

Design and Performance of Visible Light Communication Systems

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EE14RESCH11007

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1 Introduction

- Literature
- Motivation

2 Problems Addressed

- Power Allocation for Uniform Illumination with Stochastic LED Arrays
- Performance of Stochastic LED Arrays based VLC
- Closed-form expressions of BER for BPP in circular region
- Optimal Power Allocation for Uniform Illumination

3 References

Introduction

Literature

- What is Visible Light Communication (VLC) ?
- How light is used for communication ?
- Why light-emitting diodes (LED) are used ?

Introduction

Literature

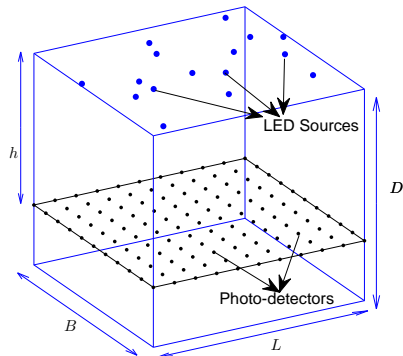


Figure: System model

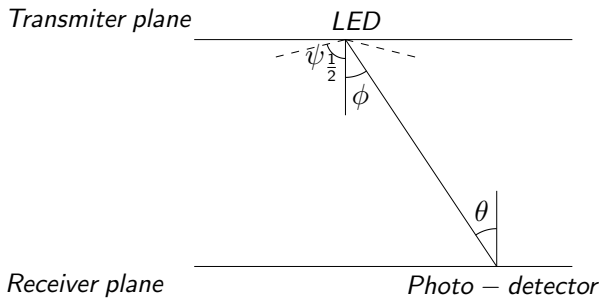


Figure: Propagation model

Introduction

Literature

- Using the Lambertian radiation pattern to model the LED radiant intensity,

$$\mathcal{R}(\phi) = \frac{(m+1)\cos^m(\phi)}{2\pi},$$

$m = \frac{\ln \frac{1}{2}}{\ln \left(\cos \left(\psi_{\frac{1}{2}} \right) \right)}$ is the order of Lambertian emission, with $\psi_{\frac{1}{2}}$ being the LED semi-angle at half power, provided by the manufacturer.

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- The channel direct current (DC) gain can then be expressed as

$$H = \frac{\mathcal{R}(\phi) \cos(\theta) A}{d^2} = \frac{(m+1) \cos^m(\phi) A \cos(\theta)}{2\pi d^2}$$

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where,

$$H_{ij} = \frac{(m+1) Ah^{m+1}}{2\pi d_{ij}^{m+3}}$$

and n_j is additive white Gaussian noise (AWGN) with $n_j \sim \mathcal{N}(0, \sigma_j^2)$.

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$$\sigma_j^2 = \sigma_{shot}^2 + \sigma_{thermal}^2$$

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$$\begin{aligned}\sigma_{shot}^2 &= 2qRP_{r_j}B + 2qI_{bg}I_2B \\ \sigma_{thermal}^2 &= \frac{8\pi kT_k}{G}\eta A I_2 B^2 + \frac{16\pi^2 kT_k \Gamma}{g_m}\eta^2 A^2 I_3 B^3\end{aligned}$$

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Signal to Noise Ratio (SNR) at photo-detector j is defined as

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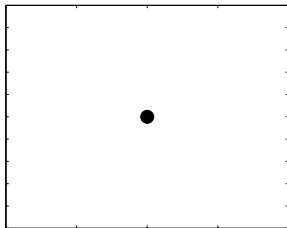
The quality factor, for measuring the performance of the light source, can be expressed as

$$F_\Lambda = \frac{\bar{\Lambda}}{2\sqrt{\text{var}(\Lambda)}},$$

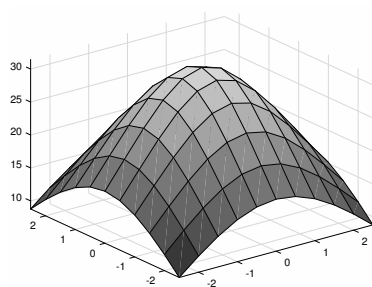
where $\bar{\Lambda}$ and $\text{var}(\Lambda)$ are the mean and variance of $\{\Lambda_j\}_{j=1}^K$, where K is the number of photodetectors.

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Literature



(a) Single point source



(b) SNR profile

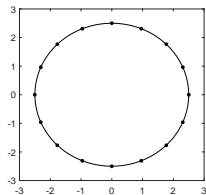
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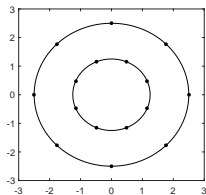


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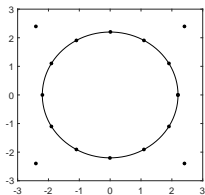
Motivation



