Poisson Hole Process: Theory and Applications to Wireless Networks

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Abstract—Interference field in wireless networks is often modeled by a homogeneous Poisson Point Process (PPP). While it is realistic in modeling the inherent node irregularity and provides meaningful first-order results, it falls short in modeling the effect of interference management techniques, which typically introduce some form of spatial interaction among active transmitters. In some applications, such as cognitive radio and device-to-device networks, this interaction may result in the formation of holes in an otherwise homogeneous interference field. The resulting interference field can be accurately modeled as a Poisson Hole Process (PHP). Despite the importance of PHP in many applications, the exact characterization of interference experienced by a typical node in a PHP is not known. In this paper, we derive several tight upper and lower bounds on the Laplace transform in an otherwise homogeneous interference field. The resulting interference field is a PPP from which holes are carved out. When the locations of holes of a given radius are carved out. When the locations of holes also form an independent PPP, the resulting point process is usually termed as a PHP and Rayleigh fading.

Index Terms—Interference modeling, stochastic geometry, Poisson Point Process, Poisson Hole Process, coverage probability.

I. INTRODUCTION

The received signal-to-interference-plus-noise ratio (SINR) is known to be a strong indicator of the performance and reliability of a wireless link. Several key performance metrics of interest, such as outage probability, ergodic capacity, and outage capacity, are strongly dictated by the received SINR. By definition, the SINR distribution depends upon the joint distribution of the received powers from the serving node and the interfering nodes, which ultimately depend on the network topology. Therefore, accurate modeling of the network topology becomes a key step towards meaningful performance analysis of wireless networks. Owing to its tractability and realism in modeling irregular node locations, stochastic geometry has emerged as an important tool for the realistic analysis of wireless networks [2–5]. Initially popular for the modeling of wireless ad hoc and sensor networks, e.g. see [6], [7], it has recently been adopted for the analysis of cellular and heterogeneous cellular networks as well [8–10]. Irrespective of the nature of the wireless network, the interference field is almost always modeled by a homogeneous PPP to maintain tractability. While this leads to remarkably simple results for key performance metrics, such as coverage and rate, it is not quite suitable for modeling the effect of interference management techniques, which often introduce some form of spatial interaction among transmitters. In this paper, we focus on spatial separation, where holes (also called exclusion zones) are created around nodes/networks that need to be protected from excessive interference [11]. In particular, we assume that the baseline interference field is a PPP from which holes of a given radius are carved out. When the locations of the holes also form an independent PPP, the resulting point process is usually termed as a PHP, which is the main focus of this paper.

A. Related Work and Applications

We first discuss a few of possibly numerous instances in wireless networks where PHP is a more appropriate model for node locations. In particular, we discuss how PHP has been used to model cognitive radio networks, heterogeneous cellular networks, and device-to-device (D2D) networks. The main objective of cognitive radio networks is to improve spectrum utilization by allowing unlicensed secondary users to use licensed spectrum as long as they do not cause excessive interference to the licensed primary users. One of the ways to ensure this is by creating exclusion zones (holes) around primary users, where secondary transmissions are not allowed. This spatial separation was elegantly modeled by using a PHP in [12]. In particular, the locations of both primary and secondary users were first modeled by independent PPPs. Assuming secondary transmissions were not allowed within a given distance from the primary users, the locations of active secondary users were then modeled by a PHP.

The PHP has also been used recently to model inter-tier dependence in the base station locations in a heterogeneous cellular network in [13–15]. Modeling the macrocell locations by a PPP, it was assumed that the small cells are deployed farther than a minimum distance from the macrocells, i.e., outside exclusion zones of a given radius. In such a case, small cells form a PHP. This model introduces repulsion between the locations of macro and small cells, which is desirable due to several reasons, such as interference mitigation at macrocells due to small cell transmissions, and the higher advantage of
deploying small cells towards the cell edges of macrocells, especially in the coverage-centric deployments.

Similarly, for underlay D2D communication in cellular networks, inhibition zones may be created around cellular links where no D2D transmissions are allowed, thus saving cellular links from excessive D2D interference. The active D2D transmitters outside the holes form a PHP \[16, 17\]. In this regard, cognitive D2D communication in cellular network when transmitters are powered by harvesting energy from the ambient interference is studied in \[18\]. In \[19\], a Poisson Cluster Process (PCP) and a PHP are merged to develop a new spatial model for integrated D2D and cellular networks. In particular, a modified Thomas cluster process is used to model device locations where instead of modeling the cluster centers as a homogeneous PPP, they are modeled as a PHP to account for the inhibition zones around cellular links.

Despite the importance of a PHP in modeling wireless networks, the exact characterization of interference experienced by a typical receiver in a PHP is a challenging problem. There are two main directions taken in the literature for the analysis of wireless networks modeled by PPPs. The first approach, termed first-order statistic approximation, approximates PHP by a homogeneous PPP with the same density \[1\]. The second approach ignores the holes altogether to approximate the PHP by its baseline PPP. This overestimates the interference and the accuracy of the approximation is a function of system parameters \[12, 16-18\]. Besides, the PHP is sometimes approximated with a PCP by matching the first and second order statistics \[12-14\]. The resulting expressions for performance metrics are usually more complicated in this case compared to the above two. While all these approaches are reasonable, they are typically not accurate beyond a specific range of system parameters. In this paper, we take a fresh look at this problem and derive tight upper and lower bounds for the Laplace transform of interference experienced by the typical node in a PHP. The main contributions are summarized next.

### B. Contributions

**New approach to the analysis of a PHP:** Unlike existing approaches that approximate a PHP with either a PPP or a PCP, we develop a new approach that is amenable to shot-noise analysis and leads to tight provable bounds on the Laplace transform of interference experienced by a typical node of a PHP. A lower bound is first derived by overestimating interference by ignoring all the holes except the closest one. We provide an equivalent interpretation of this result in which the closest hole is dissolved in such a way that it results in a tractable non-homogeneous PPP. The resulting bound is shown to be remarkably tight. Extending this approach to multiple holes, we derive an upper bound on the Laplace transform of interference by carving out each hole separately without accounting for the overlaps between them. This leads to the removal of some points from the baseline PPP multiple times, thus underestimating the interference power experienced by the typical point. This bound is also shown to be remarkably tight across a variety of scenarios, including the ones in which the holes exhibit significant overlaps.

**Approaches to incorporate the effect of overlaps in the holes.** In the first set of bounds discussed above, we carefully circumvented the need for incorporating the effect of overlaps between holes. While these simple and easy-to-use bounds are tight, we also provide ideas for incorporating the overlaps between holes, which tighten these bounds even further. In the first result, we generalize the lower bound discussed above by considering two closest holes in the interference field while incorporating the exact effect of overlap between them. In the second result, instead of trying to incorporate the exact effect of overlaps, we propose a new procedure for bounding the overlap area, which allows us to derive a provable lower bound on the Laplace transform while considering multiple holes in the interference field. In the third and final result, we propose a new approach that allows to incorporate the mean effect of overlaps in the holes.

**New insights.** Our results concretely demonstrate that for accurate analysis of interference in a PHP, it is very important to preserve the local neighborhood around the typical point. For instance, we show that considering even a single hole in the interference field results in a tighter characterization of interference power at the typical point of a PHP compared to seemingly more refined prior approach of first-order statistic approximation in which the PHP is approximated by a PPP with the matching density. This is because by considering a single hole, the local neighborhood around the typical point is accurately captured, whereas it is distorted in the other approach due to independent thinning involved in the density matching of a PPP.

