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A Tractable Closed-Form Expression of the Coverage Probability in Poisson Cellular Networks

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Abstract—The coverage of cellular networks is usually defined as the probability that the Signal-to-Interference+Noise-Ratio (SINR) is greater than a reliability threshold. Based on this definition, the coverage cannot, in general, be formulated in a tractable closed-form expression. In [1], the authors have introduced a new definition of coverage that is shown to be mathematically tractable for system-level analysis and optimization. In this letter, we conduct a thorough comparison between the two definitions of coverage and show that the coverage introduced in [1] provides one with a tractable and closed-form approximation for the SINR-coverage, which is proved to be an upper-bound in relevant operating regimes. By using the new definition of coverage, the impact of several network parameters, e.g., the density of base stations and mobile terminals, the transmit power, and the transmission bandwidth, can be analytically proved.

Index Terms—Cellular networks, stochastic geometry.

I. INTRODUCTION

During the last few years, stochastic geometry tools, and, in particular, the theory of Poisson Point Processes (PPPs) have been widely used for modeling and analyzing cellular networks [2]. By assuming Base Stations (BSs) and Mobile Terminals (MTs) distributed according to two independent homogeneous PPPs, a cell association criterion based on the highest average received power, a singular path-loss model, Rayleigh fading, as well as BSs and MTs equipped with a single antenna, the authors of [3] have obtained the following tractable expression of the coverage probability (P_{cov}):

$$P_{\text{cov}}^{(\text{SINR})}(\gamma_D) = \Pr\{\text{SINR} \geq \gamma_D\} = \pi \lambda_{\text{BS}} \int_0^{+\infty} e^{-f(x)} dx \quad (1)$$

where symbols and notation in Table I are used, $\text{SINR} = \mathcal{U}/(\sigma_N^2 + \mathcal{I})$ is the Signal-to-Interference+Noise-Ratio, and $f(x) = (\kappa \sigma_N^2 / P_{\text{tx}}) \gamma_D x^{\beta/2} + \pi \lambda_{\text{BS}} (1 + \Upsilon \mathcal{L}(\lambda_{\text{MT}}/\lambda_{\text{BS}})) x$.

The SINR-coverage in (1) cannot, in general, be formulated in closed-form. An exception is the interference-limited regime, where $P_{\text{cov}}^{(\text{SINR})}(\gamma_D) = P_{\text{cov}}^{(\text{SIR})}(\gamma_D) = \Pr\{\text{SIR} \geq \gamma_D\} = (1 + \Upsilon \mathcal{L}(\lambda_{\text{MT}}/\lambda_{\text{BS}}))^{-1}$, and $\text{SIR} = \mathcal{U}/\mathcal{I}$ is the Signal-to-Interference-Ratio. The SIR-coverage is independent of P_{tx} , which may lead to meaningless optimization problems [1].

To overcome these limitations, the authors of [1] have introduced a new definition of coverage, which accounts for the sensitivity of the MTs during both the data detection and

TABLE I

SYMBOLS AND NOTATION ($\alpha = 3.5$, $\delta = 2/\beta$, $\beta > 2$, $\eta = \kappa \sigma_N^2 \gamma_A$).

ALSO: $\bar{\mathcal{U}} = P_{\text{tx}}/\ell(r_0)$, $\mathcal{U} = g_0 P_{\text{tx}}/\ell(r_0)$,

$$\begin{aligned} \mathcal{I} &= \sum_{r_i \in \Psi_{\text{BS}} \setminus r_0} g_i P_{\text{tx}}/\ell(r_i), \Pr\{r_0 \leq r\} = 1 - e^{-\pi \lambda_{\text{BS}} r^2}, \\ \Pr\{g_0 \leq g\} &= \Pr\{g_i \leq g\} = 1 - e^{-g}, \\ \Upsilon &= {}_2F_1(-\delta, 1, 1 - \delta, -\gamma_D) - 1 \geq 0, \mathcal{L}(x) = 1 - (1 + x/\alpha)^{-\alpha}, \\ \mathcal{Q}(x, y, z) &= 1 - \exp\left(-\pi x(y/\eta)^\delta (1 + \Upsilon \mathcal{L}(z))\right). \end{aligned}$$

Symbol/Function	Definition
$\mathbb{E}\{\cdot\}$, $\Pr\{\cdot\}$	Expectation, probability measure
Ψ_{BS} , λ_{BS} , λ_{MT}	PPP and density of BSs, density of MTs
P_{tx} , B_W	Transmit power, transmission bandwidth
N_0 , $\sigma_N^2 = B_W N_0$	Noise power density, noise variance
r_0 , g_0	Distance and fading power of serving BS
r_i , g_i	Distance and fading power of interfering BSs
$\ell(r) = \kappa r^\beta$, κ , β	Path-loss function, constant, slope
γ_D , γ_A	Threshold for decoding, cell association
${}_2F_1(\cdot, \cdot, \cdot, \cdot)$	Gauss hypergeometric function
$\Gamma(\cdot)$, $\text{erfc}(\cdot)$	Gamma function, complementary error function
P_{cov} , PSE	Coverage probability, potential spectral efficiency
$\dot{z}_x(x, \cdot)$, $\ddot{z}_x(x, \cdot)$	1st and 2nd derivative with respect to x

