}_{4-2}^{M}=\operatorname{SP}\left(2,3, \mathbb{S}_{4-0}^{M}\right)$, is given as:

$$
\left.\begin{array}{rl}
\mathbb{S}_{4-2}^{2}= & \left(\begin{array}{c}
{\left[\begin{array}{c}
\mathbf{x} \\
I_{0} \\
I_{0}
\end{array}\right]} \\
000 X X X
\end{array}\right]\left[\begin{array}{c}
\mathbf{x} \\
m_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
\mathbf{x} \\
m_{1} \\
m_{0}
\end{array}\right],\left[\begin{array}{c}
\mathbf{x} \\
m_{1} \\
I_{0}
\end{array}\right], \\
& {\left[\begin{array}{c}
\mathbf{x} \\
I_{0} \\
m_{0}
\end{array}\right],\left[\begin{array}{c}
\mathbf{x} \\
I_{0} \\
m_{1}
\end{array}\right],\left[\begin{array}{c}
\mathbf{x} \\
m_{1} \\
m_{1}
\end{array}\right],\left[\begin{array}{c}
\mathbf{x} \\
m_{0} \\
m_{1}
\end{array}\right]} \\
100 X X X & {[101 X X X X X X}
\end{array}\right] .
$$

In this case, the length of the vector $\mathbf{x}$ is two. At the last iteration, we have $J_{\mathrm{b}}=2$ and the output is $\mathcal{S}^{\text {GTIM }}=\mathbb{S}_{4-4}^{M}$ as
expressed as:

$$
\begin{aligned}
& \mathbb{S}_{4-4}^{2}=\left(\underset{I_{0}}{\left[\begin{array}{c}
I_{0} \\
I_{0} \\
I_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
m_{0} \\
I_{0} \\
I_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
m_{1} \\
m_{0} \\
I_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
m_{1} \\
I_{0} \\
I_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
I_{0} \\
m_{0} \\
I_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
I_{0} \\
m_{1} \\
I_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
m_{1} \\
m_{1} \\
I_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
m_{0} \\
m_{1} \\
I_{0} \\
I_{0}
\end{array}\right], ~}\right. \\
& {\left[\begin{array}{c}
I_{0} \\
I_{0} \\
m_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
m_{0} \\
I_{0} \\
m_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
m_{1} \\
m_{0} \\
m_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
m_{1} \\
I_{0} \\
m_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
I_{0} \\
m_{0} \\
m_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
I_{0} \\
m_{1} \\
m_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
m_{1} \\
m_{1} \\
m_{0} \\
I_{0}
\end{array}\right],\left[\begin{array}{c}
m_{0} \\
m_{1} \\
m_{0} \\
I_{0}
\end{array}\right], \cdots{ }_{00111101}^{0010111101}}
\end{aligned}
$$

Notice that the vector $\mathbf{x}$ is replaced with the result as depicted in Fig. 7(c), and the binary representation is appended by replacing the LSBs with new ordered binary vectors.
3) The algorithm with $\left(M=3, B=6, J=3, J_{b}=2\right)$ : The main purpose of this example is to visualize the modularity of the algorithm. At the first iteration, the process $\mathbb{S}_{3-1}^{3}=\operatorname{SP}\left(1,2, \mathbb{S}_{3-0}^{3}\right)$ is carried out. The output can be expressed as in (23). Visually, it is equivalent to order and label the symbol $s_{2}$ in $\mathcal{S}^{\text {GTIM }}$, which can be shown in Fig. 9 along $s_{2}$-axis. Then, on each plane with respect to the $s_{2}-$ axis, $\mathcal{S}^{\text {GTIM }}=\mathbb{S}_{3-3}^{3}=\operatorname{SP}\left(2,4, \mathbb{S}_{3-1}^{3}\right)$ is performed. Notice that the constellation for $\operatorname{SP}\left(2,4, \mathbb{S}_{3-1}^{3}\right)$ is shown in Fig. 8(b). Therefore, the final result can be seen in Fig. 9, which shows the arrangement of Fig. $8(b)$ on each plane. The location of each plane is denoted by (23), which is carried out in the first iteration, i.e., $\operatorname{SP}\left(1,2, \mathbb{S}_{3-0}^{3}\right)$.

