

Lodro, Mir Muhammad and Greedy, Steeve and Mahoto, Naeem A. and Unar, Mukhtiar A. (2017) Throughput analysis of full-duplex communication cognitive radio network. Wireless Personal Communications, 97 (2). pp. 3081-3095. ISSN 1572-834X

Access from the University of Nottingham repository:

http://eprints.nottingham.ac.uk/44722/1/Mir%20Lodro%20article.pdf

Copyright and reuse:

The Nottingham ePrints service makes this work by researchers of the University of Nottingham available open access under the following conditions.

This article is made available under the Creative Commons Attribution licence and may be reused according to the conditions of the licence. For more details see: http://creativecommons.org/licenses/by/2.5/

A note on versions:

The version presented here may differ from the published version or from the version of record. If you wish to cite this item you are advised to consult the publisher's version. Please see the repository url above for details on accessing the published version and note that access may require a subscription.

For more information, please contact eprints@nottingham.ac.uk



Throughput Analysis of Full-Duplex Communication Cognitive Radio Network

Mir Muhammad Lodro¹ · Steeve Greedy² · Naeem A. Mahoto³ · Mukhtiar A. Unar³

© The Author(s) 2017. This article is an open access publication

Abstract In this paper we deal with the throughput of full-duplex cognitive communication radio which exploits unused band of primary user (PU) network. Classical cognitive radio uses half-duplex communication spectrum sensing to perform spectrum sensing and data transmission at different time intervals. It's well-established fact that in half-duplex communication cognitive radio spectrum sensing time increases at low SNR which gives rise to lesser data transmission time for secondary user (SU) and hence results in less throughput for SU. It's useful idea to do spectrum sensing and data transmission at the same time with two different antennas co-located on the SU transceiver. This shall not only guarantee high probability of detection of PU but also increased data transmission which means more throughput for SU. However, simultaneous sensing and data transmission has inherent problem of self-interference. One of the possible solution is to use polarisation discrimination in which sensing and data transmission antennas must use different polarisation. This is feasible if there is prior information about the polarisation of the signals emitted by the PUs. It shall be of special interest to assess throughput using analytical expressions for probability of detection P_D , probability of false alarm P_{FA} at various values of SNR for time-slotted cognitive radio which uses half-duplex spectrum sensing and nontime-slotted cognitive radio which uses full-duplex communication cognitive radio.

Keywords Full-duplex cognitive radio \cdot Spectrum sensing \cdot Sensing-throughput \cdot TV white space

Mir Muhammad Lodro mir.lodro@iba-suk.edu.pk; mir.lodro@nottingham.ac.uk

¹ The Department of Electrical Engineering, Sukkur Institute of Business Administration, Sukkur 65200, Sindh, Pakistan

² The University of Nottingham, Nottingham, UK

³ Mehran UET Jamshoro, Jamshoro, Pakistan

1 Introduction

Recently because of the influx of smartphone and due to digital dividend there has been increased trend for application that demand high data rates. Due to digital dividend and shift in user pattern a lot of under-utilised spectrum is observed typically in VHF and UHF band of electromagnetic spectrum and such sparsely used bands are also known as TV white spaces (TVWS). Cognitive radio (CR) has been considered as a powerful technique to increase the spectral efficiency by enabling unlicensed secondary users to access underutilized spectrum opportunistically. Federal Communication Commission (FCC) in USA and Office Communication (Ofcom) in UK have legalised the usage of TVWS in VHF and UHF part of the spectrum. In USA non-contiguous TVWS are located in frequency range of 54-698 MHz and portion of frequency ranging from 470 MHz to 790 MHz exists in Europe. There are some new standards which involve cognitive radio principle few to mention are wireless metropolitan area network (WMAN) IEEE 802.22 for broadband applications in remote areas, wireless local area network (WLAN) standard known as IEEE 802.11af which uses master and slave architecture and notion of white space database, IEEE 802.15.4m an low data rate short range wireless personal area network (WPAN) system with cognitive radio and a host of other network types shall experience cognitive radio based transformation. Performance of cognitive radio is assessed by the type of detector it uses and the number of sub-bands it can harness and relinquish if the incumbent user reactivates and reclaims its licensed band. A number of detectors has been considered for the implementation in CR transceivers for example energy detector [1, 2], matched filter detector [3, 4], Eigen-value based detector [5, 6], detectors involving cyclostationarity [7] and wavelet-based detection [8] etc.