cell association phases. Based on [1], the coverage probability can be formulated in a tractable closed-form expression:

$$\begin{aligned} P_{\text{cov}}(\gamma_D, \gamma_A) &= \Pr\{\text{SIR} \geq \gamma_D, \text{ASNR} \geq \gamma_A\} \\ &= \pi \lambda_{\text{BS}} \int_0^{(P_{\text{tx}}/\eta)^\delta} e^{-\pi \lambda_{\text{BS}} (1 + \Upsilon \mathcal{L}(\lambda_{\text{MT}}/\lambda_{\text{BS}})) x} dx \quad (2) \\ &= \mathcal{Q}(\lambda_{\text{BS}}, P_{\text{tx}}, \lambda_{\text{MT}}/\lambda_{\text{BS}}) (1 + \Upsilon \mathcal{L}(\lambda_{\text{MT}}/\lambda_{\text{BS}}))^{-1} \end{aligned}$$

where $\text{ASNR} = \bar{\mathcal{U}}/\sigma_N^2$ is the Signal-to-Noise-Ratio (SNR) averaged over Rayleigh fading, which accounts for the smallest average SNR needed to ensure a successful cell association.

Problem Statement: Unlike the SINR-coverage that cannot be formulated in closed-form and the SIR-coverage that is independent of key system parameters, the coverage in (2) is given in closed-form, depends on the most relevant network parameters, and is proved in [1] to be simple enough, tractable, and meaningful for system optimization. Motivated by these strengths, this letter has four objectives: i) to identify a suitable choice of γ_A that makes (2) a good approximation of (1), ii) to evaluate the accuracy of the proposed approximation, iii) to study the performance trends of (2) as a function of key parameters to optimize cellular networks, i.e., P_{tx} , B_W , λ_{BS} , λ_{MT} , and iv) to evaluate whether the proposed approximation results in the same performance trends as the SINR-coverage.

II. PROPOSED APPROXIMATION

The rationale of the proposed approximation consists of ensuring that (1) and (2) coincide with (or are asymptotically close to) each other when the corresponding network models operate in *their own* interference-limited and noise-limited regimes. As for (1), these regimes correspond to $P_{\text{cov}}^{(\text{SINR})}(\gamma_D) = \Pr\{\text{SIR} \geq \gamma_D\}$ and $P_{\text{cov}}^{(\text{SINR})}(\gamma_D) =$

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$\Pr\{\text{SNR} \geq \gamma_D\}$, respectively. As for (2), they correspond to $P_{\text{cov}}(\gamma_D, \gamma_A) = \Pr\{\text{SIR} \geq \gamma_D\}$ and $P_{\text{cov}}(\gamma_D, \gamma_A) = \Pr\{\text{ASNR} \geq \gamma_A\}$, respectively. Thus, the following holds.

In the interference-limited regime, i.e., $\sigma_N^2 = 0$, we obtain, by direct inspection of (1) and (2), $P_{\text{cov}}^{(\text{SINR})}(\gamma_D) = P_{\text{cov}}(\gamma_D, \gamma_A) = P_{\text{cov}}^{(\text{SIR})}(\gamma_D)$, which holds true for any γ_A .

In the noise-limited regime, i.e., $\mathcal{I} = 0$, (1) and (2) are:

$$P_{\text{cov}}^{(\text{SINR})}(\gamma_D) = \Pr\{\text{SNR} \geq \gamma_D\} \stackrel{(a)}{=} \mathbb{E}_{g_0} \left\{ \Pr\left\{r_0 \leq (g_0 r_D)^{1/\beta}\right\} \right\} \stackrel{(b)}{\leq} 1 - e^{-\pi \lambda_{\text{BS}} \Gamma(1+\delta) r_D^\delta} \quad (3)$$

$$P_{\text{cov}}(\gamma_D, \gamma_A) = \Pr\{\text{ASNR} \geq \gamma_A\} \stackrel{(a)}{=} \Pr\left\{r_0 \leq r_A^{1/\beta}\right\} = 1 - e^{-\pi \lambda_{\text{BS}} r_A^\delta} \quad (4)$$

where $r_D = P_{\text{tx}}/(\kappa \sigma_N^2 \gamma_D)$, $r_A = P_{\text{tx}}/(\kappa \sigma_N^2 \gamma_A)$, (a) follows from the definitions of SNR and ASNR, and (b) follows by applying the Jensen's inequality and $\mathbb{E}_{g_0}\{g_0^\delta\} = \Gamma(1+\delta)$.

