Exact SNR and SIR analysis in Poisson wireless networks

G.P. Efthymoglou[™], P.S. Bithas and A.G. Kanatas

The probability density function and cumulative distribution function of the received signal-to-noise ratio (SNR) and the received signalto-interference ratio (SIR), for interference-limited systems is derived, at the *n*th nearest neighbour node in a Poisson point process wireless random network. The analytical expressions are given in terms of the Meijer G-function and reveal the impact of node spatial density, transmit power, interference power, and path-loss exponent on the connectivity probability of a broadcast wireless transmission. The analytical results are validated with computer simulation.

Introduction: Many works studied distance distributions in random infinite and finite networks [1–3]. A model that is usually used to characterise the spatial distribution of these nodes is the Poisson point process (PPP). For example, the work in [4] studied the connectivity probability for vehicular-to-vehicular (V2V) and vehicular-to-infrastructure communication scenarios under the assumption that vehicles are distributed on the road following a Poisson distribution. For a random network with node distribution based on the PPP model and the transmitter located at the origin, the distance between this node and its *n*th nearest neighbour is a random variable that follows the generalised Gamma distribution [1, 5]. Therefore, for constant transmit power and a power-law path-loss model, the received average signal-to-noise ratio (SNR) or the average received signal-to-interference ratio (SIR), for the case of interference-limited systems, are random variables.

In this Letter, we consider the combined effect of path-loss and Nakagami-*m* fading on the outage probability of the *n*th nearest neighbour node. Closed-form expressions for the probability density function (PDF) and cumulative distribution function (CDF) of the received SNR or received SIR at the *n*th nearest neighbour are derived. The analytical results can be utilised to determine the probability of correct detection at the *n*th nearest neighbour node in a PPP wireless network with different node densities.

Signal to noise analysis: The Poisson process is suitable for modelling uniformly random networks. The Euclidean distance between a point at the origin and its *n*th nearest neighbour, R_n , is distributed according to the generalised Gamma distribution [1]:

$$f_{R_n}(r) = \frac{m(\lambda c_m r^m)^n}{r\Gamma(n)} \exp(-\lambda c_m r^m)$$
(1)

where λ is the node spatial density, $c_m r^m$ is the volume of the *m*-dimensional ball of radius *r*, c_m is given by

$$c_m = \begin{cases} \frac{\pi^{m/2}}{(m/2)!}, & \text{even } m \\ \frac{\pi^{(m-1)/2} 2^m ((m-1)/2)}{m!} & \text{odd } m \end{cases}$$
(2)

and $\Gamma(n)$ is the gamma function evaluated at *n*. In addition, if we want to consider only neighbouring nodes that lie within a sector with opening angle ϕ , this simply corresponds to a change of the volume from an *m*-ball to an *m*-sector (with opening angle ϕ) whose volume is $c_{\phi,m}r^m$. Therefore, the PDF of the distance to the *n*th nearest neighbour in a sector ϕ is given by replacing c_m by $c_{\phi,m}$ in (2). For m = 1, 2, 3, we have $c_{\phi,1} = 1$, $c_{\phi,2} = \phi$, and $c_{\phi,3} = (2\pi/3)(1 - \cos(\phi))$, respectively. Moreover, it is assumed that the transmitted signal undergoes small-scale Nakagami-*m* fading. The corresponding instantaneous received SNR, *X*, follows the gamma distribution with PDF

$$f_X(x) = \left(\frac{m_{\rm s}}{\Omega_{\rm s}}\right)^{m_{\rm s}} \frac{x^{m_{\rm s}-1}}{\Gamma(m_{\rm s})} \exp\left(-\frac{m_{\rm s}}{\Omega_{\rm s}}x\right) \tag{3}$$

where $m_{\rm s}$ and $\Omega_{\rm s}$ are the distribution's shaping and scaling parameters. When the path-loss follows the decaying power law, the average SNR $\Omega_{\rm s}$ at distance r is given by

$$\Omega_{\rm s} = P_{\rm t} K r^{-\alpha} / N = \tilde{P}_{\rm t} r^{-\alpha} \tag{4}$$

where P_t is the transmit power, K is a constant that depends on the antenna characteristics and free-space path-loss up to distance $r_0 = 1 \text{ m}$, r is a random variable which follows the distribution in (1),