## E. Optimality Discussion

First, we will discuss the effect of $J_{\mathrm{b}}$, which parameterizes algorithm 1. Recall that the arrangement in Fig. 8(c) has $J_{\mathrm{b}}=2$. Alternatively, we can also set $J_{\mathrm{b}}=1$ for the same problem. The result is shown in Fig. 10 a). It is obvious that the arrangement in Fig. 10(a) has a worse BER compared to Fig. 8(c). An easy way to see this is by looking at pairs of symbols that have the Hamming distance of 2. There are pairs with the Hamming distance of 2 in Fig. 10. a) that have a smaller Euclidean distance than that of Fig. 8(c). Furthermore, we can also find the same observation with $J_{\mathrm{b}}=3$ and $J_{\mathrm{b}}=2$. In this paper, we claim that with respect to $\sqrt[19]{ }$ the proposed algorithm is suboptimal and a detailed optimality discussion will be left for future studies. For a certain case such as the arrangement in Fig. 8(b), which resembles that of 16-QAM, Lemma 9 in [31] can be used to prove the optimality. However, for a general $J>2$ Lemma 9 in [31] cannot be directly applied. Following [31], some definitions and trivial operations such as the average distance spectrum, binary and codeword permutation needs to be defined first. The proposed algorithm gives a trade-off between the exhaustive search with the search size as defined in (18) and a good set of $\mathcal{S}^{\text {GTIM }}$ in terms of the Euclidean and Hamming distances.
It is also worth discussing here that if $\left|m_{0}-I_{0}\right|<\mid m_{i}-$ $m_{j} \mid, \forall i \neq j \in\{0,1, \cdots, M-1\}$ then the BER is limited by $\left|m_{0}-I_{0}\right|$. Therefore, in order to maintain the BER gain of SCGTIM, a minor modification for the function $S P$ is required.


Fig. 8. SP(•) with: (a) $M=3, B=3$ and $\eta=1.5$; (b) $M=3, B=4$ and $\eta=2$; (c) $M=4, B=4$ and $\eta=2$; (d) $M=5, B=5$ and $\eta=2.5$.


Fig. 9. An illustration for the proposed algorithm with (3, 6, 3, 2).

The modification is carried out such that the set $\mathbb{P}_{\mathrm{dc}}^{M, \tilde{J}}$ is predefined to include the vectors $\mathbf{p} \in \mathbb{P}^{M, J}$ that contain the value $m_{0}$. It means that the vectors containing the value $m_{0}$ are not used. Fig. 10.b) illustrates the aforementioned modification with the red cross, which denotes the unused vectors.

## V. ReSUlts and Discussion

## A. Simulation Setup


(a)

(b)

Fig. 10. (a) An output of algorithm 1 with $(4,4,2,1)$ and (b) a modified construction of Fig. 8 (c) if $\left|m_{0}-I_{0}\right|<\left|m_{i}-m_{j}\right|, \forall i \neq j \in\{0,1, \cdots, M-1\}$.

In order to compare different results with different $I_{0}$, a ratio $v$ is defined as follows:

$$
\begin{equation*}
v=\frac{I_{\mathrm{a}}-I_{0}}{d_{\min , M}} \tag{27}
\end{equation*}
$$

where $d_{\text {min, } \mathrm{M}}$ denotes the minimum distance of the conventional $M$-PAM and defined as $d_{\text {min, }}=\left(I_{\mathrm{b}}-I_{\mathrm{a}}\right) /(M-1)$. The case where $v>1$ shows that the $\left|I_{\mathrm{a}}-I_{0}\right|$ is greater than the minimum distance of the conventional $M$-PAM. As the performance gains of SC-TIM and SC-GTIM compared to the SC-PAM depend on $v$, we are interested in two different cases, which are $v<1$ (namely $C_{v<1}$ ) and $v>1$ (namely $C_{v>1}$ ). The case $C_{v>1}$ is the case where the LED is allowed