2 Background Study

The problem of reactivation failure is associated with the non-time-slotted cognitive radio networks (NTS-CRN) where the PU and SU traffic is not synchronised i.e. PU randomly leaves and reclaims its licensed channel. In time-slotted CRN PU and SU traffic is synchronised i.e. SU transmits after its senses that there is no incumbent user present on that band. In non-time-slotted CRN PU may sense the busy channel since it can't distinguish between PUs traffic and SU traffic and thus goes into back off stage and retransmits with persistent carrier sense multiple access (p-CSMA) which might result in collision and ultimately affect the PU throughput if the SU doesn't sense the PU reactivation and leave licensed PU sub-channel. For time-slotted CRN we use half-duplex spectrum sensing scheme and medium access control (MAC) protocol. Similarly, for non-time-slotted CRN a novel full-duplex spectrum sensing scheme and a novel cognitive medium access control (CMAC) protocol can be used.

Liang et al. [9] developed seminal expression on sensing-throughput of cognitive radio where he found optimal sensing period which achieves maximum throughput and reference [10] proposes full-duplex spectrum sensing algorithms for non-time-slotted full-duplex cognitive radio. Because of ubiquitous nature of OFDM based communication systems work in [11] focuses on sensing-throughput tradeoff of the OFDM based cognitive radio and the probability of false alarm expression in Nakagami-m and Rician fading channels. Reference [12] studies the interference caused by SU due to missed detection of PU and the extent to which PU is affected in terms of capacity-outage performance under Rayleigh and

Nakagami fading channels and suggests beacon placement at transmitter and receiver. Novel and computationally efficient detection algorithm for OFDM based PUs is discussed in [13] where the performance of various detectors in AWGN and Rayleigh fading channels is also assessed. Multi-antenna sensing and Generalized Likelihood Ratio (GLR) detector is investigated when channel gain, noise variance and PU signal values are unknown [14]. Ghassemi et al. [15] proposes peak-to-average-power ratio (PAPR) reduction in non-contiguous OFDM (NC-OFDM) using selected mapping (SLM) sequence which results in out-of-band (OOB) reduction and improved PAPR and bit error rate (BER) for NC-OFDM signals, whereas PAPR reduction in OFDM based CR using clockwise subcarrier activation is presented in [16] and a novel signal cancellation method is proposed in [17] for joint PAPR reduction and sidelobe suppression in NC-OFDM based CR network. Work in [18] deals with cooperative spectrum sensing with energy-detector over multipath fading and shadowing with different fusion strategies and using single and multiple cognitive relays. Work in [19] calculates energy efficiency versus optimal sensing time for CR network. Reference [20] presents the cooperative wideband spectrum sensing over fading channels and models the aliasing effects of sub-Nyquist sampling. Yunfei Chen [21] gave an energy detector with enhanced performance for random signals in Gaussian noise. Energy efficient cognitive radio for joint spectrum sensing and data transmission which uses co-located layer of relays with amplify-and-forward strategy to relay data from source to destination has been proposed in [22].

The remainder of the paper is structured as follows: Sect. 3 explains Full-duplex communication for non-time-slotted cognitive radio and the control model for NTS-CR and TS-CR. It also presents the Full-duplex spectrum sensing algorithm for FDC-CR with which it declares the absence and presence of PU. Section 1 defines the system model for NTS-CR and presents the probability of detection and probability of false alarm used for receiver operating characteristic curve (ROC) to see the detector performance and the throughput for secondary user. Following this Sect. 5 compares the throughput of half-duplex and full-duplex communication cognitive radio. Finally, Sect. 6 contains the numerical results on probability of the HDC and FDC cognitive radio.



to access the licensed TV band



Fig. 1 Multi-channel cognitive radio network model



Fig. 2 Control model for NTS-CR

3 Full-Duplex Communication Non-time Slotted CR

In NTS-CR using FDC-SS we use two antennas for sensing and transmission simultaneously in contrast to TS-CR in which one antenna is used for periodic sensing and transmission. In TS-CR MAC frame is divided into sensing-time and data transmission time. It is a proven fact that PU protection is guaranteed at increased sensing time which reduces the data transmission time, thereby affecting the throughput of SU. In the following section we discuss problem of reactivation failure for NTS-CR network (NTS-CRN).