(a) Circular geometry



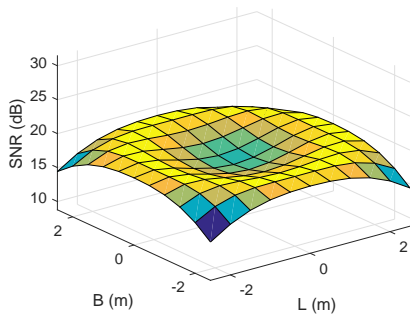
(b) Concentric circular geometry



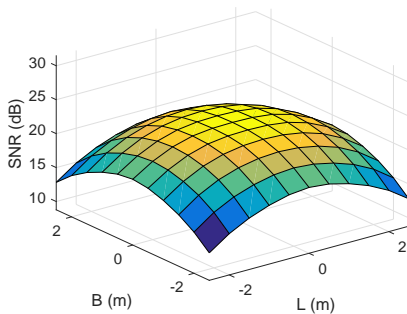
(c) Circle-square geometry

Introduction

Motivation



(a) Circular geometry

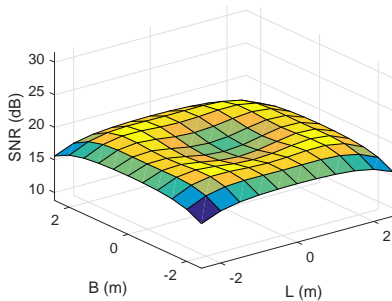


(b) Concentric Circular

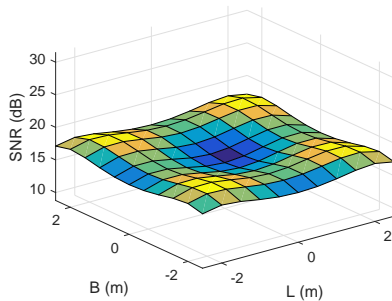
Figure: SNR distribution with equal power allocation

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Motivation



(a) With equal power allocation and optimal location

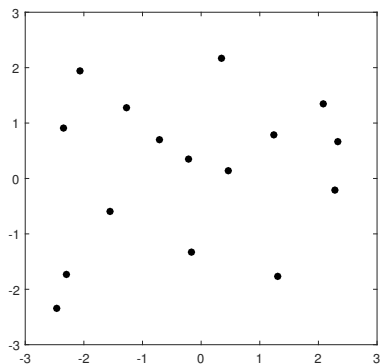


(b) With optimal power allocation and optimal location

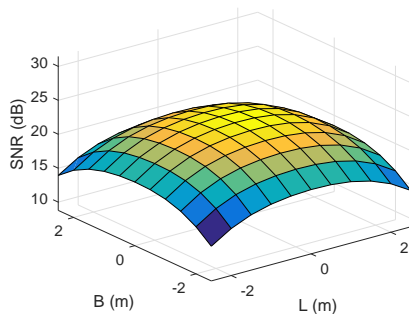
Figure: SNR distribution for circle-square geometry

Power Allocation for Uniform Illumination with Stochastic LED Arrays

Power Allocation for Uniform Illumination with Stochastic LED Arrays



(a) A realization of BPP



(b) Average SNR profile with equal power allocation

Power Allocation for Uniform Illumination with Stochastic LED Arrays

Heuristic power allocation:

$$P_{t_i} = \frac{r_i^\alpha}{\sum_{n=1}^N r_n^\alpha} P,$$

Power Allocation for Uniform Illumination with Stochastic LED Arrays

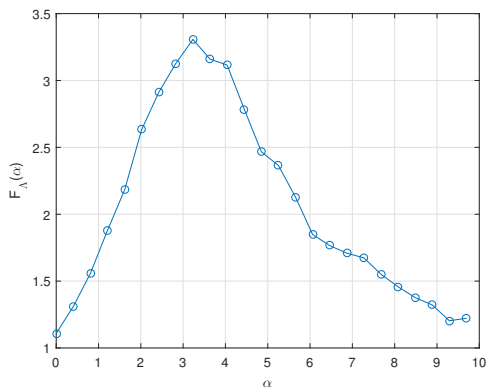
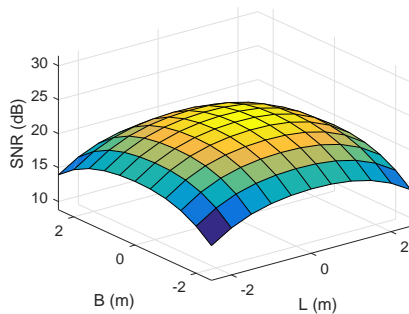


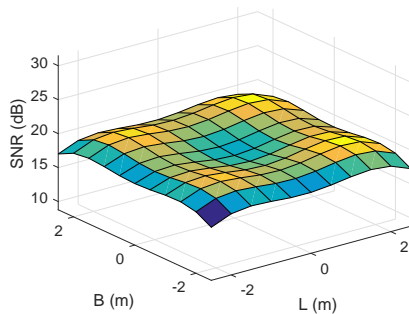
Figure: $F_{\Lambda}(\alpha)$ has a maximum.

Optimum value of α is obtained from golden section search algorithm

Power Allocation for Uniform Illumination with Stochastic LED Arrays



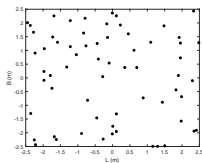
(a) Equal power allocation



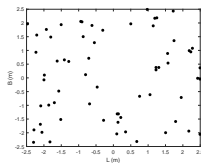
(b) Suboptimal power allocation

Figure: Average SNR for a BPP. $N = 16$

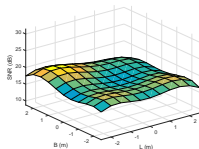
Power Allocation for Uniform Illumination with Stochastic LED Arrays



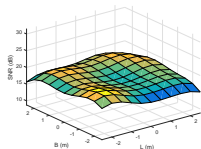
(a) $N = 64$, BPP realization 1



(b) $N = 64$, BPP realization 2



(c) SNR profile for realization 1



(d) SNR profile for realization 2

Uniform illumination possible with random distribution.