## II. Network Model

### A. System Model

We consider a wireless network that is modeled by a PHP in \(\mathbb{R}^2\). A PHP can be formally defined in terms of two independent homogeneous PPPs \(\Phi_1\) and \(\Phi_2\), where \(\Phi_2\) represents the baseline PPP from which the holes are carved out and \(\Phi_1\) represents the locations of the holes. Let the densities of \(\Phi_1\) and \(\Phi_2\) be \(\lambda_1\) and \(\lambda_2\), respectively, with \(\lambda_2 > \lambda_1\). Denoting the radius of each hole by \(D\), the region covered by the holes can be expressed as

\[
\Xi_D \triangleq \bigcup_{y \in \Phi_1} \mathbf{b}(y, D), \quad \mathbf{b}(y, D) \equiv \{ \mathbf{z} \in \mathbb{R}^2 : \| \mathbf{z} - y \| < D \}.
\]  

### TABLE I

| Symbol | Description |
|--------|-------------|
| \(\Phi_1\) | Independent PPP modeling the locations of hole centers; density of \(\Phi_1\) \(\lambda_1\) |
| \(\Phi_2\) | Independent PPP from which the holes are carved out; density of \(\Phi_2\) \(\lambda_2\) |
| \(\Psi\) | PHP formed by carving out holes with centers \(\Phi_1\) from \(\Phi_2\) |
| \(D\) | Radius of each hole carved out from \(\Phi_2\) |
| \(P_0\) | Transmit power; serving distance for the link of interest |
| \(\alpha\) | Path-loss exponent for all the wireless links |
| \(C\) | Ball of radius \(D\) centered at \(y\) |
| \(b_a\) | Fading gain; \(b_a \sim \exp(1)\) for Rayleigh fading |
| \(\gamma\) | Coverage probability (in terms of SIR); SIR threshold |
| \(\xi\) | Laplace transform of interference experienced by the typical node in a PHP |
| \(\Xi_D\) | Ball of radius \(D\) centered at \(y\) |
| \(\mathbf{b}(y, D)\) | Ball of radius \(D\) centered at \(y\) |
| \(\mathbf{z}\) | \(\mathbf{z} \in \mathbb{R}^2\) |
| \(\| \mathbf{z} - y \| \) | Euclidean distance between \(\mathbf{z}\) and \(y\) |
| \(\Xi\) | Region covered by the holes |
| \(\Xi\) | Union of all \(\Xi_D\) for all \(y \in \Phi_1\) |

\(\Xi_D\) denotes the region covered by holes of radius \(D\) centered at points \(y \in \Phi_1\). The region \(\Xi\) is the union of all such regions. The notation \(\Xi_D\) is used to denote the region covered by holes of radius \(D\) centered at points \(y \in \Phi_1\) and \(\Xi\) is the union of all such regions.

\[
\Xi_D \triangleq \bigcup_{y \in \Phi_1} \mathbf{b}(y, D), \quad \mathbf{b}(y, D) \equiv \{ \mathbf{z} \in \mathbb{R}^2 : \| \mathbf{z} - y \| < D \}.
\]
The points of $\Phi_2$ lying outside $\Xi_D$, form a PHP, which can be formally expressed as

$$\Psi = \{x \in \Phi_2 : x \notin \Xi_D\} = \Phi_2 \setminus \Xi_D.$$  \hfill (2)

It should be noted that the PHP $\Psi$ has also been known as a Hole-1 process in the literature [20].

We characterize the interference experienced by a typical node in $\Psi$ due to the transmission of the other nodes of $\Psi$. Due to the stationarity of the process, the typical node can be assumed to lie at the origin $o$, and due to Slivnyak’s theorem, we can condition on $o \in \Psi$ without changing the distribution of the rest of the process [5]. This equivalently means that the interferers are modeled by the PHP $\Psi$ with the typical receiver being an additional point placed at the origin. Note that Slivnyak’s theorem for a PPP is applicable here because conditioned on the locations of the holes (i.e., conditioned on $\Phi_1$), PHP $\Psi$ is simply a PPP of density $\lambda_2$ defined on $\mathbb{R}^2 \setminus \Xi_D$. Since the typical point is located outside the holes by construction, there are no points of $\Phi_1$ within a disk of radius $D$ around the typical point. For this receiver, we assume that the serving transmitter is located at the fixed distance $r_0$. It should be noted that we could have considered more sophisticated models for the serving link of interest but we chose to consider this simple setup because our emphasis is on characterizing interference in a PHP.

For the wireless channel between points $x$ and $y$, we consider a standard power law path-loss $l(x,y) = \|x-y\|^{-\alpha}$ with path-loss exponent $\alpha$. All the wireless links are assumed to experience independent Rayleigh fading. All the transmitters are assumed to transmit at a fixed power $P$. The received power at the typical node from its transmitter of interest is $P_{s}$ that the serving transmitter is located at the fixed distance $r_0$. It should be noted that we could have considered more sophisticated models for the serving link of interest but we chose to consider this simple setup because our emphasis is on characterizing interference in a PHP.

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$$I = \sum_{x \in \Psi} Ph_x |x|^{-\alpha},$$ \hfill (3)

where $h_x \sim \exp(1)$ models Rayleigh fading gain for the link from interferer $x \in \Psi$ to the typical receiver. For this setup, we define coverage probability next.

B. SIR and Coverage Probability

Using the received power over the link of interest and the interference power defined in the previous subsection, the signal to interference ratio (SIR) can be expressed as

$$\text{SIR}(r_0) = \frac{Ph_{s}r_0^{-\alpha}}{\sum_{x \in \Xi} Ph_x |x|^{-\alpha}}.$$ \hfill (4)

Denote the minimum SIR required for successful decoding and demodulation at the typical receiver by $\gamma$. A useful metric of interest in wireless networks is the SIR coverage probability $P_c$, which is the probability that the SIR at the receiver exceeds the threshold $\gamma$. Mathematically,

$$P_c = \mathbb{P}\{\text{SIR}(r_0) > \gamma\} = \mathbb{P}\left\{h > \frac{\gamma r_0^\alpha}{P} I \right\} = (a) \mathbb{E}\left[\exp\left(-\frac{\gamma r_0^\alpha}{P} I\right)\right] = (b) \mathcal{L}_I\left(\frac{\gamma r_0^\alpha}{P}\right),$$ \hfill (5)

where $(a)$ follows from the fact that $h \sim \exp(1)$, and $(b)$ from the definition of Laplace transform of interference power $\mathcal{L}_I(s) = \mathbb{E}[\exp(-sI)]$. Note that for this setup, it is sufficient to focus on the Laplace transform of interference in order to study coverage probability. In general, accurate characterization of $\mathcal{L}_I(s)$ is the first step in the analysis of more general classes of wireless networks, including cellular networks [9]. Therefore, we will focus on $\mathcal{L}_I(s)$ in the technical sections of the paper with the understanding that the coverage probability can be easily derived for our setup using (5). We begin our technical discussion by summarizing two key prior approaches used in the literature for characterizing $\mathcal{L}_I(s)$ in a PHP.

For the ease of reference, the notation used in the paper is summarized in Table I.

### III. Key Prior Approaches

In this section, we summarize two popular approaches that have been used in the literature to derive the Laplace transform of interference in a PHP. At the end of the Section, we also provide insights into the strengths and weaknesses of each approach.

A. Lower Bound by Ignoring Holes: Approximating $\Psi$ by $\Phi_2$

The first approach is to ignore the effect of holes and approximate the interference field $\Psi$ by the baseline PPP $\Phi_2$ of density $\lambda_2$. By construction, this approach overestimates the interference power and hence leads to the lower bound on the Laplace transform of interference [12]. This well-known result is stated below for completeness.

**Lemma 1** (Lower bound). Ignoring the impact of holes (approximating $\Psi$ by $\Phi_2$), the Laplace transform of aggregate interference $I = \sum_{x \in \Xi} Ph_x |x|^{-\alpha}$ is lower bounded by:

$$\mathcal{L}_I(s) \geq \mathbb{E}\left[e^{-\pi \lambda_2 (sP)^{2/\alpha} \text{sin}(2/\alpha)}\right].$$ \hfill (6)

**Proof.** See Appendix A

The tightness of the above bound will be demonstrated in the Numerical Results section.

B. Approximating PHP by a PPP with the Same Density

The second approach to the derivation of the Laplace transform of PHP is the first-order statistic approximation [3]. In this approach, the baseline PPP $\Phi_2$ is independently thinned such that the resulting density of the PPP is the same as that of the PHP $\Psi$, which we denote by $\lambda_{\text{PHP}}$. The first step in this approach is therefore to derive $\lambda_{\text{PHP}}$ in terms of the given system parameters, which was done in [3]. For completeness, we discuss its proof briefly below. To derive $\lambda_{\text{PHP}}$, we first need to derive an expression for the average number of points of the PHP $\Psi$ lying in a given set $B \subset \mathbb{R}^2$, which by definition is

$$\mathbb{E}\left[\sum_{x \in \Phi_2 \cap B} \prod_{y \in \Phi_1} (1 - b(x,y))\right]$$

where $b(x,y)$ is the indicator function of the region where the typical node is located.
where (a) is due to the independence of point processes \( \Phi_1 \) and \( \Phi_2 \), (b) follows from the probability generating functional (PGFL) of a PPP, and (c) follows from the Campbell theorem [2]. From the above expression, we can readily infer that \( \lambda_{\text{PHP}} = \lambda_2 \exp(-\lambda_1 \pi D^2) \). Now to derive the Laplace transform of interference in this case, we just need to replace \( \lambda_2 \) in the result of Lemma [1] with \( \lambda_{\text{PHP}} \). The result is stated below for completeness.