Remark 1: Jensen's inequality is necessary in order to have a tractable closed-form expression of (3) and to avoid using special functions, e.g., the Meijer's G-function. By direct inspection, it is easy to show that the tightness of the upper-bound in (3) increases as either λ_{BS} or P_{tx} decrease. Thus, (3) is expected to be accurate in the noise-limited regime. $\square$

Approach: The SINR-coverage in (1) is approximated as:

$$P_{\text{cov}}^{(\text{SINR})}(\gamma_D) \approx P_{\text{cov}}\left(\gamma_D, \gamma_A = \gamma_D \Gamma(1+\delta)^{-1/\delta}\right) \quad (5)$$

Rationale: Equation (5) follows by equating (3) and (4), which ensures that (2) is a tight upper-bound of (1) if $\mathcal{I} = 0$.

Remark 2: Based on (5), $\gamma_A = \gamma_D \Gamma(1+\delta)^{-1/\delta}$. Arguably, this is a suitable result for system-level optimization, since γ_A depends only on β and γ_D , while it is constant as a function of the key parameters usually used to optimize cellular networks, i.e., P_{tx} , B_W , λ_{BS} , λ_{MT} . Based on (5), e.g., the conclusions drawn in [1] about the optimal P_{tx} and λ_{BS} that maximize the energy efficiency based on (2) apply unaltered to (1). $\square$

A. Expected Accuracy

From the previous section, we evince that the approximation in (5) is exact in the interference-limited regime and is a tight upper-bound in the noise-limited regime. In this section, we study the accuracy of (5) beyond these asymptotic regimes. Due to the lack of a closed-form expression for (1), this is a challenging task for arbitrary system setups. Then, we consider two case studies for which the integral in (1) can be computed.

Lemma 1: If $\beta = 4$, the proposed approximation in (5) is an upper-bound of the SINR-coverage in (1), i.e., $P_{\text{cov}}^{(\text{SINR})}(\gamma_D) \leq P_{\text{cov}}\left(\gamma_D, \gamma_A = \gamma_D \Gamma(1+\delta)^{-1/\delta}\right)$.

Proof: Let $\mathcal{A} = (1/2) \pi \lambda_{\text{BS}} (1 + \Upsilon \mathcal{L}(\lambda_{\text{MT}}/\lambda_{\text{BS}})) r_D^{1/2} \geq 0$, the SINR-coverage in (1) simplifies to $P_{\text{cov}}^{(\text{SINR})}(\gamma_D) = \sqrt{\pi} (1 + \Upsilon \mathcal{L}(\lambda_{\text{MT}}/\lambda_{\text{BS}}))^{-1} \text{Aerfc}(\mathcal{A}) e^{\mathcal{A}^2}$ and (2) reduces to $P_{\text{cov}}\left(\gamma_D, \gamma_A = \gamma_D \Gamma(1+\delta)^{-1/\delta}\right) = (1 - e^{-2\Gamma(3/2)\mathcal{A}}) \times (1 + \Upsilon \mathcal{L}(\lambda_{\text{MT}}/\lambda_{\text{BS}}))^{-1}$. The proof follows, since $\sqrt{\pi} \text{Aerfc}(\mathcal{A}) e^{\mathcal{A}^2} + e^{-2\Gamma(3/2)\mathcal{A}} \leq 1$ for $\mathcal{A} \geq 0$. $\square$

Lemma 2: In the low-noise regime, i.e., $\sigma_N^2 \ll \mathcal{I}$ but $\sigma_N^2 \neq 0$, the approximation in (5) is an upper-bound of (1).

Proof: In the low-noise regime, the approximation $e^{-(1/r_D)x^{1/\delta}} \approx 1 - (1/r_D)x^{1/\delta}$ can be used and (1) becomes:

$$P_{\text{cov}}^{(\text{SINR})}(\gamma_D) \approx P_{\text{cov}}\left(\gamma_D, \gamma_A = \gamma_D \Gamma(1+\delta)^{-1/\delta}\right) + \frac{e^{-\mathcal{B}} - \Gamma(1+1/\delta) \Gamma(1+\delta)^{1/\delta} \mathcal{B}^{-1/\delta}}{1 + \Upsilon \mathcal{L}(\lambda_{\text{MT}}/\lambda_{\text{BS}})} \quad (6)$$

with $\mathcal{B} = \pi \lambda_{\text{BS}} (1 + \Upsilon \mathcal{L}(\lambda_{\text{MT}}/\lambda_{\text{BS}})) r_A^\delta \geq 0$. Consider the function $\mathcal{B}^{1/\delta} e^{-\mathcal{B}}$. It is positive and attains its maximum at $\mathcal{B} = 1/\delta$. Thus, $e^{-\mathcal{B}} \mathcal{B}^{1/\delta} \Gamma(1+\delta)^{-1/\delta} \Gamma(1+1/\delta)^{-1} \leq (1/\delta)^{1/\delta} e^{-1/\delta} \Gamma(1+\delta)^{-1/\delta} \Gamma(1+1/\delta)^{-1} \leq 1$ for $\beta > 2$. So, the second line of (6) is negative and the proof follows. $\square$

B. Interplay of Transmit Power and Density of Base Stations

Given a desired coverage probability, P_{cov}^* , to be fulfilled, three questions are of importance for the design and optimization of cellular networks: i) Is P_{cov}^* achievable? ii) If so, what is the functional relation between P_{tx} and λ_{BS} to achieve P_{cov}^* ? iii) Is this functional relation an injective function?

