

# Capacity Bounds for Power- and Band-Limited Wireless Infrared Channels Corrupted by Gaussian Noise \*

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## Abstract

This paper finds asymptotically exact upper and lower bounds on the channel capacity of power and band-limited optical intensity channels corrupted by white Gaussian noise. This work differs from the oft investigated case of the Poisson photon counting channel in that not only are rectangular pulse amplitude schemes considered, but results for more general time-disjoint intensity modulation schemes are presented. The role of bandwidth is expressed by way of the effective dimension of the set of signals. The bounds show that at high optical signal-to-noise ratios the use of bandwidth efficient pulse sets is essential to achieve high spectral efficiencies. This result can be considered as an extension of previous work on photon counting channels which more closely model low optical intensity channels.

## 1 Introduction

Previous investigations into the capacity of optical intensity systems has focused primarily on channels in which the dominant noise source is quantum in nature. In these channels the transmitted optical intensity is constant in discrete time intervals. The received signal is modeled by a Poisson distributed count of the number of received photons in each discrete interval. The capacity of such channels has been reported under a variety of peak and average optical power constraints [2–5]. It has also been shown that schemes based on photon counting in discrete intervals require an exponential increase in bandwidth as a function of the rate (in nats/photon) for reliable communication [6].

In this work we present capacity bounds for a fundamentally different optical intensity channel. The indoor free-space optical channel can be modeled as a lowpass, linear

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channel with additive, white, signal independent, Gaussian noise [7]. Unlike previous treatments, capacity bounds are computed for a wider class of time-disjoint modulation schemes under a constraint on the bandwidth of any codeword.

## 2 The Optical Intensity Channel

Optical intensity channels transmit information by modulating the optical power of a laser or LED light source in proportion to an input electrical current,  $x(t)$ . The channel is well modeled as being linear and lowpass due to multipath distortion and the electrical characteristics of the optoelectronics [7]. The transmitted signal is corrupted by noise,  $z(t)$ , which can be modeled as being additive, white, zero-mean and Gaussian distributed [7]. Assuming that the channel is flat in the band of interest, the received electrical signal,  $y(t)$  can be written as

$$y(t) = x(t) + z(t).$$

Since  $x(t)$  is proportional to the transmitted optical intensity,  $\forall t x(t) \geq 0$ . Additionally, due to eye and skin safety regulations the average optical power is limited, and hence the average amplitude of  $x(t)$  is limited.

Note that this channel model applies not only to free-space optical channels but also to fiber optic links with negligible dispersion and signal independent, additive, white, Gaussian noise.

## 3 Signal Space Model

The free-space optical channel can be viewed as a vector channel with respect to the time-disjoint, orthonormal  $M$ -dimensional signal basis  $\{\phi_1(t), \phi_2(t), \dots, \phi_M(t)\}$ , where  $\phi_m(t) = 0$  for  $t \notin [0, T)$ . The vector channel can then be represented as  $\mathbf{Y} = \mathbf{X} + \mathbf{Z}$ , where each term is an  $M$ -dimensional random vector distributed as  $f_{\mathbf{Y}}(\mathbf{y})$ ,  $f_{\mathbf{X}}(\mathbf{x})$  and  $f_{\mathbf{Z}}(\mathbf{z})$  respectively where  $\mathbf{Z}$  is Gaussian with uncorrelated components. In order to adapt the signal space model to the optical intensity channel, we specify

$$\phi_1(t) = \frac{1}{\sqrt{T}}, \quad t \in [0, T) \quad (1)$$

as a basis function for every intensity modulation scheme [8]. This basis function represents the average amplitude of each symbol, and as a result represents the average optical power of each symbol.

The *admissible region* of the optical intensity modulation scheme is defined as the set of all points in the signal space which describe non-negative pulses, or formally

$$\Upsilon = \left\{ (v_1, v_2, \dots, v_M) \in \mathbb{R}^M : (\forall t \in \mathbb{R}), \sum_{m=1}^M v_m \phi_m(t) \geq 0 \right\}.$$

It can be shown that  $\Upsilon$  is the convex hull of a generalized  $N$ -cone with vertex at the origin [8]. Clearly  $f_{\mathbf{X}}(\mathbf{x}) = 0$  for  $\mathbf{x} \notin \Upsilon$  to ensure the non-negativity constraint is met.

