

Article



# **Cognitive Radio Networks Using Intelligent Reflecting Surfaces**

Raed Alhamad\*

Information Technology Department, Saudi Electronic University, Riaydh, Kingdom of Saudi Arabia \*Corresponding Author: Raed Alhamad. Email: ralhamad@seu.edu.sa Received: 20 July 2021; Accepted: 26 September 2021

Abstract: In this article, we optimize harvesting and sensing duration for Cognitive Radio Networks (CRN) using Intelligent Reflecting Surfaces (IRS). The secondary source harvests energy using the received signal from node A. Then, it performs spectrum sensing to detect Primary Source Ps activity. When Ps activity is not detected, The Secondary Source S<sub>S</sub> transmits data to Secondary Destination  $S_D$  where all reflected signals on IRS are in phase at  $S_D$ . We show that IRS offers 128, 256, 512 reflectors with respect to CRN without IRS. We also suggested to add a second IRS between A and  $S_S$  to increase the harvested energy. The use of 2 IRS with M1 = 8 reflectors in the first IRS and M2 = 8 reflectors in the second IRS offers 18 dB gain (respectively 32 dB) gain with respect to a single IRS with M2 = 8 reflectors (respectively without IRS). The use of 2 IRS with M1 =16 reflectors in the first IRS and M2 = 8 reflectors in the second IRS offers 28 dB gain (respectively 42 dB) gain with respect to a single IRS with M2 = 8 reflectors (respectively without IRS). Our results are valid for Nakagami channels of fading figure *m*. We also provide the throughput of IRS with energy harvesting. We have studied packet waiting time and total delay in the presence and absence or IRS. At Signal to Noise Ratio (SNR) per bit equal to 0 dB, packet waiting time is 0.9 ms when there is no IRS and 0.5 ms when there is an IRS with M = 8 reflector. At SNR per bit equal to 0 dB, total transmission delay is 54 ms when there is no IRS and 1.5 ms when there is an IRS with M = 8 reflectors. We show that the energy efficiency is larger when both harvesting and sensing durations are optimized. The maximum of energy efficiency is 260 Mbit/s/Hz/J when harvesting and sensing durations are optimized while the maximum is 80 Mbit/s/Hz/J when harvesting and sensing durations are not optimized.

**Keywords:** Cognitive radio networks; energy harvesting; spectrum sensing; throughput analysis

## **1** Introduction

Intelligent Reflecting Surfaces (IRS) are able to enhance significantly the throughput of wireless systems [1-5]. The phase shift of each IRS reflector is optimized so that reflections are in phase at the destination [6-8]. IRS is placed between the source and the destination so that all reflections have the same phase at



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the destination [9,10]. The number of reflectors has been varied and takes values M = 8, 16, 32, 64, 128, 256, 512 and the throughput increases as M increases [11–13]. Asymptotic performance analysis of wireless systems using IRS was discussed in [14–16]. Antenna design and measurement results were discussed in [17]. Machine and deep learning algorithms were applied to wireless communications using IRS [18,19].

In this paper, we suggest to optimize harvesting and sensing duration for CRN using IRS. The Secondary Source  $S_S$  harvests energy during a T seconds where T is frame duration and  $0 \le a \le 1$  is the harvesting duration. Then S<sub>S</sub> performs spectrum sensing using the energy detector over (1-a)bT seconds to detect Primary Source  $P_S$  activity where 0 < b < 1 provides the sensing duration. When  $P_S$  activity is not detected, SS transmits data to the Secondary Destination  $S_D$  during (1-a)(1-b)T. We show that IRS offers 14, 20, 26, 32, 38, 44, 50 dB enhancement in throughput using M=8, 16, 32, 64, 128, 256, 512 reflectors with respect to CRN without IRS [20]. We also suggested to add a second IRS between A and  $S_S$  to increase the harvested energy. The use of 2 IRS with M1 = 8 reflectors in the first IRS and M2 = 8 reflectors in the second IRS offers 18 dB gain (respectively 32 dB) gain with respect to a single IRS with M2 = 8 reflectors (respectively without IRS [20]). The use of 2 IRS with M1 = 16 reflectors in the first IRS and M2 = 8 reflectors in the second IRS offers 28 dB gain (respectively 42 dB) gain with respect to a single IRS with M2 = 8 reflectors (respectively without IRS [20]). We also derive the Probability Density Function (PDF), the Cumulative Distribution Function (CDF) of SNR and the throughput of CRN using IRS where the secondary source harvests energy. Beamforming allows increasing the harvested energy as suggested in [21,22]. The weighting coefficient of different antennas can be optimized to maximize the harvested energy. Beamforming requires multiple antennas as proposed in [21,22]. In this paper, a single antenna is used at the secondary source.

