Enabling Wireless Power Transfer and Multiple Antennas Selection to IoT Network Relying on NOMA

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Abstract—Wireless Power Transfer (WPT) is a significant technique for Internet of Things (IoT) networks. Recently, more interest has been focused on multiple access technique without orthogonal signals for wireless communication. Non-orthogonal Multiple Access (NOMA) scheme is proposed to allow users the access point in IoT network. In this paper, we propose the power beacon which is able to feed energy to power-constraint relay node to further support transmission from the source to destinations in IoT networks. In this article, a NOMA system is benefited by with WPT and antenna selection technique. The system improvement can be achieved through the exact closed-form expressions of outage probability (OP). The performance gap among two users is evaluated using model of the Rayleigh fading channels. Furthermore, we compare NOMA with traditional scheme to highlight advantage of such IoT system.

Index Terms—Non-orthogonal multiple access; IoT; Outage probability.

I. INTRODUCTION

It has been predicted that Internet will be connected by 50 billion of devices, including small sensors or IoT devices, by 2020 [1]. The fast growth of IoT applications results in requirements of heterogeneous IoT sensor accessing to networks in the upcoming fifth generation (5G) networks. These applications are related to capability of devices in device-to-device (D2D) networks, machine-to-machine (M2M) networks, and other services and applications associated with IoT systems [2]–[5]. It is highly capability to implement the automation in IoT system, thanks to the technological improvement in IoT, and it does not require human control [6]. It is necessary to reliable exchange of information and data for these IoT sensors and devices related to the core of distributed automation [7]. An enormous amount of power is consumed by these small sensors and IoT devices to serve their data transmission and communication. However, popular networks contain these battery-powered small sensors or battery-powered devices. It is noted that there is higher demand to remain operation of huge number of sensors, especially providing self-sustainable green communications for IoT networks since devices are limited energy [8]. These power-constraint sensors need their lifetime to be extended, and hence it requires possible solution of energy efficient data transmission.

Unfortunately, it is difficult to replace large numbers of sensors in IoT systems. As a result, the power-constraint devices are operated in practice, and such situation limits the performance improvement. The energy harvesting technique is proposed to tackle this problem by harvesting the energy from the surrounding environments. To harvest the energy from the radio-frequency signals, Radio-frequency (RF) energy harvesting is applied [9]. The relaying networks have been widely studied by introducing flexible, sustainable, and stable energy supply to devices in such networks [10]–[14]. In [11], renewable energy is considered as a solution to employ dense small cell base stations (SBSs) to adapt to the increasing demand of communication services. The authors in [12] studied for unmanned aerial vehicle (UAV)-assisted networks in term of the resource allocation problem. In this system, multiple energy harvesting-powered D2D pairs are powered by a UAV which play UAV as an energy source providing radio-frequency energy. In [13], system throughput can be enhanced by utilizing optimal channel selection method and the harvested RF energy as well. The

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cognitive radio sensor networks benefit from the RF energy harvesting in [14].

Considering as a prominent wireless access technique for the 5G wireless communication, Non-orthogonal Multiple Access (NOMA) is introduced and analysed in [15]. To provide higher spectrum efficiency, NOMA employs non-orthogonal transmission at the transmitter. The transmitter divides power domain to serve multiple users with superposed signal transmitted. Different from Orthogonal Multiple Access (OMA), NOMA can serve multiple users over the same resource block, thus it can effectively improve sum rate in other emerging networks [16], [17]. To decode the users’ information at the receiver, the Successive Interference Cancellation (SIC) is required. Specifically, signal is decoded firstly for the user with the best channel condition while assuming other users’ signal as interference. However, results in [16] did not consider multiple antennas and multiple beacon since these techniques benefit to performance improvement. The IoT benefits from power beacon, and some metrics are studied to exhibit system performance [18]. Motivated by these papers [16–18], we formulate the problem of selection of beacon and antenna selection to evaluate outage performance of two NOMA users.

The remaining parts of this paper are arranged as follows. Section II presents the system model based on NOMA to implement IoT system. In Section III, we consider the outage performance of such NOMA applied together with Wireless Power Transfer (WPT). In addition, the traditional technique of OMA is presented in Section IV. The numerical simulations are conducted in Section V, and we provide conclusion remarks in Section VI.