Additionally,  $\Upsilon$  can be partitioned into sets of points of equal optical power,  $\Upsilon_k = \{(v_1, \dots, v_N) \in \Upsilon : v_1 = k, k \geq 0\}$ .

The average optical power,  $P$ , of an intensity signaling set can then be computed as

$$rP = \frac{1}{\sqrt{T}} \int_{\mathbf{x} \in \Upsilon} x_1 f_{\mathbf{x}}(\mathbf{x}) d\mathbf{x} \quad (2)$$

$$= \frac{1}{\sqrt{T}} P^G, \quad (3)$$

where  $P^G$  is the expected value of the  $x_1$  component of each signal vector and  $r$  is *responsivity* of the photodiode in units of Amperes per Watt. Note that  $r$  allows the optical constraints to be cast in terms of electrical quantities. In this case,  $rP$  is in units of Amperes.

## 4 Upper bound on Channel Capacity

An upper bound on the capacity of a Gaussian noise corrupted channel can be obtained by considering a sphere-packing argument in the set of all received codewords while imposing an average optical power constraint. This analysis is done in the same spirit as Shannon's sphere packing argument for channels subject to an average electrical power constraint [9]. Determining this bound requires that the volume of the set of received codewords be computed for a given average optical power limit.

### 4.1 Set of Transmitted Codewords

Consider transmitting a codeword  $\mathbf{x}$  formed from a series of  $N$ ,  $M$ -dimensional symbols at a low probability of error. Geometrically, in order for  $\mathbf{x}$  to be transmittable,  $\mathbf{x} \in \Upsilon^N$  where  $\Upsilon^N$  is the  $N$ -fold Cartesian product of  $\Upsilon$  with itself. It is possible to show that  $\Upsilon^N$  is itself the convex hull of a generalized cone with vertex at the origin [8]. In an analogous fashion to (1), define the  $\phi_1^{MN}$  basis vector as

$$\phi_1^{MN} = \frac{1}{\sqrt{N}} \underbrace{\underbrace{(1, 0, 0, \dots, 0)}_M, \underbrace{(0, 1, 0, 0, \dots, 0)}_M, \dots, (0, 0, \dots, 0, 1, 0, 0, \dots)}_{MN}$$

so that it represents the average optical power of each  $MN$ -dimensional codeword  $\mathbf{x} \in \Upsilon^N$ . The region  $\Upsilon^N$  is then parameterized by cross-sections for a given  $\phi_1^{MN}$  coordinate value.

For a fixed symbol period  $T$ , assume that the average optical power of each transmitted codeword is limited to be at most  $P^G/\sqrt{T}$  as defined in (3). In terms of the signal space definition for  $\Upsilon$ ,

$$\frac{1}{N} \sum_{n=1}^N x_{1,n} \leq P^G \quad (4)$$

where  $x_{1,n}$  is the coordinate value in the  $\phi_1$  direction for each constituent symbol. The transmitted  $NM$ -dimensional vector  $\mathbf{x}$  is taken from the set  $\Theta(P^G) = \Upsilon^N \cap \Psi(P^G)$  where  $\Psi(P^G)$  is a hyperplane defined so that the power constraint (4) is satisfied.

## 4.2 Set of Received Codewords

For some  $\mathbf{x} \in \Theta(P^G)$ , the received vector,  $\mathbf{Y}$  is normally distributed with mean  $\mathbf{x}$  and variance equal to the noise variance,  $\sigma^2$  per dimension. Let  $\Omega_{MN}$  denote the set of all possible received vectors. By the law of large numbers, with high probability  $\mathbf{Y}$  will lie near the surface of a sphere of radius  $\sqrt{MN(\sigma^2 + \epsilon)}$  where  $\epsilon$  can be made arbitrarily small by increasing  $N$ . A codeword is decoded by assigning all vectors contained inside the sphere to the given codeword.