3.1 NTS-CRN Control Model

In TS-CRN PU traffic and SU traffic is synchronous i.e. SU scans for the availability of under-utilised channel in its fixed sensing duration and quits after it performs data transmission in a fixed transmission time. Figure 2 shows control model of the NTS-CRN where we can see the PU traffic and SU traffic is asynchronous because PU and SU don't quit and reclaim the channel within fixed time interval. In NTS-CRN we divide transmission period of the SU into P sensing periods for sensing the channel with different antenna.

3.2 Reactivation Failure of NTS-CRN

In NTS-CRN since the PU and SU arrive and quite channel randomly therefore collision may take place if the PU wants to reclaim licensed channel when SU is transmitting. In this scenario PU can't recognise the traffic of peer PU or SU i.e. whether licensed channel is occupied by peer PU or SU, therefore PU goes in back-off stage and transmits with p-CSMA. If the licensed channel is occupied by SU then collision takes place and PU goes into back-off stage and senses the channel but if the network time is expired PU discards the traffic. PU's such failure to reclaim its licensed channel is known as reactivation failure. This problem of reactivation failure can be solved with FDC-SS algorithm as shown in Algorithm 1 in which SU senses for the arrival of PU during transmission phase.

Algorithm 1 FDC-SS Scheme [10]				
1: TS-CR sensing period (SP) is divided into V sub-periods				
2: if TS-CR attempts the idle channel then				
3: TS-CR uses legacy HDC-SS scheme within each sens-				
ing period				
4: else				
5: use channel Ψ_1				
6: The SU senses the channel Ψ_1 with FDC-SS within				
each sensing period				
7: end if				

3.3 FDC-SS Scheme

Figure 4 shows non-slotted multi-channel cognitive radio network model where PUs use their L licensed sub-channels and M SUs opportunistically contend for the PU licensed sub-channels.

4 FDC-SS System Model

We can write the sampled output of CR employing energy-detector as follows:

$$y(n) = \sqrt{khs(n) + w(n) n} = 0, 1, 2, \dots, N-1$$
(1)

where *h* is channel gain between PU and SU and has Rayleigh distribution. *s*(*n*) is sampled PU signal, *w*(*n*) is sampled AWGN with zero mean and variance σ_w^2 i.e. $\mathcal{N} \sim (0, \sigma_w^2)$ and *k* is self-interference introduced because of the simultaneous sensing and transmission that takes place in NTS-CR. Where k = 0 means maximum self-interference is introduced and no self-interference for k = 1. The test statistics using FDC-SS at the output of the energy detector of NTS-CR can be given as in Eq. 2.

$$T_{N}(y) = \begin{cases} \frac{1}{N} \left(\sum_{n=0}^{u} |\sqrt{k}hs(n) + w(n)|^{2} + \sum_{n=u+1}^{N} |w(n)|^{2} \right), & \text{if } \mathcal{H}_{10} \\ \frac{1}{N} \left(\sum_{n=0}^{v} |w(n)|^{2} + \sum_{n=v+1}^{N} |\sqrt{k}hs(n) + w(n)|^{2} \right), & \text{if } \mathcal{H}_{01} \end{cases}$$
(2)

Where hypothesis \mathcal{H}_{10} means PU is active for first *u* samples and then becomes inactive for rest of the samples n = u + 1 to *N*. and hypothesis \mathcal{H}_{01} means the PU was inactive for the first *v* samples and then became active for rest of the samples n = v + 1 to *N*. *N* is total number of samples for entire sensing period. Here we consider PU signal *s*(*n*) and noise *w*(*n*) as circularly symmetric complex Gaussian (CSCG) which represents signals with rich-ISI such as OFDM. For large number of samples *N* we can invoke central limit theorem (CLT) under hypothesis \mathcal{H}_{10} and the PDF of the test statistics T_{NY} , $p_{10}(x)$ can be approximated by Gaussian distribution. Thus the P_D and P_{FA} of NTS-CR respectively can be given mathematically as in Eqs. 3 and 4 respectively.