II. SYSTEM MODEL

In this model, Fig. 1 shows the IoT system containing the access point (AP), the relay R, two users D1, D2, and power beacon (B). Wireless channels denoted as in Fig. 1 in such IoT system relying NOMA are subjected to Rayleigh flat fading plus additive White Gaussian noise. The complex channel coefficients for the links AP → K, B → R, R → D1, K → D2 are represented by $|h_{KR}|^2 \sim CN(0, \lambda_{KR})$, $|h_{BR}|^2 \sim CN(0, \lambda_{BR})$, and $|h_{BD1}|^2 \sim CN(0, \lambda_{BD1})$, respectively. There are K antennas equipped at the AP.

There are two links from the AP to NOMA users. In the first phase, the received signal at the relay is given by

$$y_{RS,R} = \sqrt{P_{RS}} h_{KR} \left( \sqrt{a_1} x_1 + \sqrt{a_2} x_2 \right) + \omega_{R},$$

where $P_{RS}$ is the transmit power of AP, $x_i, i \in \{1, 2\}$ is the information symbol for $D_i$, $a_i$ is the power allocation coefficient for $x_i$ with $a_1 + a_2 = 1$ and $a_1 > a_2$, and $\omega_R$ is the Additive White Gaussian Noise (AWGN) at the relay. Optimization of such power allocation is out of concern in this paper. It can be achieved by the signal to interference plus noise ratio (SINR) at the receivers. In order to decode $x_1$ at R, the corresponding SINR is expressed by

$$\gamma_{S,R,x_1} = \frac{a_1 P_{RS} |h_{KR}|^2}{a_1 P_{RS} |h_{KR}|^2 + N_0}.$$ (2)

In NOMA, SIC is employed to eliminate interference, the SINR to decode $x_2$ is given by

$$\gamma_{S,R,x_2} = \frac{a_2 P_{RS} |h_{KR}|^2}{N_0}.$$ (3)

During time for signal processing in the second phase, R transmits the signal consisting of the decoded and re-encoded symbols to the destinations. The received signal at two users $D_i$ is given by

$$y_{R,D_i} = \sqrt{P_{R}} h_{RD_i} \left( \sqrt{a_1} x_1 + \sqrt{a_2} x_2 \right) + \omega_{D_i},$$

where $P_{R}$ is the transmit power of $R$, $\omega_{D_i}$ is the AWGN at $D_i$.

The signal to interference plus noise ratio (SINR) at each receiver needs to be calculated. In this case, the destination is required to decode $x_i$ at $D_i$ as below

$$\gamma_{D_i,x_i} = \frac{a_i P_{R} |h_{RD_i}|^2}{a_i P_{R} |h_{RD_i}|^2 + N_0}.$$ (5)

The SINR to decode $x_1$ at $D_2$ is given by

$$\gamma_{D_2,x_1} = \frac{a_1 P_{R} |h_{RD_2}|^2}{a_2 P_{R} |h_{RD_2}|^2 + N_0}.$$ (6)

After SIC, the SINR to decode $x_2$ is given by

$$\gamma_{D_2,x_2} = \frac{a_2 P_{R} |h_{RD_2}|^2}{N_0}.$$ (7)

The best channel is selected with the index of antennas

Fig. 1. System Model of multiple power beacon-assisted NOMA.
equipped at the AP as

\[ k^* = \arg \max_{k=1,...,K} \left| h_{ki} \right|^2. \]  

(8)

Together with (8), the cumulative distribution function (CDF) and probability distribution function (PDF) related to selected channels are given as

\[ F_{\left| h_{ki} \right|} (x) = 1 - \sum_{k=1}^{K} \left( \frac{K}{k} \right) \left( 1 - \frac{2}{\lambda_{SR}} \right)^k \exp\left( -\frac{2}{\lambda_{SR}} \right), \]  

(9)

and

\[ f_{\left| h_{ki} \right|} (x) = \sum_{k=1}^{K} \left( \frac{K}{k} \right) \left( 1 - \frac{2}{\lambda_{SR}} \right)^{k-1} \frac{k}{\lambda_{SR}} \exp\left( -\frac{2}{\lambda_{SR}} \right). \]  

(10)

In the considered system, the relay harvests energy from beacons. The operation of the second stage of signal processing is supported by harvesting energy at relay.