Define the region  $\Omega_\infty$  as

$$\begin{aligned}\Omega_\infty &= \{\mathbf{x} + \mathbf{b} : \mathbf{x} \in \Theta(P^G), \mathbf{b} \in \rho B^{MN}\} \\ &= \Theta(P^G) \oplus \rho B^{MN}\end{aligned}$$

where  $\rho = \sqrt{MN\sigma^2}$ ,  $\oplus$  is the *Minkowski addition* of two sets and  $B^{MN}$  is the  $MN$ -dimensional unit ball. Since  $\Theta(P^G)$  is convex,  $\Omega_\infty$  is a parallel convex set of radius  $\rho$ , that is, the set of all points with distance at most  $\rho$  from  $\Theta(P^G)$ . It can be shown that for a  $\mathbf{y} \in \Omega_{MN}$  and the corresponding  $\mathbf{x} \in \Theta(P^G)$ , the probability that  $\mathbf{y}$  does not lie in  $\Omega_\infty$  can be made arbitrarily small by increasing  $N$ . Thus, the properties of  $\Omega_\infty$  must be determined in order to determine an upper bound.

Clearly,  $\Theta(P^G) \subset \Omega_\infty$  since  $\mathbf{0} \in B^{MN}$ . Where ever the boundary of  $\Theta(P^G)$  is smooth, the boundary points of  $\Omega_\infty$  are a subset of the points parallel to  $\Theta(P^G)$  at distance  $\rho$  away. Form the parallel extension of  $\Theta(P^G)$  as the region  $\Theta(P^G + p_\rho) - h$ , for some  $h, p_\rho > 0$  as the set of points which are at most distance of  $\rho$  away from  $\Theta(P^G)$  whenever the boundary of  $\Theta(P^G)$  is smooth. At points of discontinuity, that is, in the ‘‘corners’’ of the bodies in question, the points in  $\Omega_\infty$  lie inside the parallel extension of  $\Theta(P^G)$  at a distance  $\rho$  away due to the triangle inequality. In other words,

$$\Theta(P^G) \subset \Omega_\infty \subset \Theta(P^G + p_\rho) - h. \quad (5)$$

Let  $V(\cdot)$  evaluate to the volume of the region. Since all the regions are closed, an upper bound on  $V(\Omega_\infty)$  can be found using (5) to give,

$$V(\Theta(P^G + p_\rho)) > V(\Omega_\infty) > V(\Theta(P^G)).$$

By exploiting the geometry of the regions, it is possible to show that for large  $N$ ,  $p_\rho \rightarrow 2\sigma\sqrt{M}$  to give

$$V(\Theta(P^G + p_\rho)) = V(\Upsilon_1)^N \frac{(M-1)!^N}{(MN)!} (N(P^G + 2\sigma\sqrt{M}))^{MN}. \quad (6)$$

## 4.3 Upper bound Computation

The channel capacity in bits/symbol can be upper bounded using the sphere packing argument developed for electrical power constrained channels [9]. The maximum rate is upper bounded by the asymptotic number of non-overlapping spheres that can be packed in  $\Omega_{MN}$  as  $N$  goes to infinity. Using the previously defined regions,

$$\begin{aligned}C_s(\Phi) &\leq \lim_{N \rightarrow \infty} \frac{1}{N} \log_2 \frac{V(\Omega_{MN})}{V(\rho B^{MN})} \\ &\leq \lim_{N \rightarrow \infty} \frac{1}{N} \log_2 \frac{V(\Theta(P^G + p_\rho))}{V(\rho B^{MN})}.\end{aligned}$$

Using (6) and taking the limit the capacity of the channel can be upper bounded as

$$C_s(\Phi) \leq M \log_2 \left[ \left( \sqrt{T} \frac{rP}{\sigma} + 2\sqrt{M} \right) \frac{V(\Upsilon_1)^{1/M} (M-1)!^{1/M}}{M} \sqrt{\frac{e}{2\pi}} \right] \quad (7)$$

in units of bits/symbol for some symbol period  $T$ .