Table I. Simulation Parameters

| Parameters | Notation | Values | Unit |
| :--- | :--- | :--- | :--- |
| 3-dB LED bandwidth | $f_{\text {fed }}$ | 35 | MHz |
| LED Lambertian order | $m$ | 1 | - |
| PD's responsivity | $R$ | 0.7 | $\mathrm{~A} / \mathrm{W}$ |
| PD's gain | $M$ | 50 | - |
| PD's FoV | FoV | $90^{\circ}$ | - |
| PD's effective detection area | $A_{\mathrm{AP}}$ | 1 | $\mathrm{~cm}^{2}$ |
| Symbol rate | $R_{\mathrm{s}}$ | 50 | $\mathrm{MSymb} / \mathrm{s}$ |
| FFT size | $N$ | 32 | - |
| FEC threshold | $P_{\mathrm{b}, \text { target }}$ | $3.8 \times 10^{-3}$ | - |

to be completely off during the transmission, i.e., $I_{0}=0$. On the other hand, the case $C_{v<1}$ is the case where $I_{0} \neq 0$ and the distance between $I_{\mathrm{a}}$ and $I_{0}$ is smaller than the minimum distance of SC-PAM. This is when the SC-TIM can be worse than SC-PAM. In general, the cases $C_{v>1}$ and $C_{v<1}$ are defined to show extreme performance gains of SC-TIM and SC-GTIM, i.e., significant BER gains are achieved in $C_{v>1}$ and slight BER gains are achieved in the other case.

For our simulation setup, the characteristic of a high-speed IR LED VSMY3940X01 [40] is used to have a practical linear dynamic range value. Fitting the $P-I$ curve with the $5^{\text {th }}-$ order polynomial function, we observe that $I_{\mathrm{a}}=600 \mathrm{~mA}$ and $I_{\mathrm{b}}=800 \mathrm{~mA}$. As for the $I_{0}$, we pick $I_{0}=0$ for $C_{v>1}$ and $I_{0}=550 \mathrm{~mA}$ for $C_{v<1}$ with $M \in\{2,4\}$. The reason for this is to show the dependency of $I_{0}$ and the performance gains of SC-TIM and SC-GTIM.

Other parameter values are shown in Table. I] The LED is modeled as a first order Butterworth filter with the $3-\mathrm{dB}$ bandwidth $f_{\text {led }}=35 \mathrm{MHz}$. The LED Lambertian order for the optical channel calculation in (5) as $m=1$. The ADC is modeled as a zero-order hold with a fifth order Bessel lowpass filter (LPF) whose $3-\mathrm{dB}$ bandwidth is $R_{\mathrm{s}}\left(L_{\mathrm{cp}}+N\right) /(2 N)$. The PD's responsivity is chosen to be $R=0.7 \mathrm{~A} / \mathrm{W}$ based on, for example, [41]. Lastly, the FFT size is chosen to be $N=32$ [36] and the BER target used in this paper is $3.8 \times 10^{-3}$ [42].

For the optical channel setup, the vertical and horizontal distances between the UE and the AP are set to be $d_{\mathrm{v}}=1.6$ m and $d_{\mathrm{h}}=1.55 \mathrm{~m}$, respectively. For the orientation of the UE, we set $\theta$ to be 27.75, see the configuration $C_{1}$ in 34 , Fig. 7].