At energy harvesting phase, the time switching (TS) based energy harvesting technique is applied. In a transmission block time \( T \) (in which a block of information is sent from the beacon to the relay), the relay takes \( \alpha T \) to harvest energy from the beacon, in which \( \alpha \) is the energy harvesting time fraction that depends on the schedule of \( B \). We allocate the time slot of \( (1-\alpha)T \) into two equal sub-time slots for the link from the AP to relay \( * \) and the link relay-destinations. Then, we can compute harvested energy at the relay as in [10]

\[ E_h = \tau P_h \alpha T \left| h_{bh} \right|^2, \]  

(11)

where \( 0 < \tau < 1 \) stands for the efficiency coefficient of the energy conversion process, \( 0 < \alpha < 1 \) is the percentage of energy harvesting, \( P_h \) is the transmit power of the beacon, assuming that these power beacons have the same power level, respectively (optimizing time switching factor \( \alpha \) is out of the scope of this paper). Under the assumption that the processing energy at \( R \) is negligible, the transmit power of the relay is

\[ P_R = \frac{2\tau P_h \alpha \left| h_{bh} \right|^2}{(1-\alpha)} = \xi \left| h_{bh} \right|^2, \]  

(12)

where \( \xi = \frac{2\tau P_h \alpha}{(1-\alpha)} \).

III. OUTAGE PERFORMANCE ANALYSIS FOR NOMA

In this section, we consider the outage probability analyses for the IoT system to look for impact of harvested energy and the number of transmit antennas at the AP. In particular, we derive the closed-form expressions to show the outage probabilities, and performance difference happens as comparing two users’ performance. To provide insights, asymptotic outage performance analyses for the considered system are determined in the high transmit signal to noise ratio (SNR) region.

A. Outage Analysis at \( D_1 \)

With respect to the system performance evaluation, the outage probabilities can be achieved related to ability to detect signal at relay and destinations as well. The expression of outage probability for the first user can be defined as

\[ OP_1 = \Pr \left( \min_{\gamma_{D_1,R_1}, \gamma_{D_2,R_2}} \gamma_{D_1,R_1} < \gamma_{D_2,R_2} \right) = \]  

\[ = 1 - \Pr \left( \gamma_{D_1,R_1} > \gamma_{D_2,R_2} \right) = \]  

\[ = 1 - \frac{\Pr \left( \gamma_{D_1,R_1} > \gamma_{D_2,R_2} \right)}{\gamma_{D_1,R_1}} \times \frac{\gamma_{D_2,R_2}}{\gamma_{D_1,R_1}}. \]  

(13)

where \( \gamma_i = 2^\nu_i - 1, \ R_i \) is the target rate to decode \( x_i, i = \{1, 2\} \).

Proposition 1: The close-form expression is computed to provide outage analysis at \( D_1 \) as

\[ OP_1 = 1 - \frac{\sum_{k=1}^{K} \left( \frac{K}{k} \right) \left( 1 - \frac{2}{\lambda_{SR}} \right)^k}{\sum_{k=1}^{K} \left( \frac{K}{k} \right) \left( 1 - \frac{2}{\lambda_{SR}} \right)^{k-1} \frac{k}{\lambda_{SR}} \exp\left( -\frac{2}{\lambda_{SR}} \right)} \times \]  

\[ \frac{4\tau_i N_0}{\xi (a_i - \nu_i) \lambda_{BR} \lambda_{RD}} K \left( \frac{4\tau_i N_0}{\xi (a_i - \nu_i) \lambda_{BR} \lambda_{RD}} \right). \]  

(14)

where \( K(.) \) is the second kind of modified Bessel functions.

Proof: Please refer to Appendix A.

B. Outage Analysis at \( D_2 \) with Perfect SIC

At the second user, it can be seen outage performance evaluated by

\[ OP_2 = \Pr \left( \min_{\gamma_{D_1,R_1}, \gamma_{D_2,R_2}} \gamma_{D_1,R_1} < \gamma_{D_2,R_2} \right) = \]  

\[ = 1 - \Pr \left( \gamma_{D_1,R_1} > \gamma_{D_2,R_2} \right) = \]  

\[ = 1 - \frac{\Pr \left( \gamma_{D_1,R_1} > \gamma_{D_2,R_2} \right)}{\gamma_{D_1,R_1}} \times \frac{\gamma_{D_2,R_2}}{\gamma_{D_1,R_1}}. \]  

(15)