## 5 Lower bound on Channel Capacity

A lower bound on the capacity of the optical intensity channel can be found by computing the mutual information between the channel input and output for any input distribution. An asymptotically tight lower bound for high optical SNR can be achieved if the max-entropic source distribution, subject to an average optical power constraint, is used to compute this lower bound. It is possible to show that this choice of source distribution causes the upper bound (7) and lower bound to converge at high optical SNR.

Due to the signal space definition, the average optical power depends solely on the  $\phi_1$  coordinate and can be represented as in (2). By the maximum entropy principle, the maxentropic source distribution subject to this constraint must take the form  $f_{\mathbf{x}}^*(\mathbf{x}) = K \exp(-\lambda x_1)$ , for  $\mathbf{x} \in \Upsilon$  and for some  $K, \lambda > 0$  [10]. The constants  $K$  and  $\lambda$  can be found by using the form of the distribution and solving the following

$$\begin{aligned} \int_{\mathbf{x} \in \Upsilon} f_{\mathbf{x}}^*(\mathbf{x}) d\mathbf{x} &= 1 \\ \int_{\mathbf{x} \in \Upsilon} x_1 f_{\mathbf{x}}^*(\mathbf{x}) d\mathbf{x} &= P^G \end{aligned}$$

to yield

$$f_{\mathbf{x}}^*(\mathbf{x}) = \left( \frac{M}{P^G} \right)^M \frac{1}{V(\Upsilon_1)(M-1)!} \exp\left(-M \frac{x_1}{P^G}\right) \quad (8)$$

for  $\mathbf{x} = (x_1, x_2, \dots, x_M) \in \Upsilon$ . Notice that  $f_{\mathbf{x}}^*(\mathbf{x})$  is a function of solely the coordinate in the  $\phi_1$  direction which represents the average optical power of each symbol. The conditional distribution for a given  $x_1 = k$  is uniform over all elements of  $\Upsilon_k$ , which is entropy maximizing in the absence of constraints.

## 6 Bandwidth Constraint

Previous work on the photon counting channel indicated that under an average optical power constraint the rate was unbounded at the the price of an infinite bandwidth requirement [6]. It is clear that in order to have a consistent bound or notion of maximum rate for this channel that a bandwidth constraint must be placed on the space of signals transmitted.

Imposing a bandwidth constraint on a set of time-limited signals is not straightforward since the Fourier spectrum is necessarily time-unlimited. Let  $\mathbf{L}^2[0, T]$  denote the set of all finite energy signals with support contained in  $[0, T)$ . Define the  $(1 - \epsilon)$ -fractional energy bandwidth,  $W_\epsilon(x)$ , of a transmitted symbol  $x(t) \in \mathbf{L}^2[0, T]$  with Fourier transform  $X(f)$

as  $W_\epsilon(x) = \inf\{W \in [0, \infty) : \int_{-W}^W |X(f)|^2 df \geq (1 - \epsilon) \int_{-\infty}^{\infty} |X(f)|^2 df\}$  where  $\epsilon \in (0, 1)$  is fixed to some value, typically  $10^{-2}$  or  $10^{-3}$ . This bandwidth measure quantifies the frequency concentration of  $x(t)$ .

Consider approximating  $x(t) \in L^2[0, T]$  as a linear combination of some orthonormal basis functions. For a given  $W_\epsilon(x)$  and  $T$ , the best such basis, in the sense of minimizing the energy in the error of the approximation, is the family of prolate spheroidal wave functions,  $\varphi_n(f)$  [11]. The  $\varphi_n(f)$  are functions strictly time-limited to  $[0, T]$  which have the maximum energy in  $[-W_\epsilon(x), W_\epsilon(x)]$  of all unit energy functions [12]. The error in the approximation can be upper bounded as [13]

$$\inf_{\{a_i\}} \int_{-\infty}^{\infty} \left| X(f) - \sum_{n=0}^{\lceil 2W_\epsilon(x)T \rceil} a_n \varphi_n(f) \right|^2 df < 12\epsilon^2. \quad (9)$$