**Lemma 2 (Approximation).** The Laplace transform of interference power \( I = \sum_{x \in \Psi} Ph_x ||x||^{-\alpha} \) when PHP \( \Psi \) is approximated by a PPP with density \( \lambda_{\text{PHP}} \) is

\[
\mathcal{L}_I(s) \approx \exp \left( -\pi \lambda_{\text{PHP}} \frac{(sP)^{2/\alpha}}{\text{sinc}(2/\alpha)} \right).
\]

**Remark 1.** Both the approaches discussed above approximate \( \Psi \) with a homogeneous PPP: the one first with density \( \lambda_2 \) (the baseline PPP), and the second one with density \( \lambda_{\text{PHP}} < \lambda_2 \) (the density of the PHP \( \Psi \)). While the second approach is a seemingly more refined approach, a careful thought reveals that in order to match the density of the PPP with that of a PHP, the baseline PPP has to be independently thinned, which disturbs the local neighborhood of points around the typical point, thus resulting in a loose bound. On the other hand, approximating \( \Psi \) simply by the baseline PPP \( \Phi_2 \) preserves the local neighborhood resulting in a relatively tighter approximation. More insights will be provided in the numerical results section.

Before concluding this section, it is important to note that there is one more fitting-based approach in which the PHP is approximated by a PCP with matching first and second order statistics. The Laplace transform of interference and other performance metrics are then studied using the fitted PCP. Since the formal description of this technique requires significantly more details than the techniques discussed above, we refer the reader to [12]–[14], where this approach has been used for the analysis of cognitive radio and heterogeneous cellular networks modeled as PHPs. We now discuss the proposed approaches next.

**IV. PROPOSED APPROACHES TO LAPLACE TRANSFORM OF INTERFERENCE IN A PHP**

We now introduce our proposed approach to characterize the Laplace transform of interference in a PHP. In the first intermediate step, we model the locations of interferers by a homogeneous PPP \( \Phi_2 \) of density \( \lambda_2 \) from which only one hole \( C \) of radius \( D \) is carved out at a deterministic location. Let the location of the center of this hole be \( y \in \mathbb{R} \) and hence its distance from the origin be \( ||y|| \). The resulting setup is illustrated in Fig. 1. Note that the interference field in this case is non-isotropic due to the fixed location of the hole. The Laplace transform of the interference power at the origin from the nodes of \( \Phi_2 \) outside \( C \) is characterized next. This intermediate result will be used later in this section to derive upper and lower bounds on the Laplace transform of interference experienced by a typical node in a PHP.

**Lemma 3.** Let \( I = \sum_{x \in \Phi_2 \setminus B(y, D)} Ph_x ||x||^{-\alpha} \), the Laplace transform of interference conditioned on \( ||y|| \) is

\[
\mathcal{L}_I(||y||)(s) = \exp \left( -\pi \lambda_2 \frac{(sP)^{2/\alpha}}{\text{sinc}(2/\alpha)} \right) \times \exp \left( \frac{(||y||+D)^2}{||y|| D^2} - \frac{2\pi \lambda(r)}{1 + \frac{r}{\pi}} \right),
\]

where \( \lambda(r) = \frac{\lambda_2}{\pi} \arccos \left( \frac{r^2 + ||y||^2 - D^2}{2||y|| r} \right) \), \( C = b(y, D) \) denotes the hole centered at \( y \) with radius \( D \).

**Proof.** See Appendix B.

**Remark 2 (Dissolving the hole).** The above result has an interesting interpretation that will be useful in visualizing the proposed results. Note that since received power is a radially symmetric function, it solely depends upon the distance of the transmitter to the origin. Therefore, we can in principle, dissolve the hole as long as the number of points lying in a thin strip of radius \( ||y|| - D \leq r \leq ||y|| + D \) and vanishingly small width \( dr \) is not changed. Please refer to Fig. 7 for an illustration of this strip. Taking a closer look at the interference originating from this strip, we note that the only thing that matters is the number of points distributed in the part of the strip which is outside the hole. The area of this region is \( 2\pi r dr (\pi - \theta(r)) \), where the angle \( \theta(r) = \arccos \left( \frac{r^2 + ||y||^2 - D^2}{2||y|| r} \right) \) is defined in Fig. 7. Therefore, the number of interfering points lying within this strip is Poisson distributed with mean \( \lambda_2 2\pi r dr (\pi - \theta(r)) \). Since the exact locations of these points within the strip doesn’t matter, we can dissolve the hole and redistribute the points uniformly inside the whole strip of area \( 2\pi r dr \). This means, the PPP with a hole can be equivalently modeled as a non-homogeneous PPP with density \( \lambda_2 (1 - \theta(r)/\pi) \), where the \( \lambda_2 \theta(r)/\pi \) term (defined as \( \lambda(r) \) in Lemma 3) captures the effect of hole.
Fig. 2. Illustration showing that the closest point of \( \Phi \) is at least a distance \( D \) away from the typical point of the PHP \( \Psi \).

Using this intermediate result and the above insights, we now derive tight bounds on the Laplace transform of interference experienced by a typical node in a PHP.

A. Lower Bound on the Laplace Transform of Interference in a PHP

Before going into the technical details, note that due to path-loss, the effect of holes that are close to the typical point will be much more significant compared to the holes that are farther away. Therefore, to derive an easy-to-use lower bound on the Laplace transform of interference, we consider only one hole; the one that is closest to the typical point; and ignore the other holes. Denoting the location of the closest hole by \( y_1 \), the interference field in this case is \( (\Phi_2 \cap b^c(y_1, D)) \), which clearly overestimates the interference of PHP and hence leads to a lower bound on the Laplace transform. Note that in Lemma 3 we have already derived the conditional Laplace transform for the case when there is one hole and its distance to the origin is known. To derive a lower bound, we simply need to assume this hole to be the closest point of \( \Phi_1 \) to the origin and decondition the result of Lemma 3 with respect to the distribution of \( V_1 = \|y_1\| \). To this end, we first derive the probability density function (PDF) of \( V_1 \) next.

**Lemma 4.** The PDF of the distance \( V_1 = \|y_1\| \) between the typical node at the origin and the closest point of \( \Phi_1 \) is given by

\[
f_{V_1}(v_1) = 2\pi\lambda_1 v_1 \exp(-\pi\lambda_1 (v_1^2 - D^2)), \quad v_1 \geq D.
\]

**Proof.** As discussed in Section II, the typical point of a PHP lies outside the holes by construction. Therefore, as illustrated in Fig. 2, the minimum distance between the typical point and the closest hole (closest point of \( \Phi_1 \)) is \( D \). Using this fact along with the properties of a PPP, the distribution of \( V_1 \) can be derived as follows:

\[
\mathbb{P}(V_1 > v_1) = \mathbb{P}[(\text{Number of points of } \Phi_1 \text{ in the set } \{b(0, v_1) \setminus b(0, D)\} = 0) = \exp(-\pi\lambda_1 (v_1^2 - D^2)), \quad v_1 \geq D.
\]

The result now follows by differentiating the above expression.

Deconditioning the result of Lemma 3 with respect to the distribution of the distance to the closest hole derived above, the proposed lower bound is derived below.

**Theorem 1** (New Lower Bound 1). Let \( I = \sum_{\mathbf{x} \in \Psi} P_{h_\mathbf{x}}||x||^{-\alpha} \), the Laplace transform of interference is lower bounded by

\[
L_1(s) \geq \exp \left( -\pi \lambda_2 \frac{(sP)^{2/\alpha}}{\text{sinc}(2/\alpha)} \times \int_D^\infty \exp(g(v_1)) 2\pi \lambda_1 v_1 \exp(-\pi\lambda_1 (v_1^2 - D^2)) dv_1 \right), \tag{10}
\]

where \( g(v_1) = \int_{v_1-D}^{v_1+D} \arccos \left( \frac{r^2 + v_1^2 - D^2}{2rv_1} \right) \frac{2\lambda_1}{1+\frac{r^2}{2}} dr \).