$$P_{FA}(\xi, N, u, k) = Pr \cdot \left(T_N(y) > \xi \mid \mathcal{H}_{10}\right) = \int_{\xi}^{\infty} p_{10}(x)dx$$

$$= Q\left(\frac{\frac{\xi}{\sigma_w^2} - \frac{u}{N}k\gamma_{ps} - 1}{\sqrt{\frac{u}{N^2}(k\gamma_{ps} + 1)^2 + \frac{N-u}{N^2}}}\right)$$

$$P_D(\xi, N, v, k) = Pr \cdot \left(T_N(y) > \xi \mid \mathcal{H}_{01}\right) = \int_{\xi}^{\infty} p_{01}(x)dx$$

$$= Q\left(\frac{\frac{\xi}{\sigma_w^2} - \frac{N-v}{N}k\gamma_{ps} - 1}{\sqrt{\frac{N-v}{N^2}(k\gamma_{ps} + 1)^2 + \frac{v}{N^2}}}\right)$$
(3)

where ξ is the threshold of the energy detector employed at the receiver. Now we can set u = 0, v = 0 and k = 1 in above equations and we can obtain $P_{FA}(\xi, N, u, k)$ and $P_D(\xi, N, v, k)$ for TS-CR as follows:

$$P_{FA}(\xi, N, u, k) = Q\left(\left(\frac{\xi}{\sigma_w^2} - 1\right)\sqrt{N}\right)$$
(5)

$$P_D(\xi, N, \nu, k) = \left(\left(\frac{\xi}{\sigma_w^2} - \gamma_{ps} - 1 \right) \frac{\sqrt{N}}{\gamma_{ps} + 1} \right)$$
(6)

From above two equations we can see that the P_{FA} and P_D for TS-CR are special case of NTS-CR. For a given \bar{P}_D , P_{FA} relates to P_D and for a given \bar{P}_{FA} , P_D relates to P_{FA} respectively as follows [9]:

$$P_{FA} = Q\left(\sqrt{2\gamma + 1}Q^{-1}(\bar{P}_D) + \sqrt{\tau f_s}\gamma\right)$$
(7)

$$P_D = Q\left(\frac{1}{\sqrt{2\gamma + 1}}\left(Q^{-1}(\bar{P}_{FA}) - \sqrt{\tau f_s}\gamma\right)\right)$$
(8)

Similarly, for given $(\bar{P}_D, \bar{P}_{FA})$, the minimum number of samples are given by:

$$N_{min} = \frac{1}{2} \left[Q^{-1}(\bar{P}_{FA}) - Q^{-1}(\bar{P}_D)\sqrt{2\gamma + 1} \right]^2$$
(9)

5 HDC-SS and FDC-SS Sensing-Throughput Comparison

HDC-SS involves MAC frame with duration T which consists of sensing time τ and data transmission time $(T - \tau)$. Whereas FDC-SS involves sensing and data transmission time equal to entire frame duration T with collocated antennas used for spectrum sensing and data transmission respectively. There are two scenarios for which SU can operate in PU's licensed band:

Scenario I When PU is not present and no false alarm is generated by SU Scenario II When PU is present but it's not detected by SU.

In both scenarios the achievable throughputs of the SU link are $\left(\frac{T-\tau}{T}\right)\log_2\left(1+P_s/N_0\right)$ and $\left(\frac{T-\tau}{T}\right)\log_2\left(1+P_s/P_o+N_0\right)$ respectively. Where P_s = received power of SU, P_p = PU power as an interference to SU and N_0 is the noise power. The probability of occurrence of scenario I and scenario II are $\left(1-P_{FA}(\xi,N,u,k)\right)P(\mathcal{H}_0)$ and $\left(1-P_D(\xi,N,v,k)\right)P(\mathcal{H}_1)$. Let us say if:

$$R_{0}(\xi, N, u, k) = \left(\frac{T - \tau}{T}\right) \log_{2} \left(1 + SNR_{s}\right) \times \left(1 - P_{FA}(\xi, N, u, k)\right) P(\mathcal{H}_{0})$$

$$(10)$$

$$R_{1}(\xi, N, v, k) = \left(\frac{T - \tau}{T}\right) \log_{2} \left(1 + \frac{SNR_{s}}{SNR_{p}}\right) \times \left(1 - P_{D}(\xi, N, v, k)\right) P(\mathcal{H}_{1})$$
(11)

Then the average throughput of the SU is:

$$R = R_0(\xi, N, u, k) + R_1(\xi, N, v, k)$$
(12)