In this sense the signal  $x(t)$  can be thought of as being indistinguishable from some linear combination of prolate spheroidal basis functions. It can then be said that  $x(t)$  is essentially  $2W_\epsilon(x)T$  dimensional with the error in the approximation tending to zero as  $\epsilon \rightarrow 0$ . For every  $\mathbf{x} \in \Upsilon$  define

$$\kappa(\Phi) = \max_{\mathbf{x} \in \Upsilon} 2W_\epsilon \left( \sum_{m=1}^M x_m \phi_m(t) \right) T \quad (10)$$

as the effective dimension of the signal space associated with the optical intensity basis  $\Phi$ . This channel bandwidth constraint can be interpreted as ensuring that the channel is able to support the transmission of at most  $\kappa(\Phi)$  dimensions per symbol. Since each transmitted symbol in the model is at most  $\kappa(\Phi)$  dimensional, the received symbols are uncorrupted by the channel, i.e., the received signals are indistinguishable from the transmitted signals in the sense of (9).

The upper bound on channel capacity for a given  $\Phi$  in (7) can be represented as a bound on the maximum spectral efficiency for a channel bandwidth of  $W_{\text{ch}}$  Hz using the effective dimension  $\kappa(\Phi)$  (10) and (3) as,

$$C_\eta(\Phi) \leq \frac{2M}{\kappa(\Phi)} \log_2 \left[ \left( \sqrt{\frac{\kappa(\Phi)}{2W_{\text{ch}}}} \frac{rP}{\sigma} + 2\sqrt{M} \right) \frac{V(\Upsilon_1)^{1/M} (M-1)!^{1/M} \sqrt{\frac{e}{2\pi}}}{M} \right] \quad (11)$$

in units of bits/s/Hz. Unlike the band-limited case where the dimension of each basis signal is one, here the effective dimension of each signal in  $\Upsilon$  must be computed. The factor  $M/\kappa(\Phi)$  can be thought of as a measure of the dimensional efficiency of a given model since  $M$  represents the dimension of each transmitted signal while  $\kappa(\Phi)$  is the maximum dimension of the set of signals determined by  $\Phi$  using a  $(1 - \epsilon)$ -fractional energy bandwidth measure.

At high optical SNRs, the lower bound on capacity tends to the true capacity since it is chosen to be the maxentropic source distribution. It is possible to show that using this bandwidth constraint the upper and lower capacity bounds converge at high optical signal-to-noise ratios. As a result, we make the claim that the upper and lower capacity bounds computed here are asymptotically exact.

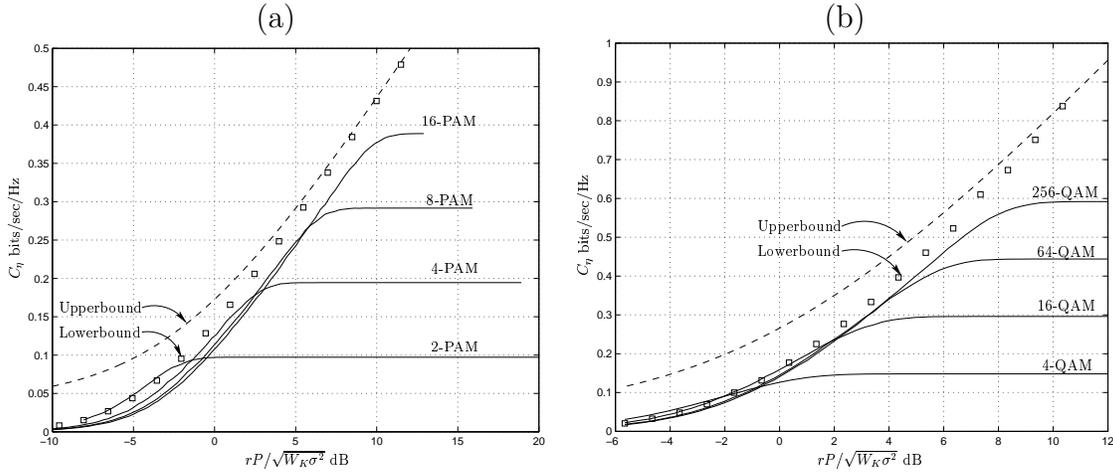


Figure 1: Bounds on the achievable spectral efficiencies using (a) rectangular PAM and (b) raised-QAM along with results for some uniform discrete constellations.