Equation (5) is instrumental to answer these questions.

Theorem 1: Let $0 \leq P_{\text{cov}}^* \leq 1$ be the coverage to be fulfilled. It can be achieved if $P_{\text{cov}}^* < (1 + \Upsilon \mathcal{L}(\lambda_{\text{MT}}/\lambda_{\text{BS}}))^{-1}$. P_{tx} can be formulated as $P_{\text{tx}} = \kappa \sigma_N^2 \gamma_D \mathcal{T}(\lambda_{\text{BS}})$, where:

$$\mathcal{T}(\xi) = \left(\frac{-\ln(1 - P_{\text{cov}}^* (1 + \Upsilon \mathcal{L}(\lambda_{\text{MT}}/\xi)))}{\pi \xi (1 + \Upsilon \mathcal{L}(\lambda_{\text{MT}}/\xi)) \Gamma(1+\delta)} \right)^{1/\delta} \quad (7)$$

and $\mathcal{T}(\xi = \lambda_{\text{BS}})$ is an injective function in λ_{BS} .

Proof: Equation (7) follows from (5) by setting $P_{\text{cov}}\left(\gamma_D, \gamma_A = \gamma_D \Gamma(1+\delta)^{-1/\delta}\right) = P_{\text{cov}}^*$. The feasibility condition on P_{cov}^* follows from (7), since the argument of the logarithmic function cannot be negative. The injectivity of $\mathcal{T}(\cdot)$ follows by computing its first-order derivative and noting that it is non-positive, i.e., $\dot{\mathcal{T}}_{\lambda_{\text{BS}}}(\lambda_{\text{BS}}) \leq 0$ for $\lambda_{\text{BS}} \geq 0$ and $0 \leq P_{\text{cov}}^* < (1 + \Upsilon \mathcal{L}(\lambda_{\text{MT}}/\lambda_{\text{BS}}))^{-1} \leq 1$, since $\dot{\mathcal{L}}_{\lambda_{\text{BS}}}(\lambda_{\text{MT}}/\lambda_{\text{BS}}) \geq 0$ and $\mathcal{L}(\lambda_{\text{MT}}/\lambda_{\text{BS}}) + \lambda_{\text{BS}} \dot{\mathcal{L}}_{\lambda_{\text{BS}}}(\lambda_{\text{MT}}/\lambda_{\text{BS}}) \geq 0$. $\square$

Remark 3: For brevity, we focus our attention only on the interplay between P_{tx} and λ_{BS} , but (5) can be used to study functional relations among any system parameters. $\square$

To the best of our knowledge, *Theorem 1* constitutes the only available analytical expression that, for cellular networks, provides one with an explicit relation between P_{tx} and λ_{BS} . To get deeper insight into the relation of P_{tx} as a function of λ_{BS} , we study, in detail, two relevant asymptotic regimes.

Corollary 1: Let the BSs be sparsely deployed compared with the MTs, i.e., $\lambda_{\text{BS}}/\lambda_{\text{MT}} \ll 1$. $\mathcal{T}(\xi = \lambda_{\text{BS}})$ is a single variable posynomial function in λ_{BS} of degree $-1/\delta = -\beta/2$:

$$\mathcal{T}(\lambda_{\text{BS}}) \approx \left(\frac{-\ln(1 - P_{\text{cov}}^* (1 + \Upsilon))}{\pi (1 + \Upsilon) \Gamma(1+\delta)} \right)^{1/\delta} \lambda_{\text{BS}}^{-1/\delta} \quad (8)$$

Proof: If $\lambda_{\text{BS}}/\lambda_{\text{MT}} \ll 1$, then $\mathcal{L}(\lambda_{\text{MT}}/\lambda_{\text{BS}}) \approx 1$ holds true. The proof follows directly from (7). $\square$

Corollary 2: Let the BSs be densely deployed compared with the MTs, i.e., $\lambda_{\text{BS}}/\lambda_{\text{MT}} \gg 1$. Let us assume, in addition, $(\lambda_{\text{MT}}/\lambda_{\text{BS}}) \Upsilon \ll 1$. $\mathcal{T}(\xi = \lambda_{\text{BS}})$ is a single variable posynomial function in λ_{BS} of degree $-1/\delta = -\beta/2$:

$$\mathcal{T}(\lambda_{\text{BS}}) \approx (-\ln(1 - P_{\text{cov}}^*)/(\pi \Gamma(1+\delta)))^{1/\delta} \lambda_{\text{BS}}^{-1/\delta} \quad (9)$$