Next section optimizes the throughput when there is a single IRS. Section 3 proposes the use of a second IRS to increase the harvested energy. Section 4 describes the numerical results and Section 5 concludes the paper.

#### 2 A Single IRS

The system model of Fig. 1 contains a Primary Source and a Primary Destination  $P_S$  and  $P_D$ , a Secondary Source and Destination  $S_S$  and  $S_D$ , node A and IRS.  $S_S$  harvests energy over a T seconds. T is frame duration and 0 < a < 1 is the harvesting duration. Then,  $S_S$  performs spectrum sensing to detect  $P_S$  activity during (1-a) bT seconds where 0 < b < 1 provides the sensing duration. When  $P_S$  activity is not detected,  $S_S$  transmits a packet of symbols to  $S_D$  over (1-a)(1-b)T seconds. A Nakagami channel of fading figure m is assumed.



Figure 1: System model

The harvested energy at S<sub>S</sub> is equal to

$$E = \delta a T P_A |f|^2 = \delta a \frac{T}{T_s} E_A |f|^2, \tag{1}$$

where  $\delta$  is the efficiency of energy conversion process, Ts is symbol duration,  $P_A = E_A/Ts$  is the power of A, f is channel gain between A and  $S_S$ . We have  $E(|f|^2) = 1/D_1^{PLE}$  where  $D_1$  is the distance between A and  $P_S$ , E(.) is the expectation operator and PLE is the Path Loss Exponent.

The symbol energy of S<sub>S</sub> is computed as:

$$E_{S_S} = \frac{E}{\frac{T}{T_s}(1-a)(1-b)} = \delta \frac{a}{(1-a)(1-b)} E_A |f|^2$$
(2)

Let  $h_q$  the channel gain between  $S_S$  and q-th reflector of IRS. Let  $g_q$  the channel gain between q-th reflector of IRS and  $S_D$ . We have:  $E(|h_q|^2) = 1/D2^{PLE}$  where D2 is the distance between  $S_S$  and IRS.  $E(|g_q|^2) = 1/D3^{PLE}$  where D3 is the distance between IRS and  $S_D$ .

We have  $hq = a_q e^{-jbq}$  where  $a_q = |h_q|$  and bq is the phase of hq:  $E(aq) = \Gamma(m + 0.5)/(\Gamma(m) * \sqrt{MD2PLE})$  and  $E(aq^2) = E(|hq|^2) = 1/D2^{PLE}$  [23]. We have  $gq = c_q e^{-jdq}$  where  $cq = |g_q|$  and dq is the phase of gq. We have  $E(cq) = 1 = \Gamma(m + 0.5)/(\Gamma(m) * \sqrt{MD3PLE})$  and  $E(cq^2) = E(|gq|^2) = 1/D3^{PLE}$  [23].

The phase of q-th IRS reflector is given by [1]

$$\varphi_q = b_q + d_q,\tag{3}$$

The received signal at S<sub>D</sub> is given by

$$r_p = s_p \sqrt{E_{S_S}} \sum_{q=1}^M h_q g_q e^{j\varphi_q} + n_p, \tag{4}$$

where sp is the p-th symbol and np is a Gaussian noise of variance N0.