Now for HDC-SS longer the sensing time τ , shorter is the data transmission time $T - \tau$ which results in less throughput for SU. Similarly for FDC-SS let us assume that there is no self-interference and SU operates over entire frame duration *T*. Following the same steps we can derive average throughout for NTS-CR with HDC-SS. Now average throughput of the TS-CR and NTS-CR with HDC-SS and FDC-SS respectively is as follows:

$$R_0(\xi, N, u, k) + R_1(\xi, N, v, k) < R_0^F(\xi, N, u, k) + R_1^F(\xi, N, v, k)$$
(13)

$$\left(\frac{T-\tau}{T}\right) \left\{ \log_2 \left(1 + SNR_s\right) \left(1 - P_{FA}(\xi, N, u, k)\right) P(\mathcal{H}_0) + \log_2 \left(1 + \frac{SNR_s}{1 + SNR_p}\right) \left(1 - P_D(\xi, N, v, k)\right) P(\mathcal{H}_1) \right\}$$

$$< \log_2 \left(1 + SNR_s\right) \left(1 - P_{FA}(\xi, N, u, k)\right) P(\mathcal{H}_0) + \log_2 \left(1 + \frac{SNR_s}{1 + SNR_p}\right) \left(1 - P_D(\xi, N, v, k)\right) P(\mathcal{H}_1)$$
(14)

From Eq. (13) it is clear that the average throughput of NTS-CR employing FDC-SS is greater than the throughput of the TS-CR employing HDC-SS.

6 Numerical Results and Discussion

Figure 3 shows the probability of detection versus energy-detection threshold for the for NTS-CR without self-interference i.e. k = 1. We can see that the probability of detection decreases as the threshold increases. It also decreases with the increase in number of samples for which PU is initially absent i.e from v = 500 to v = 1500. Figure 4 shows the probability of missed detection versus threshold for NTS-CR with no self-interference. We can see the probability of missed detection increases with the increase in threshold ϵ . It also increases with the increase in number of samples assigned to PU for which initially remains inactive.

Figure 5 shows the probability of detection of TS-CR and NTS-CR with HDC-SS and FDC-SS spectrum sensing scheme. Here we consider the effect of self-interference and compare the probability of detection with self-interference and without self-interference. We can see the P_D for TS-CR outperforms all the cases of NTS-CR because the PU exists for some samples. Moreover, the P_D decreases as the energy-detector threshold increases, it also decreases for NTS-CR with the self-interference and without self-interference. However, the P_D without self-interference is higher than the P_D with self-interference, but the performance gap is very small. Similarly, Fig. 6 shows the probability of missed detection which increases with the increase in energy-detection threshold. It also increases with the increase in number of initial sample values for which PU remains inactive. The probability of missed detection also increases with the increase in self-interference k. There is small performance gap between probability of missed detection with self-interference and without self-interference, but there is significant difference in probability of missed detection for TS-CR and the probability of missed detection when the number of initial samples for which PU is inactive increase to v = 1500. For example at energy-detection threshold of $\xi = 1$ the probability of missed detection is 0.2 which increases to P_{MD} value



Fig. 3 Probability of detection versus threshold of TS-CR with HDC-SS and NTS-CR with FDC-SS without self-interference



Fig. 4 Probability of missed detection versus threshold of TC-CR with HDC-SS and NTS-CR with FDC-SS without self-interference



Fig. 5 Probability of detection versus threshold of TS-CR with HDC-SS and NTS-CR with FDC-CR with and without interference

of 0.55 for PU which remains inactive for initial sample values of v = 1500. We summarise different values of P_D , P_{MD} and P_{FA} in the following Table 1. Note that we calculate the P_{FA} when he PU is active initially for *u* samples and then leave the channel for rest of the sample values (Tables 1, 2).



Fig. 6 Probability of missed detection versus threshold of TS-CR with HDC-SS and NTS-CR with FDC-SS with and without interference

Table 1 P_D , P_{MD} and P_{FA} values for non-time slotted CRN with self-interference and without self-interference

Threshold (ξ)	# of samples (v)	Self- interference (k)	P_D	P_{MD}	и	P_{FA}
1	500	1	0.68	0.32	500	0.70
1	500	0.9	0.65	0.35	500	0.68
1	1500	1	0.28	0.72	1500	0.92
1	1500	0.9	0.25	0.75	1500	0.90

Table 2 Throughput of TS-CR with HDC-SS and NTS-CR with FDC-SS at SNR values of -2 dB

Sensing time (τ) (ms)	Throughput (<i>b</i> / <i>s</i> / <i>Hz</i>):TS-CR with HDC-SS	Throughput $(b / s / Hz)$: NTS-CR with FDC-SS	Percent increase in throughput (%)	
6	3.6367	3.8688	5.9	
10	3.5899	3.9888	10	
12	3.5145	3.9937	11.9	
16	3.3557	3.9949	16	