Proof: If $\lambda_{BS}/\lambda_{MT} \gg 1$, then $\mathcal{L}(\lambda_{MT}/\lambda_{BS}) \approx \lambda_{MT}/\lambda_{BS}$ holds true. If, in addition, $(\lambda_{MT}/\lambda_{BS})\Upsilon \ll 1$, then $1 + (\lambda_{MT}/\lambda_{BS})\Upsilon \approx 1$. The proof follows from (7). $\square$

Remark 4: Corollaries 1 and 2 are in agreement with link budget calculations of isolated wireless links: The transmit power that is necessary to fulfill a given coverage probability requirement is a posynomial increasing function of the average cell radius whose order of growth coincides with the path-loss slope (β). This conclusion follows from (8) and (9), by taking into account that the relation between the average cell radius, R_{cell} , and λ_{BS} is, approximately, $\lambda_{BS} = 1/(\pi R_{cell}^2)$. $\square$

III. PERFORMANCE TRENDS

The aim of this section is to study the impact of the most important system parameters, i.e., P_{tx} , B_W , λ_{BS} , λ_{MT} , used to optimize cellular networks. The main objective consists of proving that the approximation in (5) allows one to unveil the impact of P_{tx} , B_W , λ_{BS} , λ_{MT} in rigorous mathematical terms. The relevance of this study for cellular network design lies in the difficulty of doing the same, for some system parameters, by direct inspection of (1). Two performance metrics are analyzed: i) the coverage probability in (5) and ii) the Potential Spectral Efficiency (PSE) defined as follows [1, Eq. (9)]:

$$\widehat{PSE}(\gamma_D) \stackrel{(a)}{\approx} B_W \rho_D \lambda_{BS} \mathcal{L}(\lambda_{MT}/\lambda_{BS}) \widehat{P}_{cov}(\gamma_D) \quad (10)$$

where the approximation in (a) originates from (5), and, for ease of writing, the following notation is used: $\rho_D = \log_2(1 + \gamma_D)$, $\widehat{\gamma}_A = \gamma_D \Gamma(1 + \delta)^{-1/\delta}$, $\widehat{P}_{cov}(\gamma_D) = P_{cov}(\gamma_D, \widehat{\gamma}_A)$, and $\widehat{PSE}(\gamma_D) = PSE(\gamma_D, \widehat{\gamma}_A)$.

In the sequel, for clarity, the network parameter of interest is denoted by ζ . The other parameters are implicitly assumed to be constant. For ease of notation, accordingly, the shorthand notation $\widehat{P}_{cov}(\gamma_D) = \widehat{P}_{cov}(\gamma_D; \zeta) = \mathcal{P}(\zeta)$ and $\widehat{PSE}(\gamma_D) = \widehat{PSE}(\gamma_D; \zeta) = \mathcal{S}(\zeta)$ for P_{cov} and PSE, respectively, is used.

A. Coverage Probability

The impact of P_{tx} , B_W , λ_{BS} , and λ_{MT} on $\mathcal{P}(\cdot)$ is summarized in the following four propositions, respectively.

Proposition 1: Let us consider $\zeta = P_{tx}$. The following holds true: i) $\mathcal{P}(\zeta)$ is monotonically increasing in ζ , ii) $\mathcal{P}(\zeta \rightarrow 0) = 0$, and iii) $\mathcal{P}(\zeta \rightarrow +\infty) = (1 + \Upsilon \mathcal{L}(\lambda_{MT}/\lambda_{BS}))^{-1}$.

Proof: It follows by direct inspection of (5). $\square$

Remark 5: By direct inspection of (1), the same trend as in Proposition 1 holds true for the SINR-coverage. $\square$

Proposition 2: Let us consider $\zeta = B_W$. The following holds true: i) $\mathcal{P}(\zeta)$ is monotonically decreasing in ζ , ii) $\mathcal{P}(\zeta \rightarrow 0) = (1 + \Upsilon \mathcal{L}(\lambda_{MT}/\lambda_{BS}))^{-1}$, iii) $\mathcal{P}(\zeta \rightarrow +\infty) = 0$.

Proof: It follows by direct inspection of (5). $\square$

Remark 6: By direct inspection of (1), the same trend as in Proposition 2 holds true for the SINR-coverage. $\square$

Proposition 3: Let us consider $\zeta = \lambda_{BS}$. The following holds true: i) $\mathcal{P}(\zeta)$ is monotonically increasing in ζ , ii) $\mathcal{P}(\zeta \rightarrow 0) = 0$, and iii) $\mathcal{P}(\zeta \rightarrow +\infty) = 1$.