Using (3), we have

$$r_p = s_p \sqrt{E_{\mathcal{S}_S}} \sum_{q=1}^M a_q c_q + n_p \tag{5}$$

The SNR at  $S_D$  is computed as [1]

$$\gamma^{S_D} = \frac{E_{S_S}}{N_0} X^2,\tag{6}$$

where

$$X = \sum_{q=1}^{M} a_q c_q,\tag{7}$$

Using (2), we deduce

$$\gamma^{S_D} = \frac{\delta a E_A |f|^2}{(1-a)(1-b)N_0} X^2,$$
(8)

X has a Gaussian distribution with mean and variance

$$m_X = \frac{M\Gamma(m+0.5)^2}{m\Gamma(m)^2 D_3^{PLE/2} D_2^{PLE/2}}$$
$$\sigma_X^2 = \frac{M}{D_3^{PLE/2} D_2^{PLE/2}} \left[ 1 - \frac{\Gamma(m+0.5)^4}{M^2 \Gamma(m)^4} \right]$$

Therefore,  $X^2$  has a non central chisquare distribution with one degree of freedom. For Nakagami channels,  $|f|^2$  has a central chi-square distribution with 2 m degrees of freedom. The SNR is the product of a non central chisquare r.v.  $X^2$  and  $f|^2$  has a central chi-square distribution with 2 m degrees of freedom. The CDF of SNR at S<sub>D</sub> is given by [23,24]

$$F_{\gamma^{S_D}}(x) = \frac{e^{-0.5\frac{m_X^2}{\sigma_X^2}}}{\Gamma(m)} \sum_{q=0}^{+\infty} \frac{\left(\frac{m_X^2}{\sigma_X^2}\right)^q}{2^q \Gamma(q+0.5)} G_{1,3}^{2,1} \left(\frac{N_0(1-a)(1-b)xmD_1^{PLE}}{2\delta a E_A} \mid \frac{1}{q+0.5, m, 0}\right),\tag{9}$$

when  $G_{n,m}^{\ \ p,l}(x)$  is the Meijer G-function.

The Packet Error Probability (PEP) at S<sub>D</sub> can be computed as [25]

$$PEP(a, b) < F_{\gamma^{S_D}}(W_0), \tag{10}$$

where

$$W_0 = \int_0^{+\infty} pep(u)du,\tag{11}$$

and pep(u) is the PEP for Quadrature Amplitude Modulation (QAM) of size Q [26]

$$pep(\mathbf{u}) = 1 - \left[1 - 2\left(1 - \frac{1}{\sqrt{Q}}\right)erfc\left(\sqrt{\frac{3ulog_2(Q)}{2Q - 2}}\right)\right]^{PL},\tag{12}$$

where PL is packet length in symbols.

The throughput at SD is computed as

$$Thr(a, b) = (1-b)(1-a)log_2(Q)P_{idle}[1-P_f(a, b)][1-PEP(a, b)]$$
(13)

where B is the used bandwidth, Pf(a, b) is the false alarm probability written as

$$P_f(a, b) = \frac{\Gamma\left(\left\lfloor (1-a)b\frac{T}{T_s}\right\rfloor, \mu/2\right)}{\Gamma\left(\left\lfloor (1-a)b\frac{T}{T_s}\right\rfloor\right)},\tag{14}$$

 $\mu$  is the energy detector threshold and

$$\Gamma(N, u) = \int_{u}^{+\infty} x^{N-1} e^{-x} dx,$$
(15)

Harvesting duration a and sensing duration b are optimized to maximize the throughput:

$$Thr^{max} = max_{a,b}Thr(a, b), \tag{16}$$

## **3 Two IRS**

In Fig. 2, IRS1 is placed between A and  $S_S$  to increase the harvested energy while IRS<sub>2</sub> is located between  $S_S$  and  $S_D$  so that all reflected signals are in phase at  $S_D$ .