**Proof.** See Appendix \( \square \)

**Remark 3.** Since the PHP is approximated by \( \Phi_2 \cap b^c(y_1, D) \) in the above result, the resulting lower bound presented in Theorem 2 is by construction tighter than the known lower bound presented in Lemma 1 where the interference field was approximated by simply \( \Phi_2 \). On the same lines as discussed for Lemma 3 in Remark 2 the above approach captures the local neighborhood of the typical node accurately, thus leading to a remarkably tight lower bound in Theorem 7. This will be demonstrated later in this section and in the numerical results section.

B. Upper Bound for the Laplace Transform of Interference in a PHP

To derive an upper bound on the Laplace transform, we extend the above approach to all the holes. To maintain tractability, each hole is carved out individually/separately from the baseline PPP \( \Phi_2 \) using the above approach. Note that since the centers of the holes follow a PPP \( \Phi_1 \), there will obviously be overlaps among holes. Therefore, when we remove points of \( \Phi_2 \) corresponding to each hole individually (without accounting for the overlaps), we may remove certain points multiple times thus underestimating the interference field, which results in an upper bound on the Laplace transform of interference. In the next section, we show that any reasonable attempt towards incorporating the exact effect of overlaps leads to a significant loss in tractability. Fortunately, the bounds derived in this section without incorporating the effect of overlaps are remarkably tight and can be considered proxies for the exact Laplace transform.

**Theorem 2** (New Upper Bound). The Laplace transform of interference experienced by a typical node in a PHP is upper bounded by

\[
L_I(s) \leq \exp \left( -\pi \lambda_2 \frac{(sP)^{2/\alpha}}{\text{sinc}(2/\alpha)} \times \int_D^\infty (1 - \exp(f(v))) dv \right), \tag{11}
\]

where \( f(v) = \int_{v-D}^{v+D} \arccos \left( \frac{r^2 + v_1^2 - D^2}{2rv_1} \right) \frac{2\lambda_1}{1+\frac{r^2}{2}} dr \).

**Proof.** See Appendix \( \square \)
Further, bounds, given by Theorem 2 and Theorem 1, respectively.

where derived in this section.

the effect of overlaps in the holes that was ignored in the results section. Next, we explore a few ways to incorporate the effect of overlaps in the holes accurately. In this section, we address this question by deriving three results for the Laplace transform of interference that incorporate the effect of holes. In the first result, we generalize the lower bound derived in Theorem 1 by considering two nearest holes instead of a single hole. The overlap among the two holes is explicitly incorporated in the analysis. The setup is presented in Fig. 3 where $C_1 = b(y_1, D)$ and $C_2 = b(y_2, D)$ denote the first and the second closest holes to the typical receiver, respectively. The angle between $y_1$ and $y_2$ is denoted by $\phi$. The interference field in this case is modeled by $(\Phi_2 \cap \{C_1 \cup C_2\}) \supset \Psi$, which overestimates the interference power and hence leads to a lower bound on the Laplace transform of interference. Before going into the main result, we first need to evaluate the joint PDF of the distances between the first and second closest holes to the origin, which are denoted by random variables $V_1$ and $V_2$, respectively. Using the same arguments as in Lemma 4 the joint PDF can be derived as

$$f_{V_1V_2}(v_1, v_2) = f_{V_2}(v_2|v_1)f_{V_1}(v_1)$$

$$= (2\pi \lambda_1)^2 v_1v_2 \exp(-\pi \lambda_1(v_2^2 - D^2)). \quad (13)$$

Using this distribution, the Laplace transform of interference for this case is derived next.

**Theorem 3 (New Lower Bound 2).** The Laplace transform of interference experienced by a typical node of a PHP $\Psi$ is lower bounded by

$$\mathcal{L}_1(s) \geq \exp\left(-\pi \lambda_2 \left(\frac{\pi P}{\sin^2(\alpha)}\right)^{2/\alpha} \right) \times \left(\frac{1}{2\pi} \int_{v_{\min}}^{v_{\max}} \int_{-\pi}^{\pi} \exp\left(\int_{v_{\min}}^{v_{\max}} 2\pi \lambda_2 r dr \right) \exp(-\pi \lambda_2 B(v_1, v_2, \phi)) \times f_{V_1V_2}(v_1, v_2) d\phi dv_1 dv_2 \right) \quad (14)$$

where $\lambda_2(r) = \frac{\lambda_2}{\alpha} \arccos\left(\frac{r^2 + v_1^2 - D^2}{2vr}\right)$, for $i = 1, 2$ and $B(v_1, v_2, \phi)$ is given by (27).

**Proof.** The key idea behind this proof is to consider $(\Phi_2 \cap \{b(y_1, D) \cup b(y_2, D)\}) \supset \Psi$ as the interference field, which clearly overestimates the interference and hence leads to the lower bound. Complete proof is provided in Appendix B.

As evident from the above result, incorporating the effect of overlap, even among two holes, results in a significantly more complex expression compared to the bounds presented in Theorems 1 and 2. This shows that incorporating the exact effect of overlaps does indeed lead to a significant loss in tractability. Therefore, instead of trying to incorporate the exact effect of overlaps, we now propose a new procedure for bounding the overlap area, which allows us to derive a lower bound on the Laplace transform while considering multiple holes in the interference field. This tightens the result of Theorem 1 where only one hole was considered. In particular, we consider $k$ closest holes from the typical field, as shown in Fig. 4. In order
to claim the result as a bound, we bound the union of $k$-closest holes, i.e., $\bigcup_{i=1}^{k} C_i = \bigcup_{i=1}^{k} b(y_i, D)$, with $\Omega_0 = C_1 \cup_{i=2}^{k} \{ b(y_i, D) \bigcap b^c(0, \max (\|y_{i-1}\| + D, \|y_i\| - D)) \}$, where $\Omega_0 \subset \bigcup_{i=1}^{k} C_i$. The set $\Omega_0$ carefully avoids the overlapping part and hence does not result in over-removal of the points from the baseline point process $\Phi_2$. The contribution of $i$th hole in $\Omega_0$ is $d(y_i, D) = \{ b(y_i, D) \bigcap b^c(0, \max (\|y_{i-1}\| + D, \|y_i\| - D)) \}$. (15)

In Fig. 4 the set $d(y_i, D)$ corresponds to the unshaded part of a hole that does not overlap with the other holes. Clearly, this approach results in the removal of less points from the baseline process $\Phi_2$ compared to a PHP, which leads to a lower bound on the Laplace transform of interference in a PHP. This lower bound is presented in the next theorem.

**Theorem 4** (New Lower Bound 3). The Laplace transform of interference is bounded by

$$L_4(s) \geq \exp \left( -\pi \lambda_2 \left( \frac{sP}{\sin(2/\alpha)} \right)^{2/\alpha} \right) \times \left( \int \cdots \int_{D \leq v_1 < v_2 < \ldots < v_k < \infty} \exp \left( \int_{v_1-D}^{v_1+D} 2\pi \lambda_2(r) \frac{P}{\frac{1}{1+\frac{r}{\pi D}}} r dr \right) \exp \left( \frac{2\pi \lambda_2(r)}{1+\frac{r}{\pi D}} r dr \right) \ldots \exp \left( \int_{\max(v_k-D, v_1+D)}^{v_k+D} 2\pi \lambda_k(r) \frac{P}{\frac{1}{1+\frac{r}{\pi D}}} r dr \right) \times f_{V_1, V_2, \ldots, V_k}(v_1, v_2, \ldots, v_k) dv_1 dv_2 \ldots dv_k \right)$$

(16)

where $\lambda_i(r) = \frac{\lambda_2}{\pi} \arccos \left( \frac{r^2 + \pi^2 - D^2}{2rv_i} \right)$ for $i = 1, 2, \ldots, k$, and joint density function of distances of $V_1, V_2, \ldots, V_k$ is $f_{V_1, V_2, \ldots, V_k}(v_1, v_2, \ldots, v_k) = (2\pi \lambda_1)^k v_1 v_2 \ldots v_k \exp(-\pi \lambda_1 (v_1^2 - D^2))$.