Performance of Stochastic LED Arrays based VLC with Uniform Irradiance

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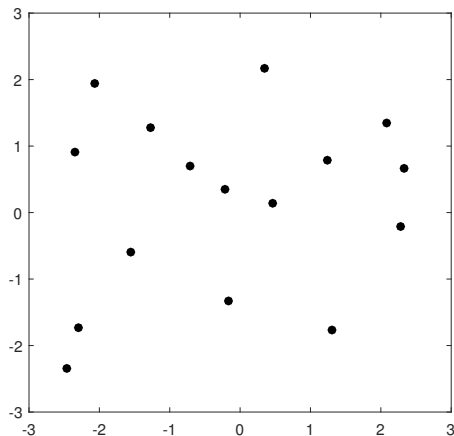


Figure: Realization of a BPP

The optical signal transmitted by the i th LED of a VLC is given by

$$p_i(t) = P_{t_i} [1 + M_I x_i(t)],$$

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$$p_i(t) = P_{t_i} [1 + M_I x_i(t)],$$

After removing DC component, the signal received at j th photo-detector from all the source LEDs is

$$y_j = RP_{r_j} + n_j$$

where,

$$P_{r_j} = \sum_{i=1}^N H_{ij} P_{t_i} M_I x_i.$$

Performance of Stochastic LED Arrays based VLC with Uniform Irradiance

The AWGN noise n_j at the photo-detector is the sum of the contributions from shot noise and thermal noise, and expressed as

$$\sigma_j^2 = \sigma_{shot}^2 + \sigma_{thermal}^2$$

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Heuristic power allocation:

The transmit power at the i th transmitter is given by the heuristic

$$P_{t_i} = \frac{r_i^\alpha}{\sum_{j=1}^N r_j^\alpha} P,$$

Performance of Stochastic LED Arrays based VLC with Uniform Irradiance

The received symbol at the central photodetector is given by

$$y = RP_{r_0} + n_0$$

and,

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The received symbol at the central photodetector is given by

$$y = RP_{r_0} + n_0$$

and,

$$P_{r_0} = C_1 \sum_{i=1}^N V_i x_i$$

where x_i is the modulating bipolar OOK signal. All LEDs are assumed to transmit the same message signal.

$$C_1 = \frac{PM_I (m+1) Ah^{m+1}}{2\pi} \quad \text{and}$$
$$V_i = \frac{r_i^\alpha}{\left(\sum_{j=1}^N r_j^\alpha\right) \left(\sqrt{h^2 + r_i^2}\right)^{m+3}}.$$

Performance of Stochastic LED Arrays based VLC with Uniform Irradiance

$$y = \left(RC_1 \sum_{i=1}^N V_i \right) x + n_0$$

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BER for this system is given by

$$P_e = \mathbb{E}_{\Phi} \left[Q \left(\frac{RC_1 \sum_{i=1}^N V_i}{\sigma_0} \right) \right]$$

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Using first order taylor approximation, Schwarz's inequality, and Jensen's inequality the above expression is reduced to

$$P_e \gtrsim Q \left(\frac{RC_1 \sum_{i=1}^N \mathbb{E}_{\Phi} [V_i]}{\sqrt{\mathbb{E}_{\Phi} [\sigma_0^2]}} \right)$$

Performance of Stochastic LED Arrays based VLC with Uniform Irradiance

For a BPP distributed over a square region of area W , the probability density function (PDF) of the distance to the i^{th} nearest LED from the origin is

$$f_{r_i} = \begin{cases} \frac{2\pi r}{W} \frac{(1-p)^{N-i} p^{i-1}}{\mathcal{B}(N-i+1, i)} & 0 < r < R_{in} \\ \frac{2(\pi-4\theta)r}{W} \frac{(1-q)^{N-i} q^{i-1}}{\mathcal{B}(N-i+1, i)} & R_{in} < r < R_c \\ 0 & R_c < r \end{cases}$$

where $\theta = \cos^{-1}(R_{in}/r)$, $p = \frac{\pi r^2}{W}$, $q = \frac{\pi r^2 - 4r^2\theta + 2r^2 \sin(2\theta)}{W}$, R_{in} and R_c are the radii of the incircle and circumcircle of W .

Performance of Stochastic LED Arrays based VLC with Uniform Irradiance

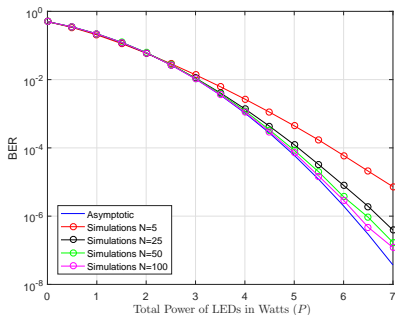
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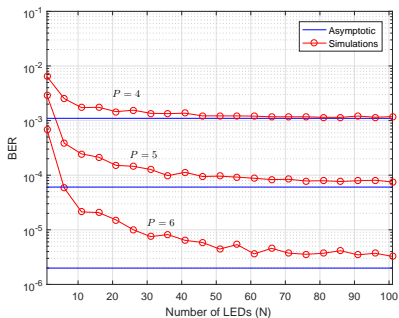
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$$\begin{aligned} \mathbb{E}_\Phi[V_i] = & \frac{1}{\sum_{j=1}^N \mathbb{E}_\Phi[r_j^\alpha]} \left[\frac{R_{in}^\alpha}{h^{m+3} \mathcal{B}(N-i+1, i)} \times \right. \\ & \sum_{k=0}^{N-i} \frac{\binom{N-i}{k} (-1)^k}{(k+i+\alpha/2)} \left(\frac{\pi R_{in}^2}{W} \right)^{k+i} {}_2F_1 \left(\frac{m+3}{2}, k+i+\alpha/2; k+i+\alpha/2+1; -\frac{R_{in}^2}{h^2} \right) \\ & \left. + \frac{R_{in}^\alpha}{\mathcal{B}(N-i+1, i)} \sum_{k=0}^{N-i} \binom{N-i}{k} (-1)^k \left(\frac{R_{in}^2}{W} \right)^{k+i} \int_0^{\frac{\pi}{4}} \frac{2(\pi-4\theta) \sin(\theta) (\pi-4\theta+2\sin(2\theta))^{k+i-1}}{\cos^{2(k+i-1)+\alpha-m}(\theta) \left(\sqrt{R_{in}^2 + h^2 \cos^2(\theta)} \right)^{m+3}} d\theta \right] \end{aligned}$$