 α is the path-loss exponent with values in the range [2, 6], N is the receiver noise power, and \tilde{P}_t is the transmit SNR. The PDF of the received SNR at the *n*th nearest neighbour is then given by

$$f_{X_n}(x) = \int_0^\infty f_X(x|r) f_{R_n}(r) dr$$

= $\frac{x^{m_s - 1}m}{\Gamma(m_s)\Gamma(n)} \left(\frac{m_s}{\tilde{P}_t}\right)^{m_s} (\lambda c_{\phi,m})^n$
 $\times \int_0^\infty r^{\alpha m_s + mn - 1} \exp\left(-\frac{m_s x}{\tilde{P}_t}r^{\alpha}\right) \exp\left(-\lambda c_{\phi,m}r^m\right) dr$ (5)

The previous integral can be solved using the following result, which is derived from [6, Equations (11) and (21)] as

$$\int_{0}^{\infty} x^{\nu} \exp(-B_{1}x^{b_{1}}) \exp(-B_{2}x^{b_{2}}) dx = \frac{1}{\sqrt{b_{1}b_{2}}} \left(\frac{b_{2}}{B_{1}}\right)^{((\nu+1)/b_{1})} \times \frac{1}{(2\pi)^{((b_{1}+b_{2}-2)/2)}} G_{b_{2}b_{1}}^{b_{1}b_{2}} \left(\frac{B_{2}^{b_{1}}b_{2}^{b_{2}}}{B_{1}^{b_{2}}b_{1}^{b_{1}}}\right) \Delta\left(b_{2}, 1 - \frac{\nu+1}{b_{1}}\right) \right)$$
(6)

where b_1 and b_2 are integers, $G(\cdot)$ is the Meijer's G-function and $\Delta(m, n) = n/m, \ldots, ((n + m - 1)/m)$. It is noted that Meijer G-functions are built-in functions in many mathematical software packages, e.g. Mathematica, Maple, and thus can be directly evaluated. For path-loss exponent expressed as $\alpha = \ell/k$, where ℓ and k are integers, based on the solution given in (6) with $b_1 = \ell$ and $b_2 = km$, the PDF of the SNR at the *n*th nearest neighbour node is given in closed form as

$$f_{X_{n}}(x) = \frac{(m_{s}/\tilde{P}_{t})^{m_{s}} (\lambda c_{\phi,m})^{n} x^{m_{s}-1}}{\Gamma(m_{s}) \Gamma(n)(2\pi)^{((\ell+km)/2)-1}} \sqrt{\frac{m}{\alpha}} \left(\frac{m_{s}x}{km\tilde{P}_{t}}\right)^{-m_{s}-(mn/\alpha)} \times G_{km,\ell}^{\ell,km} \left(\left(\frac{\lambda c_{\phi,m}}{\ell}\right)^{\ell} \left(\frac{km\tilde{P}_{t}}{m_{s}x}\right)^{km} \left| \frac{\Delta(km, 1-m_{s}-\frac{mn}{\alpha})}{\Delta(\ell, 0)} \right) \right)$$
(7)