Figures 7 and 8 on probability of false alarm follows our intuition developed after our earlier discussions on P_D and P_{MD} with and without self-interference for TS-CR and NTS-CR. We can see the probability of false alarm of TS-CR is less when the PU is initially active for u = 500 to u = 1500 samples and later subsides and leaves the channel empty. This is also true for the P_{FA} values with self-interference and without self-interference. For



Fig. 7 Probability of false alarm versus threshold of TS-CR with HDC-SS and NTS-CR with FDC-SS without self-interference



Fig. 8 Probability of false alarm versus threshold of TS-CR with HDC-SS and NTS-CR with FDC-SS with and without interference

fixed sample values let us say u = 500 the P_{FA} with interference is little higher than the P_{FA} without self-interference i.e. k = 0.9, however the performance gap is very small (Fig. 9). We can expect more degradation in P_{FA} when the self-interference factor is



Fig. 9 Throughput versus sensing time of TS-CR with HDC-SS and NTS-CR with FDC-SS at SNR values of -2 and -4 dB respectively



Fig. 10 Bar graph of throughput versus sensing time for TS-CR with HDC-SS and NTS-CR with FDC-SS at SNR values of -2 and -4 dB respectively

reduced from k = 0.9 to k = 0.1 where sever interference can be expected. Figure 10 shows the statistical attestation of the throughput of NTS-CR which uses FDC-SS is greater than the throughput of TS-CR at different vales of SNRs.

7 Conclusion

Cognitive Radio revolves around two basic and important QoS indicators which is increased throughput yield for SU and maximum protection to PU. Half-duplex communication cognitive radio which first senses and then according to sensing statistics performs data transmission. This shall reduce data transmission time for half-duplex communication cognitive radio at low SNR regimes which means throughput compromise for SU. We have seen that Full-duplex communication cognitive radio with half-duplex spectrum sensing outperforms half-duplex with half-duplex spectrum sensing schemes. We have calculated the P_D , P_{FA} , P_D and throughput of the TS-CR and the NTS-CR with half and full duplex spectrum sensing. It has been proved that full-duplex cognitive radio yields more throughout than the half-duplex spectrum. We obtain approximately 5.9% increase for NTS-CR with FDC-SS at sensing time of 6 ms. Similarly, 10% of increase in throughput at sensing time of 10 ms has been observed for NTS-CR with FDC-SS for SNR values of -2 dB.

Open Access This article is distributed under the terms of the Creative Commons Attribution 4.0 International License (http://creativecommons.org/licenses/by/4.0/), which permits unrestricted use, distribution, and reproduction in any medium, provided you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons license, and indicate if changes were made.

References

- 1. Digham, F. F., Alouini, M.-S., & Simon, M. K. (2007). On the energy detection of unknown signals over fading channels. *IEEE Transactions on Communications*, 55(1), 21–24.
- Zhang, W., Mallik, R. K., & Letaief, K. (2009). Optimization of cooperative spectrum sensing with energy detection in cognitive radio networks. *IEEE Transactions on Wireless Communications*, 8(12), 5761–5766.
- Fuhrmann, D. R., Kelly, E. J., & Nitzberg, R. (1992). A CFAR adaptive matched filter detector. *IEEE Transactions on Aerospace and Electronic Systems*, 28(1), 208–216.
- Kapoor, S., Rao, S., & Singh, G. (2011). Opportunistic spectrum sensing by employing matched filter in cognitive radio network. In 2011 International conference on communication systems and network technologies (CSNT) (pp. 580–583). IEEE.
- Zeng, Y., & Liang, Y.-C. (2007). Maximum-minimum Eigenvalue detection for cognitive radio. In IEEE 18th international symposium on personal, indoor and mobile radio communications, 2007. PIMRC 2007 (pp. 1–5). IEEE.
- 6. Zeng, Y., & Liang, Y.-C. (2009). Eigenvalue-based spectrum sensing algorithms for cognitive radio. *IEEE Transactions on Communications*, 57(6), 1784–1793.
- 7. Xu, S., Zhao, Z., & Shang, J. (2008). Spectrum sensing based on cyclostationarity. In *Workshop on power electronics and intelligent transportation system*, 2008. *PEITS'08* (pp. 171–174). IEEE.
- Tian, Z., & Giannakis, G. B. (2006). A wavelet approach to wideband spectrum sensing for cognitive radios. In 1st international conference on cognitive radio oriented wireless networks and communications, 2006 (pp. 1–5). IEEE.
- Liang, Y.-C., Zeng, Y., Peh, E. C., & Hoang, A. T. (2008). Sensing-throughput tradeoff for cognitive radio networks. *IEEE Transactions on Wireless Communications*, 7(4), 1326–1337.
- Cheng, W., Zhang, X., & Zhang, H. (2015). Full-duplex spectrum-sensing and MAC-protocol for multichannel nontime-slotted cognitive radio networks. *IEEE Journal on Selected Areas in Commu*nications, 33(5), 820–831.
- Sharkasi, Y. F., Ghogho, M., & McLernon, D. (2012). Sensing-throughput tradeoff for OFDM-based cognitive radio under outage constraints. In 2012 International symposium on wireless communication systems (ISWCS) (pp. 66–70). IEEE.
- 12. Derakhshani, M., & Le-Ngoc, T. (2012). Aggregate interference and capacity-outage analysis in a cognitive radio network. *IEEE Transactions on Vehicular Technology*, 61(1), 196–207.