Proposition 2: It can be achieved the close-form expression to provide outage analysis perfect SIC at \( D_2 \) in perfect SIC case as

\[ OP_2 = 1 - \frac{\sum_{k=1}^{K} \left( \frac{K}{k} \right) \left( 1 - \frac{2}{\lambda_{SR}} \right)^k \exp\left( -\frac{2}{\lambda_{SR}} \right)}{\sum_{k=1}^{K} \left( \frac{K}{k} \right) \left( 1 - \frac{2}{\lambda_{SR}} \right)^{k-1} \frac{k}{\lambda_{SR}} \exp\left( -\frac{2}{\lambda_{SR}} \right)} \times \]  

\[ \frac{4\tau_i N_0}{\xi (a_i - \nu_i) \lambda_{BR} \lambda_{RD}} K \left( \frac{4\tau_i N_0}{\xi (a_i - \nu_i) \lambda_{BR} \lambda_{RD}} \right). \]  

(16)

Proof: Please refer to Appendix B.

C. Scenario of Imperfect SIC

The SINR to decode \( x_2 \) at the first link AP-relay is given by
The closed-form expression for the outage probability at user $D_1$ is given by

$$\gamma_{S_1,R_1} = \frac{a_2 P_{ls} |\tilde{h}_{s_1,r}|^2}{a_1 P_{ls} g_{s_1,r}^2 + N_0},$$

(17)

where $g_{s_1,r} \sim CN(0, \mu \lambda_{s_1r})$ and $\mu$ is denoted as the level of residual interference caused by imperfect SIC as $0 \leq \mu \leq 1$.

The second link is evaluated via SINR as below. It is used to decode $x_2$ at $D_2$ and such SINR is formulated by

$$\gamma_{D_2,R_2} = \frac{a_2 P_{ls} |\tilde{h}_{s_2,r}|^2}{a_1 P_{ls} g_{s_2,r}^2 + N_0},$$

(18)

where $g_{s_2,r} \sim CN(0, \mu \lambda_{s_2r})$.

The outage performance of user $D_2$ in imperfect SIC case can be written as

$$OP_{2p} = Pr\left(\min(\gamma_{S_1,R_2}, \gamma_{D_2,R_2}) < \nu_2\right) =$$

$$= 1 - Pr\left(\gamma_{S_1,R_2} > \nu_2, \gamma_{D_2,R_2} > \nu_2\right) =$$

$$= 1 - \Pr\left(\gamma_{S_1,R_2} > \nu_2 \times \Pr\left(\gamma_{D_2,R_2} > \nu_2\right)\right) .$$

(19)

**Proposition 3:** The closed-form outage analysis at $D_2$ is formulated (imperfect SIC case) by

$$OP_{2p} = 1 - \sum_{i=1}^{K} \sum_{i=1}^{K} \left(\frac{K}{m}\right)^{i-1} \times$$

$$\times \frac{ma_x \lambda_{SR}}{k \nu_2 a_1 \lambda_{SR}} \exp\left(-\frac{k \nu_2 N_0}{a_1 P_{ls} \lambda_{SR}}\right) \times$$

$$\times \frac{a_2 \tilde{\epsilon}_{RD2}}{\nu_2 \lambda_{RD2} \lambda_{BR}} \times$$

$$\times \frac{4 \nu_2 N_0}{\sqrt{a_2 \tilde{\epsilon}_{RD2} \lambda_{RD}}} K \left(\frac{4 \nu_2 N_0}{\sqrt{a_2 \tilde{\epsilon}_{RD2} \lambda_{RD}}}\right) .$$

(20)

**Proof:** Please refer to Appendix C.

**D. Asymptotic Outage Probability Analysis**

When $P_{ls} \to \infty$, asymptotic performance of $U_1$ and $D_2$ can be obtained. We first look on the asymptotic performance of the first user in terms of outage probability as:

$$OP_{1}^{\text{asy}} = 1 - \frac{4 \nu_1 N_0}{\sqrt{\xi (a_1 - \nu_2 a_2) \lambda_{SR} \lambda_{RD}}} \times$$

$$\times K \left(\frac{4 \nu_1 N_0}{\sqrt{\xi (a_1 - \nu_2 a_2) \lambda_{RD}}}\right) .$$

(21)

$$OP_{2}^{\text{asy}} = 1 - \frac{4 \nu_2 N_0}{\sqrt{\xi (a_1 - \nu_2 a_2) \lambda_{SR} \lambda_{RD}}} \times$$

$$\times K \left(\frac{4 \nu_2 N_0}{\sqrt{\xi (a_1 - \nu_2 a_2) \lambda_{RD}}}\right) .$$