Moreover, using [6, eq. (26)], the corresponding CDF of the received SNR is given in closed form as

$$F_{X_n}(x) = \int_0^x f_{X_n}(x) dx$$

= $1 - \frac{(m_s/\tilde{P}_1)^{m_s} (\lambda c_{\phi,m})^n}{\Gamma(m_s)\Gamma(n)k(2\pi)^{((\ell+km)/2)-1}\sqrt{\alpha m}} \left(\frac{m_s}{km\tilde{P}_1}\right)^{-m_s-(mn/\alpha)}$
 $\times \left(\frac{1}{x}\right)^{mn/\alpha} G_{k,m+1,\ell+1}^{\ell,km+1} \left(\left(\frac{\lambda c_{\phi,m}}{\ell}\right)^\ell \left(\frac{km\tilde{P}_1}{m_s x}\right)^{km}\right)^{km}$ (8)
 $\left| \begin{array}{c} \Delta(km, 1-m_s - \frac{mn}{\alpha}), 1 - \frac{n}{\ell} \\ \Delta(\ell, 0), -\frac{n}{\ell} \end{array} \right)$

Signal to interference analysis: Assuming that the received signal is subject to interference, the received SIR for an interference-limited system is given by

$$\gamma_{\rm out} = \frac{X}{Y} \tag{9}$$

where Y denotes the instantaneous received interference-to-noise ratio (INR). Assuming a single dominant Nakagami-*m* interferer, the PDF of γ_{out} is given by

$$f_{\gamma_{\text{out}}}(x) = \left(\frac{m_{\text{s}}}{\Omega_{\text{s}}}\right)^{m_{\text{s}}} \left(\frac{m_{\text{I}}}{\Omega_{\text{I}}}\right)^{m_{\text{I}}} \frac{\Gamma(m_{\text{s}}+m_{\text{I}})}{\Gamma(m_{\text{s}})\Gamma(m_{\text{I}})} \frac{x^{m_{\text{s}}-1}}{\left(\frac{m_{\text{I}}}{\Omega_{\text{I}}} + \frac{m_{\text{s}}}{\Omega_{\text{s}}}x\right)^{m_{\text{s}}+m_{\text{I}}}}$$
(10)

where Ω_I is the average INR and m_I is the Nakagami fading parameter of the interference signal. Based on (10), using [6, eqs. (10), (11) and (21)] the PDF of the SIR at the *n*th nearest neighbour is given by

$$\begin{split} f_{\gamma_{\text{out,n}}}(x) &= \frac{k^{m_{\text{s}}+m_{\text{I}}+(\alpha m_{\text{s}}/m)+n-1/2}}{\Gamma(m_{\text{s}})\Gamma(m_{\text{I}})\Gamma(n)} \left(\frac{m_{\text{s}}\Omega_{\text{I}}}{m_{\text{I}}\tilde{P}_{\text{I}}}\right)^{m_{\text{s}}} \\ &\times \frac{m^{m_{\text{s}}+m_{\text{I}}}\alpha^{(\alpha m_{\text{s}}/m)+n-1/2}}{(2\pi)^{(\ell/2)+km-3/2}(\lambda c_{\phi,m})^{\alpha m_{\text{s}}/m}} x^{m_{\text{s}}-1} \\ &\times G_{\ell+km,km}^{km} \left(\left(\frac{\ell}{\lambda c_{\phi,m}}\right)^{\ell} \left(\frac{m_{\text{s}}\Omega_{\text{I}}x}{m_{\text{I}}\tilde{P}_{\text{I}}}\right)^{km} \\ & \left|\frac{\Delta(km, 1-m_{\text{s}}-m_{\text{I}}), \Delta(\ell, 1-n-\frac{\alpha m_{\text{s}}}{m})\right)\right) \quad (11) \end{split}$$