The room dimensions are assumed to be $5 \times 3.5 \times 3 \mathrm{~m}^{3}$. The reflectivities of the walls are assumed to be 0.3 , the reflectivity of the ceiling is 0.69 , and the reflectivity of the floor is 0.09. The reflectivities of the surfaces of the rectangular prism modeling the human body are assumed to be 0.6 with the exception of the upper surface, which is 0.9 . In addition, the resolution of the partition is 10 to generate the optical CIR [33]. A more detailed discussion and explanation as to where these values are taken from can be found in [32], [34].

## B. Error Performance

The BER comparisons are shown in Fig. 11. All APEPs are shown by solid lines to avoid confusion. The reason for using the solid lines for all APEPs is that the upper bounds are sufficiently tight, but not exactly equal, with each corresponding simulation result. The simulation results are shown with different markers. Note that for fair comparisons,
the set $\mathcal{S}^{\text {TIM }}$ is also constructed based on a Gray code and exhaustive search. Fig. 11(a) shows the advantages of SCGTIM over both SC-PAM and SC-TIM. Approximately 2 dB gain can be achieved by SC-TIM with $J=2$ compared to SCPAM. To show the performance of SC-TIM with $J=4$, the constellation size needs to be increased due to the limitation of vectors in $\mathcal{S}^{s}$ that can be chosen by SC-TIM. In order to obtain $\eta=1$ with $J=4$, the parameters $M$ and $K$ of the SC-TIM should be set as $M=4$ and $K=1$. As the minimum distance in the interval $\left[I_{\mathrm{a}}, I_{\mathrm{b}}\right]$ narrows, the BER performance of SC-TIM is worse than that of SC-PAM. On the other hand, as SC-GTIM can chose arbitrary vectors in $\mathcal{S}^{\text {s }}$, approximately 10 dB gain can be obtained. It is also shown that even by increasing $\eta$ to be $\eta=1.5$, the BER of SC-GTIM is still better than SC-PAM and SC-TIM.

In Fig. 11(a)-(d), we also show the results of the case when increasing $J$ for SC-GTIM does not change the BER performance. This means that by using the proposed set partitioning algorithm, the average minimum distance does not change significantly. This also means that the design problem for higher dimension $J$ can be reduced to design a good set $\mathcal{S}^{\text {GTIM }}$ for $J=2$ which will reduce the memory complexity. This issue will be discussed in the next subsection.

In Fig. 11(b), it can be seen that the performance of SCTIM with $\eta=2$ is better with respect to SC-PAM compared to those with $\eta=1$. For SC-GTIM, the BER gain decreases, but it is still significantly better than the others, i.e., approximately 5 dB better. In Fig. 8 b b) and (c), it is shown that for $J=2$, $\eta=2$ can be achieved by either $M=3$ and $M=4$. The results in Fig. 11b) show that we can take advantage of the fact that $\left|I_{0}-I_{\mathrm{a}}\right|>d_{\min , M}$. That is, as the same $\eta$ can be achieved by a lower constellation size, the lower constellation size should be chosen for designing the set $\mathcal{S}^{\mathrm{GTIM}}$. The reason for this is simply because of $d_{\min , 4}<d_{\min , 3}$. In addition, we also observe that about 2 dB gain with respect to SC-PAM can still be achieved for SC-GTIM with $\eta=3$.

The results for the case $C_{v<1}$ are shown in Fig. 11(c) and (d). At $\eta=1$, Fig. 11.c) shows the limitation of SCTIM. As shown in Fig. 5 (c), the BER is limited by two constellation points near $o_{I_{o}}$. This problem can be averaged out by increasing $J$, see (12). However, for SC-TIM with $J=4$, the problem with $d_{\text {min, } 4}<d_{\text {min,2 }}$ also appears in Fig. 11 (a); hence, the BER is still worse compared to that of SC-PAM. On the contrary, the advantage of SC-GTIM over the others is still maintained.