(22)

It is worth noting that perfect SIC and imperfect SIC cases are considered for user $D_2$, respectively, as below

$$OP_{2p} = 1 - \sum_{i=1}^{K} \sum_{i=1}^{K} \left(\frac{K}{m}\right)^{i-1} \times$$

$$\times \frac{ma_x \lambda_{SR}}{k \nu_2 a_1 \lambda_{SR}} \exp\left(-\frac{k \nu_2 N_0}{a_1 P_{ls} \lambda_{SR}}\right) \times$$

$$\times \frac{a_2 \tilde{\epsilon}_{RD2}}{\nu_2 \lambda_{RD2} \lambda_{BR}} \times$$

$$\times \frac{4 \nu_2 N_0}{\sqrt{a_2 \tilde{\epsilon}_{RD2} \lambda_{RD}}} K \left(\frac{4 \nu_2 N_0}{\sqrt{a_2 \tilde{\epsilon}_{RD2} \lambda_{RD}}}\right) .$$

(23)

**IV. OUTAGE PERFORMANCE OF OMA**

In the first phase, the received signal at the first link $AP$-relay is given by

$$\gamma_{S_1,R_1} = \sqrt{P_{ls} h_{s_1,r}} x_i + \omega_1 .$$

(24)

The SINR computed to decode $x_i$ at $R$ is given by

$$\gamma_{S_1,R_1} = \frac{P_{ls} |h_{s_1,r}|^2}{N_0} .$$

(25)

After that, $R$ transmits the mixture signal (the decoded and re-encoded symbols) to the destinations. The received signal at $D_2$ is given by

$$\gamma_{S_2,D_2} = \sqrt{P_{ls} h_{s_2,r}} x_i + \omega_1 .$$

(26)

The SINR to decode $x_i$ at $D_2$ is given by

$$\gamma_{S_2,D_2} = \frac{P_{ls} |h_{s_2,r}|^2}{N_0} .$$

(27)

**The outage probability at $D_2$ can be expressed as**

$$OP_{2p} = Pr\left(\min(\gamma_{S_2,D_2}, \gamma_{OMA,D_2}) < \nu_i\right) =$$

$$= 1 - \Pr\left(\gamma_{OMA,D_2} > \nu_i\right) \times Pr\left(\gamma_{S_2,D_2} < \nu_i\right) =$$

$$= 1 - \Pr\left(h_{s_2,r} > \nu_i N_0 \frac{\gamma_{OMA,D_2}}{P_{ls}} \times \frac{\nu_i h_{s_2,r}}{\xi} \right) =$$

$$= 1 - \sum_{i=1}^{K} \left(\frac{K}{m}\right)^{i-1} \frac{1}{P_{ls} \lambda_{SR}} \times$$

$$\times \left(\frac{K}{m}\right)^{i-1} \frac{1}{P_{ls} \lambda_{RD}} \times$$

$$= 1 - \sum_{i=1}^{K} \left(\frac{K}{m}\right)^{i-1} \frac{1}{P_{ls} \lambda_{SR}} \times$$

$$\times \left(\frac{K}{m}\right)^{i-1} \frac{1}{P_{ls} \lambda_{RD}} \times$$

$$\times \left(\frac{K}{m}\right)^{i-1} \frac{1}{P_{ls} \lambda_{RD}} \times$$

$$\times \left(\frac{K}{m}\right)^{i-1} \frac{1}{P_{ls} \lambda_{RD}} .$$

(28)
V. NUMERICAL RESULTS

In this section, we show the comparisons of IoT system related outage performance of two users using NOMA and OMA. These users are grouped in downlink of the AP using Rayleigh fading channels under different simulated parameters.

The outage probability versus the transmit SNR at the AP is illustrated in Fig. 2, where we consider two main scenarios, i.e., NOMA and OMA. The different power allocation coefficients are assigned to two users, and hence outage performance of the first user is better than that of the second user. It can be easily seen that more antennas result in lowest outage. When the SNR is greater than 30 (dB), outage probabilities for these cases go to straight line. It means that they meet saturation situation.

In addition, imperfect SIC at the second user has worse outage performance compared with perfect case. It is further confirmed that NOMA in IoT is better than in OMA case. The exactness of the asymptotic lines corresponding with derived expressions for all the considered cases is confirmed at high SNR. Similar trend can be seen in Fig. 3 as considering impact of transmit SNR at the power beacon on the outage probability.