Figure 2: IRS to increase the harvested energy

When energy harvesting uses IRS, the harvested energy is expressed as

$$E = \delta a \frac{T}{T_s} E_A Z^2, \tag{17}$$

where

$$Z = \sum_{q=1}^{M_1} \delta_q \rho_q \tag{18}$$

M1 is the number of reflectors of the first IRS,  $\delta q = |uq|$ , uq is the channel gain between A and IRS1 and  $\rho q = |vq|$ , vq is the channel gain between IRS1 and S<sub>S</sub>.

Z has a Gaussian distribution with mean and variance

$$m_{Z} = \frac{M_{1}\Gamma(m+0.5)^{2}}{m\Gamma(m)^{2}D_{5}^{PLE/2}D_{4}^{PLE/2}}$$
$$\sigma_{Z}^{2} = \frac{M_{1}}{D_{4}^{PLE/2}D_{5}^{PLE/2}} \left[1 - \frac{\Gamma(m+0.5)^{4}}{M_{1}^{2}\Gamma(m)^{4}}\right]$$

D4 is the distance between A and IRS1 and D5 is the distance between IRS1 and  $S_s$ .

We deduce

$$E_{S_S} = \frac{E}{\frac{T}{T_s}(1-a)(1-b)} = \delta \frac{a}{(1-a)(1-b)} E_A Z^2$$
(19)

The SNR at S<sub>D</sub> is equal to

$$\gamma^{S_D} = \frac{E_{S_S}}{N_0} X^2 = \frac{\delta a E_A Z^2}{(1-a)(1-b)N_0} X^2,$$
(20)

where X is defined in (7).

The CDF of SNR at SD is equal to [23,24]

$$F_{\gamma^{S_D}}(x) = e^{-0.5\frac{m_Z^2}{\sigma_Z^2}} e^{-0.5\frac{m_X^2}{\sigma_X^2}} \sum_{q=0}^{+\infty} \sum_{l=0}^{+\infty} \frac{\left(\frac{m_Z^2}{\sigma_Z^2}\right)^l \left(\frac{m_X^2}{\sigma_X^2}\right)^q}{2^{q+l}\Gamma(q+0.5)\Gamma(l+0.5)q!l!}$$

$$G_{1,3}^{2,1}\left(\frac{N_0(1-a)(1-b)xm}{\delta aE_A 2} \mid \frac{1}{q+0.5, \ l+0.5, \ 0}\right),$$
(21)

The throughput is computed and optimized using (10)–(16).

#### **4** Numerical Results

Figs. 3–5 depict the throughput for Quadrature Phase Shift Keying (QPSK) modulation, 16 and 64 QAM modulations, m = 2,  $\mu = 1$ ,  $E_A = 1$ , D1 = 1.1, D2 = 1.2, D3 = 1.4, PLE = 3. A single IRS with M = 8 reflectors was used. The distance between S<sub>S</sub> and P<sub>S</sub> is 2. We plotted the theoretical throughput and the computer simulations (sim). We observe that optimal a b offers more than 4 dB gain with respect to a = 1/3 and optimal b. The proposed optimal a and b offers up to 20 dB gain with respect to optimal a and b = 1/2. A significant throughput enhancement is observed when optimizing a and b with respect to a = 1/3 and b = 1/2 where harvesting, sensing and data transmission are performed over the same durations.



**Figure 3:** Throughput for QPSK and M = 8



**Figure 5:** Throughput for 64 QAM and M = 8

For the same parameters as Figs. 3–5, Figs. 6 and 7 depict the throughput for 16 and 64 QAM modulation and different number of reflecting surfaces M=8, 16, 32, 64, 128, 256, 512. Sensing and harvesting duration were optimized in Figs. 6 and 7. We observe 14, 20, 26, 32, 38, 44, 50 dB enhancement in throughput using M=8, 16, 32, 64, 128, 256, 51 reflectors with respect to CRN without IRS [20].