Performance of Stochastic LED Arrays based VLC with Uniform Irradiance



(a) BER performance with respect to source power for different number of LEDs



(b) Convergence of BER with respect to number of LEDs

Closed-form expressions of BER for BPP in circular region

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Source LEDs are distributed uniformly in circular region of radius R_c (circumference of dimension of room).

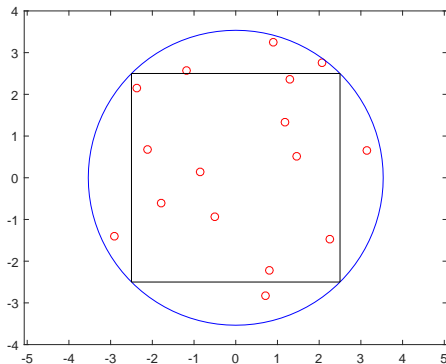


Figure: Realization of a circular BPP

Closed-form expressions of BER for BPP in circular region

The PDF of the distance of i^{th} nearest LED location from origin for circular BPP for $0 \leq r \leq R_c$ is given by

$$f_{r_i} = \frac{2r}{R_c^2 B(i, N - i + 1)} \left(\frac{r^2}{R_c^2} \right)^{i-1} \left(1 - \frac{r^2}{R_c^2} \right)^{N-i}$$

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$$\mathbb{E}_{\Phi} [r_i^{\alpha}] = \frac{R_c^{\alpha}}{\mathcal{B}(N-i+1, i)} \sum_{k=0}^{N-i} \frac{\binom{N-i}{k} (-1)^k}{\left(i + k + \frac{\alpha}{2}\right)}$$

Closed-form expressions of BER for BPP in circular region

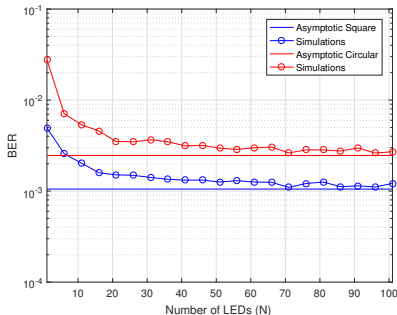
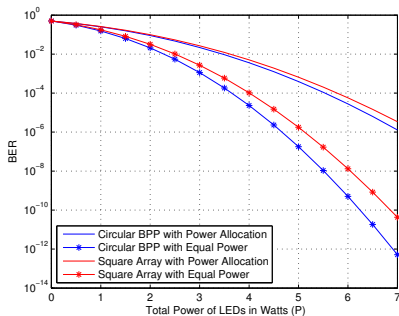
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$$\mathbb{E}_\Phi [r_i^\alpha] = \frac{R_c^\alpha}{\mathcal{B}(N-i+1, i)} \sum_{k=0}^{N-i} \frac{\binom{N-i}{k} (-1)^k}{\left(i + k + \frac{\alpha}{2}\right)}$$

$$\begin{aligned} \mathbb{E}_\Phi [V_i] &= \frac{R_c^\alpha}{h^{m+3} \mathcal{B}(N-i+1, i) \sum_{j=1}^N \mathbb{E}_\Phi [r_j^\alpha]} \sum_{k=0}^{N-i} \frac{\binom{N-i}{k} (-1)^k}{\left(i + k + \frac{\alpha}{2}\right)} \\ &\times {}_2F_1 \left(\frac{m+3}{2}, i + k + \frac{\alpha}{2}; i + k + \frac{\alpha}{2} + 1; -\frac{R_c^2}{h^2} \right) \end{aligned}$$

Closed-form expressions of BER for BPP in circular region



BER can be used to estimate the cost of the system in terms of the number of LEDs

Optimal Power Allocation for Uniform Illumination in VLC Systems

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Given a distribution of source LEDs, How to find the optimum power allocation for uniform illuminance ?

Optimal Power Allocation for Uniform Illumination in VLC Systems

Given a distribution of source LEDs, How to find the optimum power allocation for uniform illuminance ?

Minimize the variance of the received power

$$\underset{P_{t_i}}{\text{minimize}} \quad \mathbb{E} \left[(P_{r_j} - \mathbb{E} [P_{r_j}])^2 \right]$$

Optimal Power Allocation for Uniform Illumination in VLC Systems

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Minimize the variance of the received power

$$\underset{P_{t_i}}{\text{minimize}} \quad \mathbb{E} \left[(P_{r_j} - \mathbb{E} [P_{r_j}])^2 \right]$$

$$\begin{aligned} \text{var}(P_{r_j}) = & \frac{1}{2} \sum_{i=1}^N \left(\frac{2 \sum_{j=1}^K H_{ij}^2}{K} - \frac{2 \left(\sum_{j=1}^K H_{ij} \right)^2}{K^2} \right) P_{t_i}^2 \\ & + 2 \sum_{u=1}^N \sum_{v=i+1}^N \left(\frac{2 \sum_{j=1}^K H_{uj} H_{vj}}{K} \right. \\ & \quad \left. - \frac{2 \left(\sum_{j=1}^K H_{uj} \right) \left(\sum_{j=1}^K H_{vj} \right)}{K^2} \right) P_{t_u} P_{t_v} \end{aligned}$$