**Proof.** See Appendix C.

In the numerical results section, we will show that considering $k = 2$, i.e., only the two closest holes from the typical point, results in a remarkably tight bound. Note that for $k = 2$, the expression is also fairly tractable. For brevity, that expression is not stated separately.

Now, in Theorems 3, 4, we have tried to handle the overlaps in such a way that the resulting expressions: (i) can be claimed as lower bounds to the Laplace transform, and (ii) tighten the lower bound provided by Theorem 1. One last question that we address before concluding this section is whether it is possible to handle the effect of overlaps in the average sense. For this, we revisit the upper bound of Theorem 2 which was derived by carving out holes from the baseline process $\Phi_2$ individually without caring about the overlaps between them. This led to the possible removal of some points multiple times, thereby leading to an underestimation of interference. We compensate this over-removal, by rescaling the second term of Theorem 2, which is the one that captures the effect of removing points from $\Phi_2$. The rescaling term is derived by estimating the average pairwise overlap area between circles. More details of the approach are provided in the proof in Appendix D. Note that the resulting expression in this case is the same as that of Theorem 2 except a scaling factor of $1 - \min(\frac{\lambda_1 D^2}{2}, \frac{1}{2})$ that appears in the second term. The key downside to this approach compared to Theorem 2 is that it results in an approximation unlike Theorem 2 where the result was shown to be a bound.

**Proposition 2** (New Approximation). The Laplace transform of interference at a typical point in a PHP can be approximated as $L_2(s) \approx \exp \left( -\pi \lambda_2 \left( \frac{sP}{\sin(2/\alpha)} \right)^{2/\alpha} \right) \exp \left( -2\pi \lambda_1 \int_{D}^{\infty} \left( 1 - \exp \left( f(v) \times \left( 1 - \min \left( \frac{\lambda_1 P D^2}{2}, \frac{1}{2} \right) \right) \right) \right) dv \right)$

(17)

where $f(v) = \int_{v-D}^{v+D} \arccos \left( \frac{v^2 + \pi^2 - D^2}{2rv} \right) \frac{2\lambda_2}{1+\frac{v}{\pi D}} v dv$.

**Proof.** See Appendix D.

Comparison of this result with the bounds derived in this paper shows that handling overlaps in the average sense may not work particularly well, especially when the overlaps are significant. This is because such results do not capture the local neighborhood of the typical point as carefully as the bounds derived in this paper. We now move on the numerical results section, where more such insights about the relative accuracy of bounds and approximations are presented.

**VI. NUMERICAL RESULTS AND DISCUSSION**

There are three main system parameters that determine the interference experienced by the typical point in a PHP: density $\lambda_1$ of the holes, density $\lambda_2$ of the baseline PPP, and the radii $D$ of the holes. Based on the relative values of these parameters, we identify four main network configurations, which are illustrated in Fig. 5. We define the possible configurations as LD-SH: configuration with low density of holes and small holes; HD-SH: configuration with high density of holes and small holes; LD-LH: configuration with low density of holes and
Fig. 5. The PHP network model (a) First configuration: LD-SH, (b) Second configuration: HD-SH, (c) Third configuration: LD-LH, (d) Fourth configuration: HD-LH.

Fig. 6. Ratio of the proposed upper and lower bounds derived in Theorems 2 and 1, respectively.

Fig. 7. Analytical and simulation results for the coverage probability as a function of the density \( \lambda_1 \) (\( D = 1 \)).

Fig. 8. Analytical and simulation results for the coverage probability as a function of hole radius \( D \) (\( \lambda_1 = 0.1 \)).

Fig. 9. Analytical and simulation results for the coverage probability in LD-SH case (\( \lambda_1 = 0.05 \) and \( D = 0.6 \)).

large holes; HD-LH: configuration with high density of holes and large holes. Clearly, the configuration where the holes are small and sparse (LD-SH case) is more benign than the configuration in which the holes are small and dense (HD-SH). Similarly, the configuration where holes are large and sparse (LD-LH case) is more benign than the configuration where the holes are both large and dense (HD-LH case). Therefore, the result that works well in the HD-SH configuration is expected to work in the LD-SH configuration as well. The same is true for LD-LH and HD-LH cases.
Simulations are performed over circular region with radius 40 m and results are averaged over $5 \times 10^4$ iterations. Unless otherwise specified, we set the network parameters as follows: $\lambda_2 = 1$, $\alpha = 4$, $P = 1$, $r_0 = 0.1$, $\gamma = 10$ dB. We compare our proposed bounds and the new approximation with the first-order statistic approximation given by Lemma 2 and the PPP-based bound given by Lemma 7 where $\Psi$ is approximated by $\Phi_2$. Before going into more details comparisons, we demonstrate the tightness of the lower and upper bounds derived in Theorems 1 and 2 by plotting their ratio and its approximation (derived in Proposition 1) in Fig. 6. In addition to validating the tightness of the approximation given by Proposition 1, this result shows that the ratio in all cases of interest is close to one, which demonstrates the tightness of both the bounds. Note that, as expected, the ratio is comparatively higher when the holes are large and dense (HD-LH case).

We now compare the proposed bounds and approximations with the numerical results and the known approaches in terms of coverage probability. As demonstrated in [5], the coverage probability for our setup is simply the Laplace transform of the interference evaluated at $s = \frac{\gamma_0^2}{P}$. Note that when we substitute $s = \frac{\gamma_0^2}{P}$ in any of the Laplace transform expressions derived in this paper, we notice that the resulting expression is independent of $P$. This is expected, because the $\text{SIR}(r_0)$ expression given by [4] is indeed independent of $P$.

In Fig. 7 we plot the coverage probability of a typical receiver as a function of hole density $\lambda_1$ assuming all other parameters are fixed (we assume $D = 1$). Small values of $\lambda_1$ result in LD-SH configuration, whereas high values result in the HD-SH configuration. In Fig. 8 we conduct the same study but instead of varying $\lambda_1$, we now vary the hole radius $D$ while fixing $\lambda_1 = 0.1$. The low values of $D$ result in HD-SH configuration, whereas the high values result in HD-LH configuration. Comparison of the proposed results with the simulations reveal that all the bounds and approximations proposed in this paper are surprisingly tight, even for the extreme configuration like HD-LH, where the overlaps are significant. The lower bounds work even better. We also notice that the prior results (given by Lemmas 1 and 2) deviate significantly when the overlaps in the holes are significant. In particular, the approximation given by Lemma 2 becomes very loose. As discussed in Remark 1 this is because its derivation involved independent thinning of the interference field, which distorts the local neighborhood of the typical receiver.

We now plot the analytical and simulation results for the coverage probability of the typical receiver as a function of the SIR threshold in the four possible configurations in Figs. 9–12. Fig. 9 shows the results for LD-SH case with small holes of radius $D = 0.6$ and low density of $\lambda_1 = 0.05$, while results for the HD-SH case are shown in Fig. 10 with parameters $D = 0.6$ and $\lambda_1 = 0.2$. Further, Fig. 11 shows results for the LD-LH case with large holes of radius $D = 1.5$ and low density of $\lambda_1 = 0.05$, while the results for the HD-LH case with parameters $D = 1.5$ and $\lambda_1 = 0.2$ are shown in Fig. 12. The plots again confirm the accuracy of our results and show that the first-order statistic approximation given by Lemma 2 is rather loose while our proposed bounds lead to tight upper and lower bounds in all cases. Interestingly, all the proposed results work so well that they are nearly

![Fig. 10. Analytical and simulation results for the coverage probability in HD-SH case ($\lambda_1 = 0.2$ and $D = 0.6$).](image1)

![Fig. 11. Analytical and simulation results for the coverage probability in LD-LH case ($\lambda_1 = 0.05$ and $D = 1.5$).](image2)

![Fig. 12. Analytical and simulation results for the coverage probability in HD-LH case ($\lambda_1 = 0.2$ and $D = 1.5$).](image3)
Fig. 13. Analytical and simulation results for the coverage probabilities of the PHP users as a function of $\lambda_2/\lambda_1$ under configuration LD-SH ($\lambda_1 = 0.05$ and $D = 0.6$).