## 7 Examples and Discussion

### 7.1 PAM

Form an  $M$ -ary pulse-amplitude modulation scheme using the rectangular pulse shape of (1). Define the effective dimension of the scheme using the 99% fractional power bandwidth ( $K=0.99$ ) to yield  $\kappa_{\text{PAM}} = 20.572$ . Figure 1(a) presents the upper and the lower bounds on  $C_\eta(\Phi_{\text{PAM}})$  for the PAM scheme defined as well as spectral efficiency curves for discrete uniform 2, 4, 8 and 16 point constellations versus optical SNR. These spectral efficiency curves were computed numerically using Monte Carlo methods.

The upper bound on capacity is obtained by direct application of (11) to give,

$$C_\eta(\Phi_{\text{PAM}}) \leq \frac{2}{\kappa(\Phi_{\text{PAM}})} \log_2 \left[ \left( \sqrt{\frac{\kappa(\Phi_{\text{PAM}})}{2W}} \frac{rP}{\sigma} + 2 \right) \sqrt{\frac{e}{2\pi}} \right].$$

The lower bound on capacity was determined first by computing  $f_{\mathbf{Y}}(\mathbf{y})$ , which takes the form

$$\begin{aligned} f_{\mathbf{Y}}(\mathbf{y}) &= f_{\mathbf{X}}^*(\mathbf{x}) * f_{\mathbf{Z}}(\mathbf{z}) \\ &= \frac{1}{PG} \left( 1 - Q \left( \frac{y}{\sigma} - \frac{\sigma}{PG} \right) \right) \exp \left( \frac{\sigma^2 - 2y}{2PG} \right) \end{aligned}$$

where  $Q(x) \triangleq (1/\sqrt{2\pi}) \int_x^\infty \exp(-u^2/2) du$ . Since  $f_{\mathbf{Y}}(\mathbf{y})$  does not have a closed form in this case, computing the mutual information explicitly is impossible. Figure 1(a) shows the lower bound computed for a number of points. Note that at high SNR the lower and upper bounds on capacity approach one another.

### 7.2 Raised-QAM

An optical 3-dimensional raised-QAM scheme can be defined by specifying  $\phi_1(t)$  as in (1),  $\phi_2(t) = \sqrt{2/T} \cos(2\pi t/T)$  and  $\phi_3(t) = \sqrt{2/T} \sin(2\pi t/T)$ . for  $t \in [0, T)$  [8]. Figure

1(b) presents a plot of the upper bound on capacity (11) for a 3-dimensional raised-QAM scheme which takes the form

$$C_\eta(\Phi_{\text{QAM}}) \leq \frac{6}{\kappa(\Phi_{\text{QAM}})} \log_2 \left[ \left( \sqrt{\frac{\kappa(\Phi_{\text{QAM}})}{2W}} \frac{rP}{\sigma} + 2\sqrt{3} \right) \sqrt{\frac{e}{18\pi^{1/3}}} \right].$$

Using the same definition of bandwidth,  $K = 0.99$ ,  $\kappa_{\text{QAM}} = 27.038$ . As is the case with PAM, the lower bound must be computed numerically. Unfortunately, computation of  $f_{\mathbf{Y}}(\mathbf{y})$  is difficult and the lower bound was computed using a discretized version of  $f_{\mathbf{X}}^*(\mathbf{x})$  (8) and integrated using Monte Carlo methods. The upper and lower bounds approach one another at high optical SNRs. Spectral efficiency curves for 4, 16, 64 and 256 point uniform distributions were determined using Monte Carlo techniques and are also presented.

### 7.3 Prolate Spheroidal Wave Function Bases

As discussed in Section 6, for a given  $2WT$  product, the prolate spheroidal wave functions are the time-limited functions with support in  $[0, T]$  with maximum energy in the frequency band  $[-W, W]$  of all unit energy functions [12]. In light of the bandwidth constraint imposed it seems natural to form an optical intensity signaling scheme based on this orthonormal family of functions.