Optimal Power Allocation for Uniform Illumination in VLC Systems

The objective function is expressed in quadratic form as

$$\begin{aligned} & \underset{\mathbf{x}}{\text{minimize}} && \frac{1}{2} \mathbf{x}^T \mathcal{P} \mathbf{x} \\ & \text{subject to} && \mathcal{G} \mathbf{x} \succeq \mathbf{0} \\ & && \mathcal{A} \mathbf{x} = P \end{aligned}$$

where $\mathcal{G} = \text{diag}(-1, \dots, -1)$, $\mathcal{A} = [1, \dots, 1]$ and elements β_{ij} of matrix \mathcal{P} is given by

$$\beta_{ij} = \begin{cases} \frac{2 \sum_{p=1}^K a_{ip}^2}{K} - \frac{2(\sum_{p=1}^K a_{ip})^2}{K^2}, & i = j \\ \frac{2 \sum_{p=1}^K a_{ip} a_{jp}}{K} - \frac{2(\sum_{p=1}^K a_{ip})(\sum_{p=1}^K a_{jp})}{K^2}, & i \neq j \end{cases}$$

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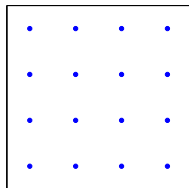
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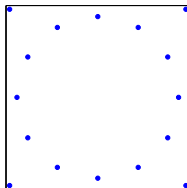
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Numerically solved using quadratic programming (QP) through the `solvers.qp` command in Python using CVXOPT solver.

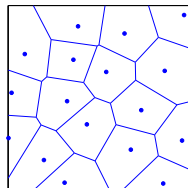
Optimal Power Allocation for Uniform Illumination in VLC Systems



(a) Square Array

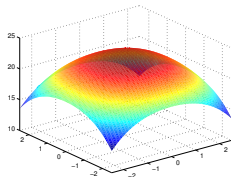


(b) Circle Square

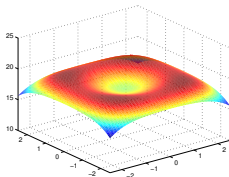


(c) Realization of HCPP

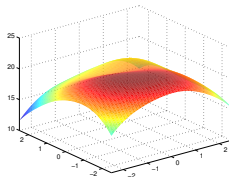
Optimal Power Allocation for Uniform Illumination in VLC Systems



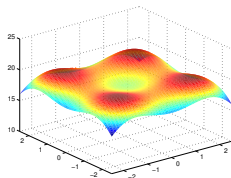
(a) Square array with equal power



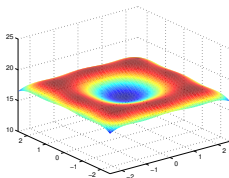
(b) Circle square with equal power



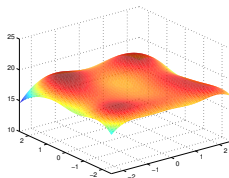
(c) HCPP with equal power



(d) Square array with optimum power



(e) Circle square with optimum power



(f) HCPP with optimum power

Optimal Power Allocation for Uniform Illumination in VLC Systems

Table: Variance of received power

	Equal power	Heuristic power	Optimum power
Square Array	6.4414e-15	6.5287e-16	6.3785e-16
Circle Square	3.0134e-15	1.1267e-15	8.0382e-16
BPP	5.1986e-14	3.9660e-14	9.8679e-15
HCPP	2.2504e-14	1.2721e-14	2.2736e-15

Table: Quality factor

	Equal power	Heuristic power	Optimum power
Square Array	1.34	3.30	3.28
Circle Square	3.79	4.48	5.06
BPP	1.27	2.80	3.49
HCPP	1.49	3.27	4.73

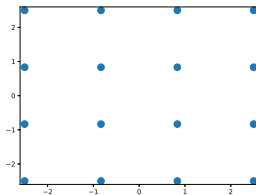
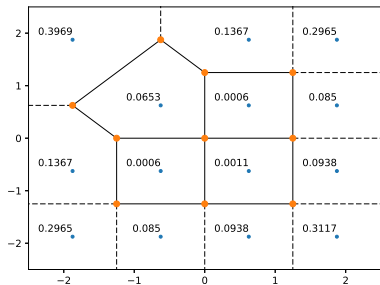
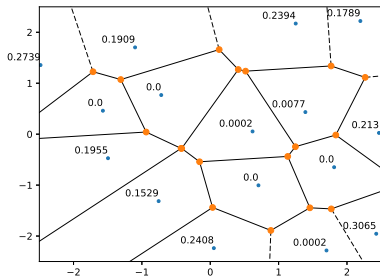


Table: Quality factor

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HCPP	1.49	3.27	4.73
Square Array <i>optimumlocation</i>	2.62	10.78	11.55



(a) Square array with optimum power allocation



(b) HCPP realization with optimum power allocation

Publications

- ① G. V. S. S. Praneeth Varma, Rayapati Sushma, Vandana Sharma, Abhinav Kumar, and G. V. V. Sharma, "Power allocation for uniform illumination with stochastic LED arrays," Optics Express **25**(8), 8659–8669 (2017).
- ② G. V. S. S. Praneeth Varma, Abhinav Kumar, and G. V. V. Sharma, "Resource Allocation for Visible Light Communication using Stochastic Geometry," IEEE Trans. Comm. (Submitted).
- ③ G. V. S. S. Praneeth Varma, Abhinav Kumar, and G. V. V. Sharma, "Resource Allocation for Visible Light Communication using Stochastic Geometry," Proc in IEEE COMSNETS. (Submitted).
- ④ G. V. S. S. Praneeth Varma, "Optimum Power Allocation for Uniform Illuminance in Visible Light Communication Systems," (In preparation).