Proof: From (5), $\dot{\mathcal{P}}_\zeta(\zeta) \geq 0$ holds true if and only if:

$$\dot{\mathcal{Q}}_\zeta(\zeta)(1 + \Upsilon \mathcal{L}(\zeta)) - \mathcal{Q}(\zeta) \Upsilon \dot{\mathcal{L}}_\zeta(\zeta) \geq 0 \quad (11)$$

where $\mathcal{Q}(\zeta) = \mathcal{Q}(\zeta, P_{tx}, \lambda_{MT}/\zeta)$ and $\mathcal{L}(\zeta) = \mathcal{L}(\lambda_{MT}/\zeta)$. (11) is true for $\zeta \geq 0$, since $\mathcal{Q}(\zeta) \geq 0$, $\dot{\mathcal{Q}}_\zeta(\zeta) \geq 0$, $\mathcal{L}(\zeta) \geq 0$, and $\dot{\mathcal{L}}_\zeta(\zeta) \leq 0$ for $\zeta \geq 0$. Then, i) is proved. ii) and iii) follow from $\mathcal{L}(\zeta \rightarrow 0) = 1$ and $\mathcal{L}(\zeta \rightarrow +\infty) = 0$, respectively. $\square$

Remark 7: Due to the lack of a closed-form expression for the SINR-coverage, it is not straightforward to prove, in mathematical terms, the impact of λ_{BS} directly from (1). Since $\dot{\mathcal{L}}(\lambda_{MT}/\zeta) \leq 0$, i.e., $\mathcal{L}(\zeta)$ is decreasing in ζ , the same trend as in Proposition 3 holds true for the SIR-coverage. $\square$

Proposition 4: Let us consider $\zeta = \lambda_{MT}$ and denote $\widehat{r}_A = P_{tx}/(\kappa \sigma_N^2 \widehat{\gamma}_A)$. The following holds true: i) $\mathcal{P}(\zeta)$ is monotonically decreasing in ζ , ii) $\mathcal{P}(\zeta \rightarrow 0) = 1 - e^{-\pi \lambda_{BS} \widehat{r}_A^{1/\delta}}$, and iii) $\mathcal{P}(\zeta \rightarrow +\infty) = (1 - e^{-\pi \lambda_{BS} \widehat{r}_A^{1/\delta}(1+\Upsilon)})/(1 + \Upsilon)$.

Proof: From (5), $\dot{\mathcal{P}}_\zeta(\zeta) \leq 0$ holds true if and only if:

$$\Upsilon \dot{\mathcal{L}}_\zeta(\zeta) \left[(1 + \mathcal{C}(1 + \Upsilon \mathcal{L}(\zeta))) e^{-\mathcal{C}(1 + \Upsilon \mathcal{L}(\zeta))} - 1 \right] \leq 0 \quad (12)$$

where $\mathcal{C} = \pi \lambda_{BS} \widehat{r}_A^{1/\delta}$, $\mathcal{Q}(\zeta) = \mathcal{Q}(\lambda_{BS}, P_{tx}, \zeta/\lambda_{BS})$, and $\mathcal{L}(\zeta) = \mathcal{L}(\zeta/\lambda_{BS})$. (12) is true for $\zeta \geq 0$, since $\dot{\mathcal{L}}_\zeta(\zeta) \geq 0$ for $\zeta \geq 0$ and the term inside the brackets is negative for $\mathcal{C}(1 + \Upsilon \mathcal{L}(\zeta)) \geq 0$. Then, i) is proved. ii) and iii) follow from $\mathcal{L}(\zeta \rightarrow 0) = 0$ and $\mathcal{L}(\zeta \rightarrow +\infty) = 1$, respectively. $\square$

Remark 8: Since $\mathcal{L}(\zeta/\lambda_{BS}) \geq 0$ and $\dot{\mathcal{L}}_\zeta(\zeta/\lambda_{BS}) \geq 0$, i.e., $\mathcal{L}(\zeta)$ is increasing in $\zeta = \lambda_{MT}$, we evince, from (1), that the SINR-coverage behaves like (5) as a function of λ_{MT} . $\square$

From Propositions 1-4, a few comments can be made:

- Remarks 5, 6, 8 are readily obtained from (1). They are, however, instrumental to explicitly substantiating that the proposed approximation in (5) yields the same performance trends as (1). The impact of λ_{BS} (see Remark 7) is evaluated in Section IV with Monte Carlo simulations.
- The performance trend as a function of λ_{MT} (see Proposition 4) can be explained as follows: By increasing λ_{MT} , the number of active BSs increases, which, in turn, enhances the aggregate other-cell interference, $\mathcal{I}$. As a result, the coverage probability decreases.
- By comparing Proposition 1 against Proposition 3, we evince that network densification, i.e., increasing λ_{BS} , has the potential of providing the typical MT with zero outage (the coverage probability tends asymptotically to one as $\lambda_{BS} \rightarrow +\infty$). This is not possible, on the other hand, by increasing P_{tx} : As $P_{tx} \rightarrow +\infty$, the coverage probability reaches an upper-limit that is less than one.