The limitation of SC-GTIM is shown in Fig. 11(d). To maintain the BER gain, it is clear that SC-GTIM cannot use the constellation diagram in Fig. 8(b) as the minimum distance is limited by all constellation points near $I_{0}$, which will result in a worse BER performance compared to SC-PAM and SC-TIM. However, due to the flexibility of SC-GTIM, the constellation diagram in Fig. 8(c) can be modified to be that as shown in Fig. 10.b). As the constellation points near $I_{0}$ acts as the limitation, the vectors $\mathbf{s}_{l}$ that contains the symbol $I_{\mathrm{a}}$ will not be picked. Therefore, the same structure in Fig. 8(b) can be used in Fig. 8 (c) by not picking $s_{j}=I_{\mathrm{a}}, \forall j$. The result is that the BER gain can still be achieved even though it is small.

(a)
$\times$ SC-PAM ---•SC-PAM approx. $\left(P_{\mathrm{b}, \mathrm{PAM}}\right) \quad \nabla$ SC-TIM $(J=2) \quad \Delta$ SC-TIM $(J=4)$
O SC-GTIM $(J=2) * \operatorname{SC}$-GTIM $(J=4)=$ APEP (upper bound) ............ $3.8 \times 10^{-3}$

Fig. 11. BER comparisons for: (a) $C_{v>1}$ and $\eta=1$; (b) $C_{v>1}$ and $\eta=2$; (c) $C_{v<1}$ and $\eta=1$; and (d) $C_{v<1}$ and $\eta=2$.

## C. Complexity and PAPR Penalties

At first, one might think that the brute-force search is the main reason for the additional complexity. In fact, the computational complexities of SC-TIM and SC-GTIM are the same as the conventional one if the Voronoi region is used provided that the total transmitted bits $B_{t}$ is the same. Fig. 8 (b) is used as an example. The goal is to decode the binary vector [0011] ${ }^{\mathrm{T}}$. Using the Voronoi region shown by red solid lines (only regions of interest are shown), the MSB (the $3^{\text {rd }}$ bit) can be decoded first. This can easily be done based on the location of the estimated $s_{1}$, namely $\hat{s}_{1}$, and the boundary $\left(m_{0}+m_{1}\right) / 2$. In addition, the operation is done over a single real value to allow a fair comparison with the SC-PAM. The rest of the bits, i.e., the $2^{\text {nd }}$ to the $0^{\text {th }}$ bits, can be estimated in the same way. Hence, the computational complexity is $O(B)$ and the same as the SC-PAM.

As previously discussed after (7) and (11), it cannot be guaranteed that the single-tap FDE can be applied to SC-TIM and SC-GTIM systems. We observe that this is true for SCTIM with $J>1$ since the matrix $\mathbf{R}_{\mathbf{s}}$ is not circulant. On the contrary, there are certain $\mathcal{S}^{\text {GTIM }}$ such that $\mathbf{R}_{\mathbf{s}}$ is circulant. For example, $\mathbf{R}_{\mathbf{s}}$ with the arrangements shown in Fig. 8 b) and Fig. 9 have a circulant $\mathbf{R}_{\text {s }}$. Generally, if:

$$
\begin{equation*}
\eta \in \mathbb{Z}^{+} \text {and } M=2^{\eta}-1 \tag{28}
\end{equation*}
$$

then $\mathbf{R}_{\mathbf{s}}$ will be circulant. A simple explanation for this is that for an equally-likely binary digit $\{0,1\}$, then all symbols will be equally-likely selected if 28 is satisfied. Hence, in terms of relative distances between constellation symbols, the constellation resembles that of SC-PAM whose $\mathbf{R}_{\mathbf{s}}$ is circulant. If the single-tap FDE can be applied, the computational complexity is $O\left(2 N \log _{2} N\right)$. Otherwise, it is $O\left(2 N^{2}\right)$.