Fig. 14. Analytical and simulation results for the coverage probabilities of the PHP users as a function of $\lambda_2/\lambda_1$ under configuration HD-SH ($\lambda_1 = 0.2$ and $D = 0.6$).

Fig. 15. Analytical and simulation results for the coverage probabilities of the PHP users as a function of $\lambda_2/\lambda_1$ under configuration LD-LH ($\lambda_1 = 0.05$ and $D = 1.5$).

Fig. 16. Analytical and simulation results for the coverage probabilities of the PHP users as a function of $\lambda_2/\lambda_1$ under configuration HD-LH ($\lambda_1 = 0.2$ and $D = 1.5$).

Figs. 15 and 16 depict results for the LD-LH and HD-LH cases. Here we consider $D = 1.5$ and $\lambda_1 = \{0.05, 0.2\}$. As was the case in the above results, our proposed lower and upper bounds provide a remarkably accurate characterization of coverage probability. This is because the local neighborhood of the typical node is carefully preserved while deriving these bounds. On the other hand, the prior results, in particular the first-order statistic approximation, leads to a fairly loose result.

VII. CONCLUSIONS

In this paper, we have focused on the accurate performance characterization of a typical user in a wireless network that is modeled as a PHP. This model is of particular interest in scenarios where interference management techniques introduce spatial separation among active transmitters in the form of holes or exclusion zones. In terms of the technical results, we have provided new easy-to-use provable lower and upper bounds on the Laplace transform of interference experienced...
by a typical user in a PHP. In addition to accurately characterizing the interference power, these bounds immediately characterize the coverage probability of a typical user in the case where all the wireless links experience independent Rayleigh fading. Since the prior work has mostly focused on reducing the PHP to a PPP either by ignoring the holes or by matching the PPP density to that of a PHP, to the best of our knowledge, the proposed bounds are the tightest known bounds for the Laplace transform of interference in a PHP.

For the analysis, we propose a new approach in which the holes are dissolved in such a way that a PHP is reduced to an equivalent (and more tractable) non-homogeneous PPP. The key in deriving tight bounds was to preserve the local neighborhood around the typical point while simplifying the far field to attain tractability. The tightness of the bounds is demonstrated analytically as well as numerically by comparing with simulations and known approaches. These results have numerous applications in a variety of wireless networks where interference management is performed by spatially separating the active links, such as in cognitive radio and D2D networks.

Since our main emphasis was on characterization of the interference power, we assumed that the serving transmitter for the typical receiver is located at a fixed distance. This corresponds to an ad hoc network scenario. Relaxation and generalization of this assumption is a fruitful area of future investigation. For instance, if the receiver of interest is a randomly chosen point in $\mathbb{R}^2$ and its serving transmitter is its closest point in a PHP, this setup can be used to study the performance of a cellular network modeled as a PHP. Another direction of future work is the extension of the current model to study exclusion zones with different shapes and sized, e.g., circles with different radii. Finally, the holes in a PHP are driven by a PPP. Extending this to other point processes, such as a Matérn process, is another promising direction of future work.

APPENDIX

A. Proof of Lemma 7

In this case, the PHP $\Psi$ is approximated by the baseline PPP $\Phi_2$, which reduces the problem to the well-studied problem of deriving Laplace transform of interference originating from the PPP field of interferers [3]. For completeness, the sketch of the derivation is provided next.

$$L_I(||y||)(s) = \mathbb{E} \left[ \exp \left( -s \sum_{x \in \Psi} Ph_x ||x||^{-\alpha} \right) \right]$$

\[= \exp \left( -\lambda \int_{\mathbb{R}^2} 1 - \mathbb{E}_{h_\Psi} \left\{ \exp(-sPh_x ||x||^{-\alpha}) \right\} dx \right) \]

\[= \exp \left( -\lambda \int_{\mathbb{R}^2} 1 - \mathbb{E}_{h_\Psi} \left\{ \exp(-sPh_x ||x||^{-\alpha}) \right\} dx \right) \]

where (a) follows from the PGFL of a PPP [2], (b) from $h_\Psi \sim \exp(1)$, and (c) using standard machinery, where the integral is first converted form Cartesian to polar coordinates and the closed form expression follows from the properties of the Gamma function [9, Appendix B].

B. Proof of Lemma 3

The Laplace transform of interference conditioned on the distance of the hole center to the origin, $||y||$, is

$$L_I(||y||)(s) = \mathbb{E} \left[ \exp \left( -s \sum_{x \in \Psi} Ph_x ||x||^{-\alpha} \right) \right]$$

\[= \mathbb{E}_{\Psi} \left[ \prod_{x \in \Psi} \mathbb{E}_{h_\Psi} \left\{ \exp(-sPh_x ||x||^{-\alpha}) \right\} \right] \]

\[= \mathbb{E}_{\Psi} \left[ \prod_{x \in \Psi} \mathbb{E}_{h_\Psi} \left\{ \frac{1}{1 + sP||x||^{-\alpha}} \right\} \right] \]

where (a) follows from the expression for the PGFL of a PPP and (b) is derived by the standard machinery, where the properties of the Gamma function (in the same way as step (c) in the proof of Lemma [1] in Appendix A) [9, Appendix B]. The second term follows from the cosine-law: $r^2 + ||y||^2 - 2r||y||\cos\theta(r) = D^2$ (Fig. 1). By substituting $\lambda(r) = \frac{\lambda}{\pi} \arccos \left( \frac{r^2 + ||y||^2 - D^2}{2||y||r} \right)$, the final expression in equation (3), is derived.

C. Proof of Theorem 7

The lower bound on the Laplace transform of interference is

$$L_I(s) \geq \mathbb{E} \left[ \exp \left( -s \sum_{x \in \Psi} Ph_x ||x||^{-\alpha} \right) \right]$$

\[= \int_D L_I(||y||)(s; \lambda, D)f_{v_1}(v_1)dv_1 \]

\[= \frac{\lambda}{2\pi} \arccos \left( \frac{r^2 + ||y||^2 - D^2}{2||y||r} \right) dr \]

\[\times 2\pi \lambda v_1 \exp(-\pi \lambda_1 (v_1^2 - D^2))dv_1 \]

\[= \lambda \frac{\arccos \left( \frac{r^2 + ||y||^2 - D^2}{2||y||r} \right) dr}{\pi^2} \]

\[\times \exp \left( -\pi \lambda \frac{(sP)^{2/\alpha}}{\text{sinc}(2/\alpha)} \right) \]

\[\times \exp \left( \int_{v_1-D}^{\infty} 2\pi \lambda v_1 \exp(-\pi \lambda_1 (v_1^2 - D^2))dv_1 \right) \]

\[\times \exp \left( \int_{D}^{\infty} \frac{(g(v_1))}{2\pi \lambda v_1} \exp(-\pi \lambda_1 (v_1^2 - D^2))dv_1 \right) \]
where \( b(y_1, D) \) denotes the exclusion zone centered at \( y_1 \) with radius \( D \), and (a) follows by substituting the conditional Laplace transform expression from Lemma 3 and the PDF of \( V_1 \) from (9). Further, \( g(v_1) = \int_{v_1 + D} \arccos \left( \frac{r^2 + v_1^2 - D^2}{2vr} \right) \frac{\lambda}{2\pi} r dr \).