B. Potential Spectral Efficiency

In this section, we turn our attention to the PSE in (10). Unlike the analysis of the coverage in Section III-A, the impact of B_W , λ_{BS} , λ_{MT} cannot be readily proved for the SINR-PSE, i.e., the PSE obtained by replacing $\widehat{P}_{cov}(\cdot)$ in (10) with the SINR-coverage in (1). The reason of the analytical complexity is the presence of the multiplicative factor $B_W \lambda_{BS} \mathcal{L}(\lambda_{MT}/\lambda_{BS})$, which depends on B_W , λ_{BS} , λ_{MT} . Therefore, the performance trends reported in this section are non-trivial, and, to the best of our knowledge, they have not been mathematically and rigorously proved elsewhere.

The impact of B_W , λ_{BS} , and λ_{MT} on $\mathcal{S}(\cdot)$ is summarized in the following three propositions, respectively. The impact of

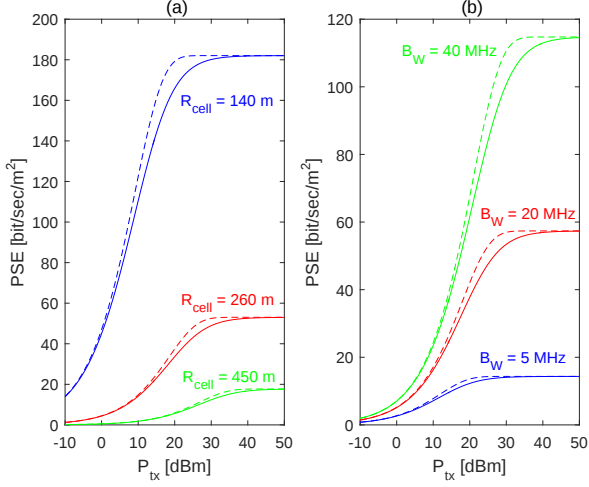


Fig. 1. PSE as a function of P_{tx} and R_{cell} (a) and B_W (b).

P_{tx} is, on the other hand, not discussed in detail. It follows, in fact, by direct inspection of the PSE in (10) and by noting that PSE and coverage have the same behavior. Thus, the trends discussed in *Proposition 1* and *Remark 5* apply to PSE as well.

Proposition 5: Let us consider $\zeta = B_W$ and denote $\bar{r}_A = P_{tx}/(\kappa N_0 \gamma_A)$. The following holds true: i) $\mathcal{S}(\zeta)$ is monotonically increasing in ζ , ii) $\mathcal{S}(\zeta \rightarrow 0) = 0$, and iii) $\mathcal{S}(\zeta \rightarrow +\infty) = +\infty$.

Proof: The PSE is $\mathcal{S}(\zeta) = \mathcal{D}\zeta(1 - e^{-\mathcal{E}\zeta^{-\delta}})$, where $\mathcal{D} = \rho_D \lambda_{BS} \mathcal{L}(\lambda_{MT}/\lambda_{BS})/(1 + \Upsilon \mathcal{L}(\lambda_{MT}/\lambda_{BS}))$ and $\mathcal{E} = \pi \lambda_{BS} (1 + \Upsilon \mathcal{L}(\lambda_{MT}/\lambda_{BS})) \bar{r}_A^\delta$. i) follows because $\dot{\mathcal{S}}_\zeta(\zeta) \geq 0$, since $\dot{\mathcal{S}}_\zeta(\zeta \rightarrow 0) = 1$, $\dot{\mathcal{S}}_\zeta(\zeta \rightarrow +\infty) = 0$, and $\dot{\mathcal{S}}_\zeta(\zeta) = -\delta \mathcal{E} \mathcal{D} e^{-\mathcal{E}\zeta^{-\delta}} (\delta \mathcal{E} \zeta^{-(1+\delta)} + (1 - \delta) \zeta^{-(1+\delta)}) \leq 0$ for $\beta > 2$. ii) directly follows from (10) and iii) is obtained by computing the limit $\mathcal{S}(\zeta \rightarrow +\infty)$ with the aid of De L'Hopital's rule. $\square$

Remark 9: Comparing *Propositions 2* and *5*, we evince that P_{cov} and PSE have opposite trends as a function of B_W . $\square$

Proposition 6: Let us consider $\zeta = \lambda_{BS}$. The following holds true: i) $\mathcal{S}(\zeta)$ is monotonically increasing in ζ , ii) $\mathcal{S}(\zeta \rightarrow 0) = 0$, and iii) $\mathcal{S}(\zeta \rightarrow +\infty) = B_W \rho_D \lambda_{MT}$.