Moreover, using [6, eq. (26)], the corresponding CDF of the received SIR is given in closed form as

$$F_{\gamma_{\text{out,n}}}(x) = \frac{k^{m_s+m_1+(\alpha m_s/m)+n-3/2}}{\Gamma(m_s)\Gamma(m_1)\Gamma(n)} \left(\frac{m_s\Omega_1}{m_1\tilde{P}_t}\right)^{m_s} \\ \times \frac{m^{m_s+m_1-1}\alpha^{(\alpha m_s/m)+n-1/2}}{(2\pi)^{\ell/2+km-3/2}(\lambda c_{\phi,m})^{\alpha m_s/m}} x^{m_s} \\ \times G_{\ell+km+1,km+1}^{km,\ell+km+1} \left(\left(\frac{\ell}{\lambda c_{\phi,m}}\right)^{\ell} \left(\frac{m_s\Omega_1 x}{m_1\tilde{P}_t}\right)^{km} \\ \left| \Delta(km, 1-m_s-m_1), \Delta(\ell, 1-n-\frac{\alpha m_s}{m}), 1-\frac{m_s}{km} \right) \right|$$

$$(12)$$



Fig. 1 Probability of connectivity based on SNR versus neighbour index n

Numerical results: The probability of coverage or probability of connectivity for the *n*th nearest neighbour can be defined as the complementary CDF of the received SNR as

$$\Pr\{X_n > \gamma_{\text{th}}\} = 1 - F_{X_n}(\gamma_{\text{th}}) \tag{13}$$

which is the probability that the SNR at the *n*th nearest neighbour is greater than the target SNR γ_{th} .

For all the numerical results, we consider a 2D case and a sector of 90° angle, by setting m=2, $\phi = \pi/4$ and $c_{\phi,2} = \phi$. This setting could model a V2V communications scenario where a car broadcasts information to the vehicles behind it. To determine the coverage area of the transmitter, the average distance to the *n*th nearest neighbour, is given by [1]

$$E[d_n] = \frac{\Gamma(n+12)}{\Gamma(n)\sqrt{\lambda\varphi}}$$
(14)

This distance determines how far a node can transmit given a minimum required SNR at the receiver, i.e. the length of the longest possible hop for a given transmit power. For example, for PPP density $\lambda = 0.001$, the average distance to neighbours $n = \{1,10,20\}$ obtained from (14) with m = 2 and $\phi = \pi/4$ are $E[d_n] = \{32, 111, 158\}$ m, whereas for $\lambda = 0.01$ we obtain $E[d_n] = \{10, 35, 50\}$ m.

Fig. 1 plots the probability of connectivity versus the neighbour index *n* for PPP density $\lambda = \{1, 5, 10\} \cdot 10^{-3}$, assuming $\tilde{P}_t = 70 \text{ dB}$, $m_s = 2$, path-loss exponents $\alpha = 3$ ($\ell = 3$, k = 1) and $\alpha = 3.5$ ($\ell = 7$, k = 2), and $\gamma_{\text{th}} = 5 \text{ dB}$. The figure shows the impact of path-loss exponent α and PPP density λ on the probability of connectivity to the *n*th nearest neighbour. We note that assuming N = -105 dBm and K = -35 dB, $\tilde{P}_t = 70 \text{ dB}$ corresponds to $P_t = 70 + 35 - 105 = 0 \text{ dBm}$.

In Fig. 2 we plot the probability of connectivity versus neighbour index *n* for an interference-limited system, that is, for a system where connectivity is based on the SIR association criterion. We assume the same system parameters with the previous plot but also consider $m_{\rm I} = 1$ and INR $\Omega_{\rm I} = 10 \, \rm dB$ for the interference signal. The plot shows the impact of interference on the probability that the *n*th nearest neighbour will successfully detect the transmission from the source node located at the origin.



Fig. 2 Probability of connectivity based on SIR versus neighbour index n

Conclusion: In this Letter, we derived closed-form expressions for the PDF and CDF of the received SNR and received SIR at the *n*th nearest neighbour in a PPP random wireless network. The analytical expressions were used to investigate the impact of node density, transmit power, path-loss exponent, and association threshold on the probability that the *n*th nearest neighbour to a source node can successfully decode a transmitted packet.

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G.P. Efthymoglou, P.S. Bithas and A.G. Kanatas (Dept. of Digital Systems, University of Piraeus, Piraeus, Greece)

□ E-mail: gefthymo@unipi.gr

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