Figure 6: Throughput for 16 QAM and different values of M



Figure 7: Throughput for 64 QAM and different values of M

Fig. 8 shows the throughput of 16 QAM modulation for the same parameters as Figs. 3–5 and when there are 2 IRS for D4=1.1 and D5=1.2. The use of 2 IRS with M1=M2=8 reflectors offers 18 dB gain (respectively 32 dB) gain with respect to a single IRS with M2=8 reflectors (respectively without IRS [20]). The use of 2 IRS with M1=16 reflectors in the first IRS and M2=8 reflectors in the second IRS offers 28 dB gain (respectively 42 dB) gain with respect to a single IRS with M2=8 reflectors (respectively 42 dB) gain with respect to a single IRS with M2=8 reflectors in the second IRS offers 28 dB gain (respectively 42 dB) gain with respect to a single IRS with M2=8 reflectors (respectively without IRS [20]).



Figure 8: Throughput for 16 QAM with 2 IRS and m = 2

Fig. 9 shows the effects of fading figure m = 1, 2, 3 on throughput of 64 QAM modulation for the same parameters as Figs. 3–5. There are M = 8 reflectors in a single IRS. We observe a throughput enhancement as m increases. Note that m = 1 corresponds to Rayleigh channels.



Figure 9: Effects of fading figure m on throughput for 16 QAM modulation

Figs. 10 and 11 depict the throughput for 8 Phase Shift Keying (8-PSK) and 32 Amplitude Shift Keying (ASK) modulations. There is a single IRS with M = 8 reflectors. Optimal harvesting and sensing durations offers the largest throughput for all modulations studied.



**Figure 10:** Throughput for 8 PSK modulation and M = 8



**Figure 11:** Throughput for 32 ASK modulation and M = 8

Fig. 12 depicts the energy efficiency vs. the throughput. The energy efficiency is defined as the throughput in bit/s/Hz divided by the spent energy in J. It is observed that the proposed optimal

harvesting and sensing durations offers the largest energy efficiency. The maximum of energy efficiency is 260 Mbit/s/Hz/J when harvesting and sensing durations are optimized while the maximum is 80 Mbit/s/Hz/J when harvesting and sensing durations are not optimized.



Figure 12: Energy efficiency for 32 ASK modulation and M = 8

Figs. 13–16 show packet waiting time and total delay in the presence and absence or IRS. Packet arrival rate is 0.01 and frame duration 1 ms. At SNR per bit equal to 0 dB, packet waiting time is 0.9 ms when there is no IRS and 0.5 ms when there is an IRS with M=8 reflector. At SNR per bit equal to 0 dB, total transmission delay is 54 ms when there is no IRS and 1.5 ms when there is an IRS with M=8 reflectors and optimized harvesting and sensing durations.



Figure 13: Packet's waiting time without IRS



Figure 14: Total delay without IRS







Figure 16: Total delay with IRS and M = 8 reflectors

#### **5** Conclusion and Perspectives

In this paper, we optimized harvesting and sensing duration for CRN using Intelligent Reflecting Surfaces (IRS).  $S_S$  performs spectrum sensing to detect primary activity. When primary source activity is not detected,  $S_S$  sends data to Secondary Destination  $S_D$  where all reflected signals on IRS are in phase at  $S_D$ . We observed 14, 20, 26, 32, 38, 44, 50 dB enhancement in throughput using M = 8, 16, 32, 64, 128, 256, 512 reflectors with respect to CRN without IRS [20]. We also suggested to add a second IRS between A and  $S_S$  to increase the harvested energy. The use of 2 IRS with M1 = M2 = 8 reflectors offers 18 dB gain (respectively 32 dB) gain with respect to a single IRS with M2 = 8 reflectors in the second IRS offers 28 dB gain (respectively 42 dB) gain with respect to a single IRS with M2 = 8 reflectors (respectively without IRS [20]). As a perspective, we can optimize harvesting and sensing durations for CRN using Non Orthogonal Multiple Access (NOMA). As a perspective, we can increase the harvested energy using beamforming when the secondary source has multiple transmit antennas.

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