D. Proof of Theorem 2

By definition, the Laplace transform of the PHP is

\[
\mathcal{L}_I(s) \triangleq \mathbb{E} \left[ \exp \left( -s \sum_{x \in \Phi_D \setminus \Xi_D} \| \Phi_x \| - \alpha \right) \right]
\]

\[
\equiv \mathbb{E}_{\Phi_1} \left[ \exp \left( -\lambda_2 \left( \int_{\| \Phi_x \| - D} \frac{dx}{2\pi} - \int_{\Xi_D} \frac{dx}{2\pi} \right) \right) \right]
\]

where \( \Xi_D \) in (a) is defined as in (1), (b) follows by taking expectations over channel gains \( h \sim \exp(1) \) and the PPP \( \Phi_D \) given \( \Xi_D \), where we use the PGFL of a PPP to take expectation over \( \Phi_D \). Note the integral over \( \Xi_D \) is not easy to compute due to the possible overlaps in the holes. Therefore, to derive the bound, we use

\[
\int_{\Xi_D} \frac{dx}{2\pi} \leq \sum_{y \in \Phi_1} \int_{b(y, D)} \frac{dx}{2\pi}
\]

which follows by ignoring the effect of overlaps. Substituting this back in the expression of \( \mathcal{L}_I(s) \); solving the first integral as done in Lemma 3 and using the result of Lemma 3 to handle the integral over \( b(y, D) \), we get

\[
\mathcal{L}_I(s) \leq \exp \left( -\pi \lambda_2 \left( \frac{\mathbb{E}P_{\max}}{\sin(\alpha)} \right) \right) \\
\times \mathbb{E}_{\Phi_1} \left[ \prod_{y \in \Phi_1} \exp \left( 2\alpha \right) \int_{\| \Phi_x \| - D} \frac{dx}{2\pi} \right]
\]

\[
\equiv \exp \left( -\pi \lambda_2 \left( \frac{\mathbb{E}P_{\max}}{\sin(\alpha)} \right) \right) \times \exp \left( -2\pi \lambda_1 \int_{v_1}^{\infty} (1 - \exp(f(v))) vdv \right)
\]

where the second term in (a) follows from the PGFL of a PPP, and then by substituting \( \|y\| = v \) and \( f(v) = \int_{v - D}^{v + D} \arccos \left( \frac{r^2 + v^2 - D^2}{2vr} \right) \frac{\lambda}{2\pi} r dr \). Since by definition of the typical point in this case, there are no points of \( \Phi_1 \) in \( b(0, D) \), the lower bound of integral in the above expression is \( D \).

E. Proof of Proposition 2

Denote the interference powers used for deriving the lower and upper bounds on the Laplace transform of interference in Theorems 1 and 2 by \( I_1 \) and \( I_2 \) by \( I_1 = \sum_{x \in \Phi_D \setminus \Xi_D} P_{\max} \| \Phi_x \| - \alpha \) and \( I_2 = \sum_{x \in \Phi_D \setminus \Xi_D} P_{\max} \| \Phi_x \| - \sum_{y \in \Phi_1} \sum_{x \in \Phi_D \setminus \Xi_D} P_{\max} \| \Phi_x \| - \alpha \), respectively. Here, \( y_1 \) denotes the location of the closest point of \( \Phi_1 \) to the origin. Using this notation, the ratio of the upper and lower bounds is

\[
\frac{L_u(s)}{L_l(s)} = \frac{\mathbb{E}[e^{-sI_u}]}{\mathbb{E}[e^{-sI_l}]}
\]

F. Proof of Theorem 3

In order to derive the lower bound on the Laplace transform of interference, the interference is overestimated by considering only two holes that are closest to the typical node. The setup is illustrated in Fig. 3. The idea is to first derive the Laplace transform conditioned on \( V_1 = \|y_1\|, V_2 = \|y_2\| \), and \( \phi = \arccos \), which is the angle between two holes. Deconditioning on these random variables will yield the final result. The details are as follows:

\[
\mathcal{L}_I(s) \geq \mathbb{E} \left[ \exp \left( -s \sum_{x \in \Phi_D \setminus \Xi_D} P_{\max} \| \Phi_x \| - \alpha \right) \right]
\]

\[
= \mathbb{E}_{\Phi_2} \left[ \prod_{x \in \Phi_1 \cap \Xi_D} \mathbb{E}_{h_x} \left[ \exp \left( -sP_{\max} \| \Phi_x \| - \alpha \right) \right] \right]
\]

\[
= \mathbb{E}_{\Phi_2} \left[ \prod_{x \in \Phi_1 \cap \Xi_D} \mathbb{E}_{h_x} \left[ \exp \left( -sP_{\max} \| \Phi_x \| - \alpha \right) \right] \right]
\]

\[
= \mathbb{E}_{\Phi_2} \left[ \prod_{x \in \Phi_1 \cap \Xi_D} \mathbb{E}_{h_x} \left[ \exp \left( -sP_{\max} \| \Phi_x \| - \alpha \right) \right] \right]
\]

\[
= \mathbb{E}_{\Phi_2} \left[ \prod_{x \in \Phi_1 \cap \Xi_D} \mathbb{E}_{h_x} \left[ \exp \left( -sP_{\max} \| \Phi_x \| - \alpha \right) \right] \right]
\]

\[
= \mathbb{E}_{\Phi_2} \left[ \prod_{x \in \Phi_1 \cap \Xi_D} \mathbb{E}_{h_x} \left[ \exp \left( -sP_{\max} \| \Phi_x \| - \alpha \right) \right] \right]
\]

\[
= \mathbb{E}_{\Phi_2} \left[ \prod_{x \in \Phi_1 \cap \Xi_D} \mathbb{E}_{h_x} \left[ \exp \left( -sP_{\max} \| \Phi_x \| - \alpha \right) \right] \right]
\]

\[
= \mathbb{E}_{\Phi_2} \left[ \prod_{x \in \Phi_1 \cap \Xi_D} \mathbb{E}_{h_x} \left[ \exp \left( -sP_{\max} \| \Phi_x \| - \alpha \right) \right] \right]
\]
two holes. Our first goal is to find the coordinates of the points. The setup is illustrated in Fig. 4. Denoting the locations of the two holes by \( \Xi_C = C_1 \cup C_2 \) with \( C_1 = b(y_1, D) \) and \( C_2 = b(y_2, D) \), and \( \lambda_1(r) \) and \( \lambda_2(r) \) in the last step are \( \lambda_1(r) = \frac{\lambda}{2} \arccos \left( \frac{r^2 + r^2 - 2r^2 - \sqrt{v(2y_2 - 2y_1)} \cos \phi}{2rt} \right) \) and \( \lambda_2(r) = \frac{\lambda}{2} \arccos \left( \frac{r^2 + v^2 - 2r^2}{2rt} \right) \). Step (b) follows from \( h_\kappa \sim \exp(1) \), (c) from the PGFL of PPP, and (d) by deconvolution on the distributions of \( V_1 \) and \( V_2 \). Note that while the joint distribution of \( V_1 \) and \( V_2 \) is given by \( (13) \), \( \phi \) is independent of all other random variables and is uniformly distributed between \( -\pi \) and \( \pi \). In step (d), the terms corresponding to the intersection and closest second holes are derived on the same lines as Lemma 3 (conditioned on \( V_1 \) and \( V_2 \)). The rest of the proof will focus on evaluating the integral \( B(v_1, v_2, \phi) \) that appears in the term corresponding to the intersection of the two holes. Our first goal is to find the coordinates of the points at which the two circles \( C_1 \) and \( C_2 \) intersect. Without loss of generality, we assume that the centers of the two circles are located at \( y_1 = (y_{1x}, 0) \) and \( y_2 = (y_{2x}, y_{2y}) \) while they are separated by distance \( w = \sqrt{v^2 + v^2 - \sqrt{v(2y_2 - 2y_1)} \cos \phi} \). Using the equations of the two circles, \( (x_1 - y_{1x})^2 + y_{1y}^2 = D^2 \), and \( (x_2 - y_{2x})^2 + (x_2 - y_{2y})^2 = D^2 \), we obtain

\[
-2x_1y_{1y} + y_{1y}^2 = -2x_1y_{2y} + y_{2y}^2 - 2y_{2x}y_{2y} + y_{2y}^2,
\]

using which we find \( x_y \) in terms of other variables. Then, substituting \( x_y \) in one of the equations of the circle, we get a quadratic equation for \( x_x \). Solving these equations, the coordinates of the intersection points can be found to be

\[
(\hat{u}_1, \hat{t}_1) = \left( \frac{1}{2} (y_{1x} + y_{2x}) + \frac{1}{2} \sqrt{\frac{4D^2}{w^2} - 1} y_{2y}, \frac{y_{2y}}{2} + \frac{1}{2} \sqrt{\frac{4D^2}{w^2} - 1} - (y_{1y} - y_{2x}) \right)
\]

\[
(\hat{u}_2, \hat{t}_2) = \left( \frac{1}{2} (y_{1x} + y_{2x}) - \frac{1}{2} \sqrt{\frac{4D^2}{w^2} - 1} y_{2y}, \frac{y_{2y}}{2} - \frac{1}{2} \sqrt{\frac{4D^2}{w^2} - 1} (y_{1y} - y_{2x}) \right).
\]