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- Y. Chen, C. W. Sung, S.-W. Ho, and W. S. Wong, “Ber analysis for interfering visible light communication systems,” in International Symposium on Communication Systems , Networks and Digital Signal Processing 564–570 (2016).

Thank You

Proof for asymptotic BER

$$P_e = \mathbb{E}_{\Phi} \left[Q \left(\frac{RC_1 \sum_{i=1}^N V_i}{\sigma_0} \right) \right] \quad (1)$$

Jensens Inequality: If X is a random variable (RV) and f is a convex function, then ,

$$f(\mathbb{E}[X]) \leq \mathbb{E}[f(X)] \quad (2)$$

$$P_e \geq Q \left(\mathbb{E}_{\Phi} \left[\frac{RC_1 \sum_{i=1}^N V_i}{\sigma_0} \right] \right) \quad (3)$$

since $Q(\cdot)$ is convex.

Lemma

Consider random variables X and Y where Y either has no mass at 0 (discrete) or has support $[0, \infty)$. Then

$$\mathbb{E}[f(X, Y)] \approx f(\mathbb{E}[X], \mathbb{E}[Y]) \quad (4)$$

Corollary

$$\mathbb{E}\left[\frac{X}{Y}\right] \approx \frac{\mathbb{E}[X]}{\mathbb{E}[Y]} \quad (5)$$

$$\mathbb{E}[X^2] \approx (\mathbb{E}[X])^2 \quad (6)$$

Since $\sigma_0 > 0$, using (5) and (6) in (3),

$$\begin{aligned} Q\left(\mathbb{E}_{\Phi}\left[\frac{RC_1 \sum_{i=1}^N V_i}{\sigma_0}\right]\right) &\approx Q\left(\frac{RC_1 \sum_{i=1}^N \mathbb{E}_{\Phi}[V_i]}{\mathbb{E}_{\Phi}[\sigma_0]}\right) \\ &= Q\left(\frac{RC_1 \sum_{i=1}^N \mathbb{E}_{\Phi}[V_i]}{\sqrt{\mathbb{E}_{\Phi}[\sigma_0^2]}}\right) \end{aligned} \quad (7)$$

resulting in

$$P_e \gtrsim Q\left(\frac{RC_1 \sum_{i=1}^N \mathbb{E}_{\Phi}[V_i]}{\sqrt{\mathbb{E}_{\Phi}[\sigma_0^2]}}\right) \quad (8)$$

BER Analysis for Square BPP

$$\begin{aligned}\mathbb{E}_{\Phi} [r_i^{\alpha}] &= \int_{-\infty}^{\infty} r^{\alpha} f_{r_i}(r) dr \\&= \int_0^{R_i} r^{\alpha} \frac{2\pi r (1-p)^{N-i} p^{i-1}}{W \mathcal{B}(N-i+1, i)} dr \\&\quad + \int_{R_i}^{R_c} r^{\alpha} \frac{2(\pi - 4\theta) r (1-q)^{N-i} q^{i-1}}{W \mathcal{B}(N-i+1, i)} dr \\&= \mathcal{I}_1 + \mathcal{I}_2\end{aligned}$$

where

$$\begin{aligned}
 \mathcal{I}_1 &= \int_0^{R_i} r^\alpha \frac{2\pi r (1-p)^{N-i} p^{i-1}}{W \mathcal{B}(N-i+1, i)} dr \\
 &= \frac{2\pi}{W \mathcal{B}(N-i+1, i)} \sum_{k=0}^{N-i} \binom{N-i}{k} (-1)^k \left(\frac{\pi}{W}\right)^{k+i-1} \\
 &\quad \times \int_0^{R_i} r^{2(k+i+\alpha/2+1/2)} dr \\
 &= \frac{1}{\mathcal{B}(N-i+1, i)} \sum_{k=0}^{N-i} \frac{\binom{N-i}{k} (-1)^k}{k+i+\alpha/2} \left(\frac{\pi}{W}\right)^{k+i} R_i^{2(k+i)+\alpha} \quad (9)
 \end{aligned}$$

$$\begin{aligned}
 \mathcal{I}_2 &= \int_{R_i}^{R_c} r^\alpha \frac{2(\pi-4\theta) r (1-q)^{N-i} q^{i-1}}{W \mathcal{B}(N-i+1, i)} dr \\
 &= \frac{1}{\mathcal{B}(N-i+1, i)} \sum_{k=0}^{N-i} \frac{\binom{N-i}{k} (-1)^k}{W^{k+i}} R_i^{2(k+i)+\alpha-1} \\
 &\quad \times \int_0^{\frac{\pi}{4}} \frac{2(\pi-4\theta) (\pi-4\theta+2\sin(2\theta))^{k+i-1}}{\cos^{2(k+i)+\alpha-1}(\theta)} d\theta \quad (10)
 \end{aligned}$$