Proof: i) follows because $\mathcal{P}(\zeta)$ (see *Proposition 3*) and $\zeta \mathcal{L}(\lambda_{MT}/\zeta)$ increase in ζ . ii) and iii) follow from (10) by using *Proposition 3* and taking into account that $\zeta \mathcal{L}(\lambda_{MT}/\zeta) \rightarrow 0$ if $\zeta \rightarrow 0$ and $\zeta \mathcal{L}(\lambda_{MT}/\zeta) \rightarrow \lambda_{MT}$ if $\zeta \rightarrow +\infty$. $\square$

Proposition 7: Let us consider $\zeta = \lambda_{MT}$ and denote $\hat{r}_A = P_{tx}/(\kappa \sigma_N^2 \gamma_A)$. The following holds true: i) $\mathcal{S}(\zeta)$ is monotonically increasing in ζ , ii) $\mathcal{S}(\zeta \rightarrow 0) = 0$, and iii) $\mathcal{S}(\zeta \rightarrow +\infty) = B_W \rho_D \lambda_{BS} (1 - e^{-\pi \lambda_{BS} \hat{r}_A^\delta (1 + \Upsilon)}) / (1 + \Upsilon)$.

Proof: From (10), $\dot{\mathcal{S}}_\zeta(\zeta) \geq 0$ holds true if and only if:

$$\dot{\mathcal{L}}_\zeta(\zeta) \mathcal{Q}(\zeta) + \mathcal{L}(\zeta) \dot{\mathcal{Q}}_\zeta(\zeta) + \Upsilon \mathcal{L}^2(\zeta) \dot{\mathcal{Q}}_\zeta(\zeta) \geq 0 \quad (13)$$

where $\mathcal{Q}(\zeta) = \mathcal{Q}(\lambda_{BS}, P_{tx}, \zeta/\lambda_{BS})$ and $\mathcal{L}(\zeta) = \mathcal{L}(\zeta/\lambda_{BS})$. (13) is true for $\zeta \geq 0$, since $\mathcal{Q}(\zeta) \geq 0$, $\dot{\mathcal{Q}}_\zeta(\zeta) \geq 0$, $\mathcal{L}(\zeta) \geq 0$, and $\dot{\mathcal{L}}_\zeta(\zeta) \geq 0$ for $\zeta \geq 0$. Then, i) is proved. ii) and iii) follow from $\mathcal{L}(\zeta \rightarrow 0) = 0$ and $\mathcal{L}(\zeta \rightarrow +\infty) = 1$, respectively. $\square$

Remark 10: Comparing *Propositions 4* and *7*, we note that P_{cov} and PSE have opposite trends as a function of λ_{MT} . $\square$

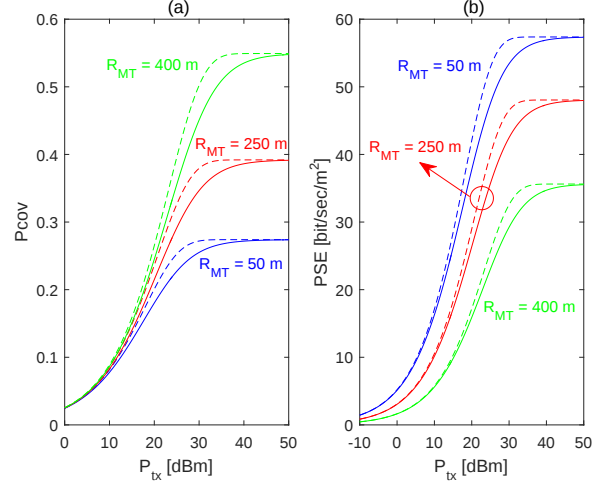


Fig. 2. P_{cov} (a) and PSE (b) as a function of P_{tx} and R_{MT} .

IV. NUMERICAL RESULTS

The aim of this section is to analyze the accuracy of the approximation in (5) and to substantiate the performance trends proved in Section III as a function of the network parameters P_{tx} , B_W , λ_{BS} , and λ_{MT} . Unless otherwise stated, the simulation setup is: $\beta = 3.5$, $f_c = 2.1$ GHz, $\kappa = (4\pi f_c/3 \cdot 10^8)^2$, $N_0 = -174$ dBm/Hz, $B_W = 20$ MHz, $\lambda_{BS} = 1/(\pi R_{cell}^2)$ BSs/m² with $R_{cell} = 250$ m, $\lambda_{MT} = 1/(\pi R_{MT}^2) = 121$ MTs/km² with $R_{MT} = 51.29$ m, and $\gamma_D = 5$ dB.

The numerical results are illustrated in Figs. 1 and 2. The solid lines report the SINR-coverage and SINR-PSE obtained from (1) and (10). The dashed lines show the proposed approximation based on (5). The numerical illustrations provide evidence that the proposed approximation is accurate enough for practical applications, and, more importantly, yields the same trends as the widely used SINR-coverage. Therefore, (5) and (10) constitute accurate and tractable analytical frameworks for the analysis and optimization of cellular networks as a function of several network parameters of interest.

V. CONCLUSION

In this letter, we have shown that the coverage probability recently introduced in [1] constitutes a closed-form upper-bound of the conventional SINR-coverage. The new definition of coverage probability provides one with a tractable analytical framework, from which the impact of the most important system parameters for the design of cellular networks can be rigorously proved. Possible generalizations of this work include the analysis of bounded path-loss models, tri-dimensional network deployments, multi-tier cellular networks.

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