The complexity penalties come from the mapping of bits to the symbol, which is the size of the lookup tables (LUTs) used to store the sets $\mathcal{S}^{\mathrm{TIM}}$ and $\mathcal{S}^{\mathrm{GTIM}}$. The case when $J=1$ is used in order for SC-PAM to have a fair comparison. For

Table II. PAPR Comparisons

|  | SC-PAM | SC-TIM | SC-GTIM |
| :--- | :--- | :--- | :--- |
| $\eta=1$ | 0.57 dB | 2.59 dB | 2.02 dB |
| $\eta=2$ | 0.58 dB | 1.83 dB | 1.8 dB |

SC-PAM, the memory complexity is $O\left(2^{B}\right)$. For SC-TIM and SC-GTIM, the complexity is $O\left(J 2^{B}\right)$ as the LUT is now a matrix. As previously mentioned, the BER performance of SCGTIM can be invariant with respect to $J$ due to the fact that the minimum distance is the same. In this case, the memory complexity of SC-GTIM can be limited to $O\left(2^{B+1}\right)$.

For PAPR comparisons, given the linear dynamic range [ $I_{\mathrm{a}}, I_{\mathrm{b}}$ ] and the new degree of freedom $I_{0}<I_{\mathrm{a}}$, it is obvious that the average energy per symbols of both SC-TIM and SCGTIM are lower; consequently, the PAPR increases as the peak value remains the same. However, $I_{0}$ can also be set such that $I_{\mathrm{b}}<I_{0}$. In this case, the PAPR does not always increase depending on $\left|I_{0}-I_{b}\right|$.

For PAPR comparisons, we only present the configurations which give the best BER performances in Fig. 11 . In addition, only the case $C_{v>1}$ is of interest as it gives a higher PAPR compared to that of the other case. Table. II summarized the PAPR comparisons. As previously mentioned before, the PAPRs of either SC-TIM and SC-GTIM are higher due to the fact that the average energy per symbol decreases. For a certain configuration of SC-TIM, when the symbol $I_{0}$ is picked more frequently than in SC-GTIM, SC-GTIM has a lower PAPR compared to SC-TIM; hence, the average energy per symbol of SC-TIM is lower than that of SC-GTIM.

## VI. Conclusions

This paper investigated the implementation of IM in singlecarrier PAM with block transmission for OWC. Specifically, we compared PAM (SC-PAM) and SC-TIM with the proposed SC-GTIM. The main difference between SC-TIM and SCGTIM was the flexibility in choosing an arbitrary number of
active time slots. Specifically, only predefined $K$ time slots were active for SC-TIM, while there was no fixed $K$ for SCGTIM. We showed that this flexibility in choosing the number of active time slots was used to mitigate the performance limitations of SC-TIM due to the linear dynamic range of LEDs. A set partitioning algorithm was also proposed to construct a codebook for SC-GTIM. In terms of BER, it was shown that more than a 3 dB gain could be achieved by SCGTIM. In terms of maximum achievable spectral efficiency for a given constellation size $M$, it was also shown that SCGTIM achieved better performance than the others. This was especially true in low-to-moderate spectral efficiency scenarios, where a $50 \%$ improvement could be achieved compared to SC-PAM. In terms of computational complexity penalties, we showed that there was a memory complexity penalty for both SC-TIM and SC-GTIM compared to that of SC-PAM due to the fact that the size of the codebook for mapping from bits to symbols was larger. In addition, there was also a PAPR penalty as the average energy per symbols for SC-TIM and SC-GTIM were lower than that of SC-PAM, while the maximum of the average transmitted energy was the same.

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[^1]:    ${ }^{1}$ If the term $K \log _{2} M$ does not exist, then it is straightforward given that $\underset{K}{\arg \max }\binom{J}{K}$ is the nearest integer of $J / 2$.

[^2]:    ${ }^{2}$ Note that the name set partitioning is also used in [27], but ours is different in the terms of the method used to partition and label the set. In addition, the algorithm in 27 does not discuss a method on how to choose a set of constellation symbols whose cardinality is smaller than the set of all possible constellation symbols. Joint Euclidean-distance and Hamming-distance search is also not discussed in 27.