For simplifying the analysis, using the approximation

$$\sum_{j=1}^N \mathbb{E}_{\Phi} [r_j^{\alpha}] \approx \sum_{j=1}^N r_j^{\alpha} \quad (11)$$

in (1),

$$\mathbb{E}_{\Phi} [V_i] = \frac{1}{\sum_{j=1}^N \mathbb{E}_{\Phi} [r_j^{\alpha}]} \mathbb{E}_{\Phi} \left[\frac{r^{\alpha}}{\left(\sqrt{r^2 + h^2} \right)^{m+3}} \right] \quad (12)$$

$$= \frac{\mathcal{J}_1 + \mathcal{J}_2}{\sum_{j=1}^N \mathbb{E}_{\Phi} [r_j^{\alpha}]} \quad (13)$$

where

$$\begin{aligned}
 \mathcal{J}_1 &= \int_0^{R_i} \frac{r^\alpha}{\left(\sqrt{r^2 + h^2}\right)^{m+3}} \frac{2\pi r}{W} \frac{(1-p)^{N-i} p^{i-1}}{\mathcal{B}(N-i+1, i)} dr \\
 &= \frac{R_i^\alpha}{h^{m+3} \mathcal{B}(N-i+1, i)} \sum_{k=0}^{N-i} \binom{N-i}{k} (-1)^k \left(\frac{\pi R_i^2}{W}\right)^{k+i} \\
 &\quad \times \int_0^{R_i^2/h^2} \frac{t^{(k+i+\alpha/2)-1}}{(1+t)^{\frac{m+3}{2}}} dt
 \end{aligned} \tag{14}$$

after some algebra.

$$\int_0^u \frac{x^{\mu-1}}{(1+\beta x)^\nu} dx = \frac{u^\mu}{\mu} {}_2F_1(\nu, \mu; 1+\mu; -\beta u) \quad [|\arg(1+\beta u)| < \pi, \operatorname{Re} \mu > 0] \quad (15)$$

Substituting the above in (14),

$$\begin{aligned} \mathcal{J}_1 &= \frac{h^\alpha}{h^{m+3} \mathcal{B}(N-i+1, i)} \sum_{k=0}^{N-i} \binom{N-i}{k} (-1)^k \left(\frac{\pi h^2}{W} \right)^{k+i} \\ &\quad \times {}_2F_1\left(\frac{m+3}{2}, (k+i+\alpha/2), k+i+\alpha/2+1; -\frac{R_i^2}{h^2}\right) \end{aligned} \quad (16)$$

Similarly,

$$\begin{aligned} \mathcal{J}_2 &= \int_{R_i}^{R_c} \frac{r^\alpha}{(\sqrt{r^2+h^2})^{m+3}} \frac{2(\pi-4\theta)r}{W} \frac{(1-q)^{N-i} q^{i-1}}{\mathcal{B}(N-i+1, i)} dr \\ &= \frac{1}{\mathcal{B}(N-i+1, i)} \sum_{k=0}^{N-i} \frac{\binom{N-i}{k} (-1)^k}{W^{k+i}} \\ &\quad \times \int_0^{\pi/4} \frac{2(\pi-4\theta) R_i^{2(k+i)+\alpha-1}}{\cos^{2(k+i-2)+\alpha-m}(\theta)} \frac{(\pi-4\theta+2\sin(2\theta))^{k+i-1}}{(\sqrt{R_i^2+h^2\cos^2(\theta)})^{m+3}} d\theta \end{aligned} \quad (17)$$

BER analysis for circular BPP

The probability density function (PDF) of the distance of i^{th} nearest LED location from origin for circular BPP for $0 \leq r \leq R_c$ is given by

$$\begin{aligned}
 f_{r_i} &= \frac{2r}{R_c^2 B(i, N-i+1)} \left(\frac{r^2}{R_c^2} \right)^{i-1} \left(1 - \frac{r^2}{R_c^2} \right)^{N-i} \\
 \Rightarrow \mathbb{E}_\Phi [r_i^\alpha] &= \int_0^\infty r^\alpha f_{r_i}(r) dr \\
 &= \int_0^{R_c} \frac{2rr^\alpha}{R_c^2 B(N-i+1, i)} \left(\frac{r^2}{R_c^2} \right)^{i-1} \left(1 - \frac{r^2}{R_c^2} \right)^{N-i} dr \\
 &= \frac{1}{B(N-i+1, i)} \sum_{k=0}^{N-i} \binom{N-i}{k} (-1)^k \int_0^{R_c^2} \frac{t^{i+k+\alpha/2-1}}{R_c^{2(i+k)}} dt \\
 &= \frac{R_c^\alpha}{B(N-i+1, i)} \sum_{k=0}^{N-i} \frac{\binom{N-i}{k} (-1)^k}{\left(i+k+\frac{\alpha}{2}\right)} \quad (18)
 \end{aligned}$$