An  $M$ -PSWF optical intensity model be formed by performing a Gram-Schmidt orthogonalization procedure with  $\phi_1(t)$  and  $\varphi_m(t)$  for  $m = 0, 1, \dots, M - 2$  at a time-bandwidth product of  $2W_\epsilon(\phi_1)T$  to form the basis set  $\Phi_{\text{PSWF}}$ . The basis functions for this model are then denoted  $\phi_1(t)$  and  $\varphi'_m(t)$ .

### 7.4 Discussion

An important difference over the electrical channels is that the upper and the lower bound depend explicitly on the pulse set chosen. Thus,  $C_\eta(\Phi)$  is a measure of the maximum spectral efficiency of the optical channel for the given pulse set. Indeed, in order to determine a bound on the maximum spectral efficiency,  $C_\eta(\Phi)$  should be maximized over all  $\Phi$ . Some early work on the photon counting channel demonstrated that narrow pulse position techniques were optimal pulse techniques in the sense of a given average distance measure [14, 15]. Capacity results for the photon counting channel nearly exclusively assume that rectangular pulse techniques are employed. Here the assumption on the shape of the pulses is removed and the maximum spectral efficiencies are computed for a given pulse set. However, the rate maximizing pulse set for an optical intensity channel under an average optical and bandwidth constraint is an open problem.

At high optical signal-to-noise ratios, pulse techniques have lower maximum spectral efficiencies than bandwidth efficient techniques. Figure 2 presents a comparison of the capacity bounds derived earlier. Note that at high SNRs signaling schemes based on  $M$ -PSWF and raised-QAM pulse sets have far greater spectral efficiencies over rectangular PAM techniques at a given SNR. At lower optical SNRs, the derived bounds are loose and do not reveal any new insight. Indeed, at low SNR, when the available spectral

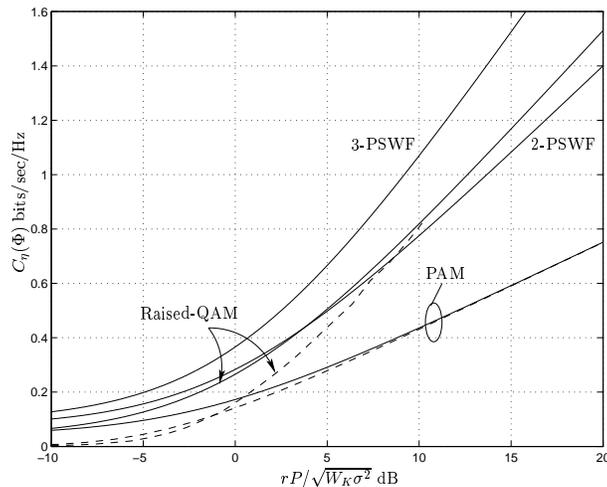


Figure 2: Comparison of achievable spectral efficiencies using rectangular PAM, raised-QAM and 2-, 3-PSWF.

efficiencies tend to zero, rectangular pulse techniques are attractive due to their ease of implementation.

## 8 Conclusions

We have derived capacity bounds for the optical intensity channel with average optical power and bandwidth constraints in Gaussian noise. These results complement rather than contradict previous work on the Poisson photon counting channel. The photon counting channel can be viewed as an optical system operating at low optical power where the quantum nature of the photons dominates performance. Rectangular pulse techniques are uniquely considered since the bandwidth of the channel is considered to be very large.

In this work, we treat a fundamentally different channel. Indoor free-space channels suffer from reduced bandwidth due to multipath distortion and from white, Gaussian noise due to high background illumination. The derived capacity bounds are not restricted to pulse techniques, as in previous work, but treat a wider class of time-disjoint optical intensity schemes. A bandwidth constraint is imposed on the set of signals that are transmitted by way of determining the effective dimension of the space of time-limited signals with a given fractional power bandwidth. The derived capacity bounds demonstrate that for a given average optical power, pulse techniques have significantly lower maximum spectral efficiencies than bandwidth efficient techniques. In particular, significant rate gains can be had by using a  $M$ -PSWF or raised-QAM pulse sets over a rectangular PAM at high optical signal-to-noise ratios.

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