Considering outage performance of two users versus transmit SNR at the AP with different power allocation factors as in Fig. 4, the users’ performance change based on the amount of power allocated. Higher $a_1$ leads to better outage performance at the first user. These trends of curves related outage behavior are similar as in Fig. 2 and Fig. 3. While considering how transmit SNR at beacon makes impact on outage probability, it can be seen similar performance as in Fig. 5.
VI. CONCLUSIONS
In this paper, we have investigated the IoT system by enabling energy harvesting and transmit antenna selection schemes. The main result of NOMA scheme provide acceptable outage performance. Such performance is improved significantly at high SNR regime. The relaying scheme with WPT technique benefits to such IoT with performance improvement for two far users who need assistance of WPT-assisted relay. Depending on power allocation factors, different performance of two users can be observed. When SIC can be operated perfectly, it is able to exhibit better performance for the second user. For the antenna selection scheme, it is unnecessary design of multiple antennas system with complex signal processing technique, it can be reduced by exploiting antenna selection as presented in this paper. We derived in closed-form for the outage probability for two distant users in the considered IoT system. More importantly, the asymptotic expressions for the outage probabilities are provided. The superior outage performance achieved by the proposed IoT system is confirmed in numerical results.

APPENDIX A

Proof of the Proposition 1:
By using (9) and (13), \( \varrho_1 \) can be formulated by
\[
\varrho_1 = \Pr \left( \gamma_{S_i,k - n} > \nu_1 \right) = \Pr \left( \frac{h_{S_i,k}}{P_{BS}} > \frac{\nu_1 N_0}{a_i - \nu_1 a_2} \right) = \sum_{k=1}^{K} \left( \frac{K}{k} \right) (-1)^{k-1} \exp \left( -\frac{k \nu_1 N_0}{P_{BS} (a_i - \nu_1 a_2) \lambda_{SR}} \right). \tag{A.1} \]

In similar way, from (10) and (13), it can be obtained \( \varrho_2 \) as
\[
\varrho_2 = \Pr \left( \gamma_{S_i,k - 1} > \nu_1 \right) = \Pr \left( \frac{h_{S_i,k}}{P_{BS}} > \frac{\nu_1 N_0}{a_i - \nu_1 a_2} \right) = \int_{0}^{\infty} \exp \left( -\frac{\nu_1 N_0}{\xi (a_i - \nu_1 a_2) \lambda_{RD2}} \right) \frac{1}{\lambda_{BB}} \exp \left( -\frac{x}{\lambda_{BB}} \right) dx = \frac{1}{\lambda_{BB}} \int_{0}^{\infty} \exp \left( -\frac{\nu_1 N_0}{\xi (a_i - \nu_1 a_2) \lambda_{RD2}} x \right) - \frac{x}{\lambda_{BB}} dx = \frac{4 \nu_1 N_0}{\xi (a_i - \nu_1 a_2) \lambda_{RD2} \lambda_{BB}} K_i \left( \frac{4 \nu_1 N_0}{\xi (a_i - \nu_1 a_2) \lambda_{RD2}} \right). \tag{A.2} \]

It is worth noting that the last equation follows from the fact that \( \int_{0}^{\infty} \exp \left( -\frac{\delta - \phi x}{4x} \right) dx = \sqrt{\frac{\delta}{\phi}} K_i \left( \sqrt{\delta \phi} \right) \) in [19, eq. (3.324)].

If we plug (A.1), (A.2) into (13), \( OP_1 \) as the proposition can be achieved.
This is end of the proof.

APPENDIX B

Proof of the Proposition 2:
Using (9) and (15), \( \psi_i \) can be computed as below
\[
\psi_i = \Pr \left( \gamma_{S_i,k - n} > \nu_2 \right) = \Pr \left( \frac{h_{S_i,k}}{P_{BS}} > \frac{\nu_2 N_0}{a_i - \nu_2 a_2} \right) = \sum_{k=1}^{K} \left( \frac{K}{k} \right) (-1)^{k-1} \exp \left( -\frac{k \nu_2 N_0}{a_i - \nu_2 a_2} \right). \tag{B.1} \]