through a change of variables

$$\begin{aligned}
\mathbb{E}_\Phi [V_i] &= \frac{1}{\sum_{j=1}^N \mathbb{E}_\Phi [r_j^\alpha]} \mathbb{E}_\Phi \left[\frac{r^\alpha}{(\sqrt{r^2 + h^2})^{m+3}} \right] \\
&= \frac{1}{\sum_{j=1}^N \mathbb{E}_\Phi [r_j^\alpha]} \int_0^{R_c} \frac{2r}{B(N-i+1, i)} \sum_{k=0}^{N-i} \binom{N-i}{k} \frac{(-1)^k r^{2(i+k+\alpha/2-1)}}{R_c^{2(i+k)} (\sqrt{r^2 + h^2})^{m+3}} dr \\
&= \frac{1}{B(N-i+1, i) \sum_{j=1}^N \mathbb{E}_\Phi [r_j^\alpha]} \int_0^{R_c^2/h^2} \sum_{k=0}^{N-i} \frac{\binom{N-i}{k} (-1)^k h^2 (h^2 t)^{(i+k+\alpha/2-1)}}{R_c^{2(i+k)} h^{m+3} (\sqrt{1+t})^{m+3}} dt \\
&= \frac{1}{B(N-i+1, i) \sum_{j=1}^N \mathbb{E}_\Phi [r_j^\alpha]} \sum_{k=0}^{N-i} \binom{N-i}{k} \frac{(-1)^k h^{2(i+k+\alpha/2)}}{h^{m+3} R_c^{2(i+k)}} \\
&\quad \times \int_0^{R_c^2/h^2} \frac{t^{(i+k+\alpha/2-1)}}{(1+t)^{(m+3)/2}} dt \\
&= \frac{R_c^\alpha h^{-(m+3)}}{B(N-i+1, i) \sum_{j=1}^N \mathbb{E}_\Phi [r_j^\alpha]} \\
&\quad \times \sum_{k=0}^{N-i} \frac{\binom{N-i}{k} (-1)^k}{(i+k+\frac{\alpha}{2})} {}_2F_1 \left(\frac{m+3}{2}, i+k+\frac{\alpha}{2}; i+k+\frac{\alpha}{2}+1; -\frac{R_c^2}{h^2} \right) \quad (19)
\end{aligned}$$

Formulation of optimization problem

$$P_{r_j} = \sum_{i=1}^N H_{ij} P_{t_i} \quad (20)$$

$$\text{var}(P_{r_j}) = \mathbb{E} \left[(P_{r_j})^2 \right] - (\mathbb{E} [P_{r_j}])^2$$

$$\begin{aligned} (\mathbb{E} [P_{r_j}])^2 &= \left(\frac{\sum_{j=1}^K P_{r_j}}{K} \right)^2 \\ &= \left(\frac{\sum_{j=1}^K \sum_{i=1}^N H_{ij} P_{t_i}}{K} \right)^2 \\ &= \left(\frac{\sum_{i=1}^N \gamma_i P_{t_i}}{K} \right)^2 \quad \text{where } \gamma_i = \sum_{j=1}^K H_{ij} \\ &= \frac{\sum_{i=1}^N \gamma_i^2 P_{t_i}^2 + 2 \sum_{i=1}^N \sum_{p=i+1}^N \gamma_i \gamma_p P_{t_i} P_{t_p}}{K^2} \end{aligned} \quad (21)$$

$$\mathbb{E} \left[(P_{r_j})^2 \right] = \mathbb{E} \left[\left(\sum_{i=1}^N H_{ij} P_{t_i} \right)^2 \right] \quad (22)$$

$$\begin{aligned} &= \mathbb{E} \left[\sum_{i=1}^N H_{ij}^2 P_{t_i}^2 + 2 \sum_{i=1}^N \sum_{q=i+1}^N H_{ij} H_{qj} P_{t_i} P_{t_q} \right] \\ &= \frac{\sum_{j=1}^K \left[\sum_{i=1}^N H_{ij}^2 P_{t_i}^2 + 2 \sum_{i=1}^N \sum_{q=i+1}^N H_{ij} H_{qj} P_{t_i} P_{t_q} \right]}{K} \\ &= \frac{\sum_{i=1}^N \mu_{ii} P_{t_i}^2 + 2 \sum_{i=1}^N \sum_{q=i+1}^N \mu_{iq} P_{t_i} P_{t_q}}{K} \end{aligned}$$

$$\text{where } \mu_{iq} = \sum_{j=1}^K H_{ij} H_{qj} \quad (23)$$

$$\begin{aligned}
\text{var}(P_{r_j}) &= \frac{\sum_{i=1}^N \mu_{ii} P_{t_i}^2 + 2 \sum_{i=1}^N \sum_{q=i+1}^N \mu_{iq} P_{t_i} P_{t_q}}{K} \\
&\quad - \frac{\sum_{i=1}^N \gamma_i^2 P_{t_i}^2 + 2 \sum_{i=1}^N \sum_{p=i+1}^N \gamma_i \gamma_p P_{t_i} P_{t_p}}{K^2} \\
&= \sum_{i=1}^N \frac{\mu_{ii}}{K} - \frac{\gamma_i^2}{K^2} P_{t_i}^2 + 2 \sum_{i=1}^N \sum_{q=i+1}^N \frac{\mu_{iq}}{K} - \frac{\gamma_i \gamma_p}{K^2} P_{t_i} P_{t_q} \\
&= \frac{1}{2} \sum_{i=1}^N \left(\frac{2 \sum_{j=1}^K H_{ij}^2}{K} - \frac{2 \left(\sum_{j=1}^K H_{ij} \right)^2}{K^2} \right) P_{t_i}^2 \\
&\quad + 2 \sum_{u=1}^N \sum_{v=i+1}^N \left(\frac{2 \sum_{j=1}^K H_{uj} H_{vj}}{K} - \frac{2 \left(\sum_{j=1}^K H_{uj} \right) \left(\sum_{j=1}^K H_{vj} \right)}{K^2} \right) P_{t_u} P_{t_v} \\
&= \frac{1}{2} [P_{t_1}, \dots, P_{t_N}] \begin{bmatrix} \beta_{11} & \cdots & \beta_{1N} \\ \vdots & \ddots & \vdots \\ \beta_{N1} & \cdots & \beta_{NN} \end{bmatrix} \begin{bmatrix} P_{t_1} \\ \vdots \\ P_{t_N} \end{bmatrix} \\
&= \frac{1}{2} \mathbf{x}^T \mathcal{P}_X
\end{aligned} \tag{24}$$