Substituting \( (y_{2x}, y_{2y}) = (v_2 \cos \phi, v_2 \sin \phi) \), \( (y_{1x}, y_{1y}) = (v_1, 0) \) in these expressions, we get

\[
(\hat{u}_1, \hat{t}_1) = \frac{1}{2} \left( v_1 + v_2 \cos \phi + \sqrt{4D^2 - \frac{w^2}{w^2}} - 1 \right) v_2 \sin \phi, \frac{y_{2y}}{2} + \frac{1}{2} \sqrt{\frac{4D^2}{w^2} - 1} (y_{1y} - v_2 \cos \phi)
\]

\[
(\hat{u}_2, \hat{t}_2) = \frac{1}{2} \left( v_1 + v_2 \cos \phi - \sqrt{4D^2 - \frac{w^2}{w^2}} - 1 \right) v_2 \sin \phi, \frac{y_{2y}}{2} - \frac{1}{2} \sqrt{\frac{4D^2}{w^2} - 1} (y_{1y} - v_2 \cos \phi).
\]

Note that the overlap between the circles will happen only when the distance between their centers is smaller than \( 2D \), i.e., \( w \leq 2D \). For a given \( v_1 \) and \( v_2 \geq v_1 \), it can be easily deduced that the overlap occurs only when \( \phi < \hat{\phi} \), where \( \hat{\phi} = \arccos \left( \frac{v_1^2 + v_2^2 - 4D^2}{2v_1v_2} \right) \). The integral in the term corresponding to the intersection of the two circles, \( B(v_1, v_2, \phi) \), can now we derived as

\[
B(v_1, v_2, \phi) = \int_{C_1 \cap C_2} \frac{dx}{1 + \frac{||x||}{\pi P}}
\]

\[
\left\{ \begin{array}{ll}
\int_{u_1}^{u_2} \frac{dx}{\sqrt{D^2 - (x_1 - y_{1x})^2}} \frac{dx_{y_2}}{\sqrt{D^2 - (x_2 - y_{2x})^2}} - \frac{dx_{y_2}}{\sqrt{D^2 - (x_2 - y_{2x})^2}} \\
\int_{u_1}^{u_2} \frac{dx_{y_2}}{\sqrt{D^2 - (x_2 - y_{2x})^2}} \frac{dx_{y_2}}{\sqrt{D^2 - (x_2 - y_{2x})^2}} - \frac{dx_{y_2}}{\sqrt{D^2 - (x_2 - y_{2x})^2}} - \frac{dx_{y_2}}{\sqrt{D^2 - (x_2 - y_{2x})^2}}
\end{array} \right.
\]

(21)

\[
\exp (-\pi \lambda \frac{sP^{2/\alpha}}{\text{sinc}(2/\alpha)})
\]

G. Proof of Theorem 7

We consider \( k \) closest holes to the typical point of a PHP. The setup is illustrated in Fig. 4. Denoting the locations of the holes by \( y_1, \ldots, y_k \), the interference field in this case is modeled by \( \Phi_2 \cap \{ C_1 \cup \{U_{i=2}^k \phi(y_i, D) \} \} \), where \( \Omega \supset \Psi \) and \( d(y_i, D) \) is defined in (15). This approach overestimates the interference power and hence leads to a lower bound on the Laplace transform of interference. Let \( I = \sum_{X \in \Omega} P_{h_x} ||x||^{-\alpha} \), the Laplace transform of interference conditioned on \( V_1 = ||y_1||, V_2 = ||y_2||, \ldots, V_k = ||y_k|| \) is

\[
L_{\alpha} (\|y_1\|, \ldots, \|y_k\|) = E \left[ \exp \left( -s \sum_{X \in \Omega} P_{h_x} ||x||^{-\alpha} \right) \right]
\]

\[
E_{\Omega} \left[ \prod_{x \in \Omega} \frac{1}{1 + sP ||x||^{-\alpha}} \right]
\]

\[
\exp \left( -\lambda_2 \int_{[0, 1] \times [1] \frac{1}{1 + \frac{||x||}{\pi P}} \Delta \right) \int_{[0, 1] \times [1] \frac{1}{1 + \frac{||x||}{\pi P}} \Delta \right)
\]

\[
\exp \left( -\lambda_2 \frac{sP^{2/\alpha}}{\text{sinc}(2/\alpha)} \right)
\]

\[
\exp \left( -\lambda_2 \frac{sP^{2/\alpha}}{\text{sinc}(2/\alpha)} \right)
\]
Area of the shaded region

\[ \pi D^2 - \frac{A_{ol}}{2} \]

Overlapping region

Fig. 17. Illustration of the overlap effect for a typical hole in a PHP. This is used in Appendix H.

\[
\lambda_{eff} \times \left( \pi D^2 - \frac{A_{ol}}{2} \right) = \lambda_2 \pi D^2 \Rightarrow \lambda_{eff} = \frac{\lambda_2}{\pi D^2}.
\]

The above expression shows that \( \frac{1}{(1-\frac{A_{ol}}{2\pi D^2})} \) times more points are removed in the upper bound given by Theorem 2. To compensate for this effect, we first proceed as in Theorem 2 and then rescale the second term (that captures the effect of removing holes) as follows:

\[ L_I(s) \simeq E_{\Phi_1} \left[ \exp \left( -\lambda_2 \left( \int_{R^2} \frac{dx}{1 + \|x\|^2} \right) \sum_{y \in \Phi_1, \, \delta(y, D) \leq y_0 \delta(y, D)} \right) \right] \]

where (a) follows from \( h_{\Psi} \sim \exp(1) \), (b) from PGFL of PPP along with the fact that the points of the baseline PPP \( \Phi_2 \) that are located in the closest hole \( \delta(y, D) \) should be removed, and (c) from the cosine-law by using the same argument applied in the proof of Lemma 3. Finally, deconditioning the resulting expression with respect to the distances of the centers of the \( k \) holes from the typical point, \( V_1, V_2, \ldots, V_k \), with joint distribution given by

\[ f_{V_1, V_2, \ldots, V_k}(v_1, v_2, \ldots, v_k) = (2\pi \lambda)^k v_1 v_2 \ldots v_k \times \exp(-\pi \lambda_1 (v_1^2 - D^2) + \ldots + (-\pi \lambda_1 (v_k^2 - D^2)), \quad D < v_1 < v_2 < \ldots < v_k, \]

completes the proof. Here, the joint PDF \( f_{V_1, V_2, \ldots, V_k}(\cdot) \) is derived by using the same argument as in equation [12].

H. Proof of Proposition 2

Consider the illustration of overlapping circles shown in Fig. 17. While defining a PHP, points of the baseline PPP \( \Phi_2 \) lying inside these circles are removed. In other words, the “union” of such overlapping circles is removed. However, in our upper bound given by Theorem 2, we removed each such circle separately of the others leading to the removal of the overlapping parts multiple times. Therefore, while the average number of points removed per unit area of these circles was \( \lambda_2 \) in a PHP, the average number of points removed per unit area of such circles in Theorem 2 denoted by \( \lambda_{eff} \), is \( \lambda_{eff} > \lambda_2 \). To maintain tractability later in the proof, we confine to the pairwise overlaps, meaning the total overlap area for a circle, denoted by \( A_{ol} \), is the sum of pairwise overlap areas of this circle with the other circles. Now to estimate \( \lambda_{eff} \), consider the shaded region in Fig. 17 with area \( \pi D^2 - \frac{A_{ol}}{2} \), which represents the effective contribution of each hole in a PHP. Due to overlaps, the total average number of points removed under Theorem 2 from this region is \( \lambda_2 \pi D^2 \). Using this fact, \( \lambda_{eff} \) can be estimated as

\[
\lambda_{eff} \times \left( \pi D^2 - \frac{A_{ol}}{2} \right) = \lambda_2 \pi D^2 \Rightarrow \lambda_{eff} = \frac{\lambda_2}{\pi D^2}.
\]
\[ \lambda_1 4\pi D^2 \] circles that overlap with the circle of interest, the average pairwise overlap area is \[ \lambda_1 4\pi D^2 \bar{A} = \lambda_1 \pi D^4. \] Since the maximum overlap is bounded above by \[ \pi D^2, \] we get \[ A_{ol} = \min(\lambda_1 \pi D^4, \pi D^2). \] This completes the proof.

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