Next, (15) is used to calculate \( \psi_2 \) as
\[
\psi_2 = \Pr \left( \gamma_{S_2,n} > \nu_3, \gamma_{S_2,n} > \nu_3 \right) = \Pr \left( \frac{h_{S_2,n}}{P_{BS}} > \frac{\nu_3 N_0}{a_i - \nu_3 a_2} \right) = \frac{1}{\xi} \int_{0}^{\infty} \exp \left( -\frac{\nu_3 N_0}{\xi (a_i - \nu_3 a_2) \lambda_{RD2}} x \right) - \frac{x}{\lambda_{BB}} dx = \frac{4 \nu_3 N_0}{\xi (a_i - \nu_3 a_2) \lambda_{RD2} \lambda_{BB}} K_i \left( \frac{4 \nu_3 N_0}{\xi (a_i - \nu_3 a_2) \lambda_{RD2}} \right). \tag{B.3} \]

Replacing (B.1), (B.3) into (15), \( OP_2 \) can be obtained. This is end of the proof.

APPENDIX C

Proof of the Proposition 3:
Plugging (9) and (19) to corresponding result, \( \psi_i \) can be written by
\[
\psi_i = \Pr \left( \gamma_{S_i,k - n} > \nu_2 \right) = \Pr \left( \frac{h_{S_i,k}}{P_{BS}} > \frac{\nu_2 a_i P_{BS} |S_{i,k}|^2 + N_0}{a_i P_{BS}} \right) = \frac{1}{\xi} \int_{0}^{\infty} \exp \left( -\frac{\nu_2 a_i P_{BS} \phi + N_0}{a_i P_{BS}} \right) - \frac{x}{\lambda_{BB}} dx. \tag{C.1} \]
Using (9) and (10), then further employs corresponding CDF and PDF, $\psi_1$ can be computed by

$$
\psi_1 = \sum_{k=1}^{K} \sum_{m=1}^{K} \frac{K}{m} (-1)^{k-m-1} \frac{m}{\lambda_{SR}} \times
\exp\left(-\frac{kv_0N_0}{a_{i}P_{BR}^2\lambda_{SR}}\right) \int_{0}^{\infty} \exp\left(-\frac{kv_0a_{i}^2}{a_{z}\lambda_{SR}} + \frac{m}{\lambda_{SR}}\right) x \, dx =
\sum_{k=1}^{K} \sum_{m=1}^{K} \frac{K}{m} (-1)^{k-m-1} \frac{m}{\lambda_{SR}} \times
\exp\left(-\frac{kv_0N_0}{a_{i}P_{BR}^2\lambda_{SR}}\right).
$$

(C.2)

Based on (19), we compute $\psi_2$ as below

$$
\psi_2 = \Pr\left(\gamma_{SR} > \gamma_2\right) =
\Pr\left(\frac{h_{BR}^2}{\gamma_{BR}} > \frac{\nu_2a_{z}^2\gamma_{BR}}{\nu_2^2a_{z}^2\gamma_{RD}^2 + N_0}\right) =
\int_{0}^{\infty} \int_{0}^{\infty} \left(1 - F_{\gamma_{BR}}(\gamma)\right) \times
f_{\gamma_{RD}^2}(x) f_{\gamma_{BR}}(y) dy dx.
$$

(C.3)

In next step, using results of CDF and PDF, $\psi_2$ is formulated as

$$
\psi_2 = \frac{1}{\gamma_{BR}^2} \frac{1}{\gamma_{RD2}^2} \int_{0}^{\infty} \left(-\frac{\nu_2a_{z}^2\gamma_{RD2}}{\nu_2a_{z}^2\gamma_{BR}} + \frac{1}{\gamma_{RD2}^2}\right) y \, dy \times
\int_{0}^{\infty} \exp\left(-\frac{\nu_2N_0}{\nu_2a_{z}^2\gamma_{RD2}} - \frac{x}{\gamma_{BR}^2}\right) dx =
\frac{a_{z}^2\gamma_{RD2}}{\nu_2a_{z}^2\gamma_{RD2}^2 + \nu_2a_{z}^2\gamma_{BR}^2} \times
\frac{4\nu_2N_0}{a_{z}^2\gamma_{RD2}^2\gamma_{BR}} K_1\left(\frac{4\nu_2N_0}{a_{z}^2\gamma_{RD2}^2\gamma_{BR}}\right).
$$

(C.4)

Plugging (C.2), (C.4) into (19), $\Omega_{2p}^*$ can be achieved. This completes the proof.

CONFLICTS OF INTEREST

The authors declare that they have no conflicts of interest.

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