- Chaudhari, S., Koivunen, V., & Poor, H. V. (2009). Autocorrelation-based decentralized sequential detection of OFDM signals in cognitive radios. *IEEE Transactions on Signal Processing*, 57(7), 2690–2700.
- Taherpour, A., Nasiri-Kenari, M., & Gazor, S. (2010). Multiple antenna spectrum sensing in cognitive radios. *IEEE Transactions on Wireless Communications*, 9(2), 814–823.
- Ghassemi, A., Lampe, L., Attar, A., & Gulliver, T. A. (2010). Joint sidelobe and peak power reduction in OFDM-based cognitive radio. In 2010 IEEE 72nd vehicular technology conference fall (VTC 2010-Fall) (pp. 1–5). IEEE.
- Ghassemi, A., & Gulliver, T. A. (2012). PAPR reduction in OFDM based cognitive radio with blockwise-subcarrier activation. In 2012 IEEE international conference on communications (ICC) (pp. 5598–5602). IEEE.
- Ni, C., Jiang, T., & Peng, W. (2015). Joint PAPR reduction and sidelobe suppression using signal cancelation in NC-OFDM-based cognitive radio systems. *IEEE Transactions on Vehicular Technology*, 64(3), 964–972.
- Atapattu, S., Tellambura, C., & Jiang, H. (2011). Energy detection based cooperative spectrum sensing in cognitive radio networks. *IEEE Transactions on Wireless Communications*, 10(4), 1232–1241.
- Li, X., Cao, J., Ji, Q., & Hei, Y. (2013). Energy efficient techniques with sensing time optimization in cognitive radio networks. In 2013 IEEE wireless communications and networking conference (WCNC) (pp. 25–28). IEEE.
- Sun, H., Nallanathan, A., Cui, S., & Wang, C.-X. (2015). Cooperative wideband spectrum sensing over fading channels. *IEEE Transactions on Vehicular Technology*, 65(3), 1382–1394.
- Chen, Y. (2010). Improved energy detector for random signals in Gaussian noise. *IEEE Transactions on Wireless Communications*, 9(2), 558–563.
- Chatterjee, S., Maity, S. P., & Acharya, T. (2014). Energy efficient cognitive radio system for joint spectrum sensing and data transmission. *IEEE Journal on Emerging and Selected Topics in Circuits and* Systems, 4(3), 292–300.



Mir Muhammad Lodro has received his B.E in Telecommunication Engineering, and M.E. in Communication Systems and Networks from Mehran UET Jamshoro in 2008 and 2014 respectively. He has also earned an MRes in Electrical and Electronic Engineering from George Green Institute for Electromagnetics Research-GGIEMR, The University of Nottingham, UK in year 2015. His research was focused on SDR and Statistical Signal Processing for Cognitive Radio in TV White Space. Additionally, he has 6 years of teaching experience at Department of Electrical Engineering, Sukkur Institute of Business Administration, Sindh, Pakistan. His areas of research interests are in broader areas of Communications, Signal Processing and Networks. He is member of IEEE USA and an affiliate member of IEEE Signal Processing, Communication and Networks. He is also reviewer of few international conferences and journals in the areas of CommSigNets and Antennas. He was awarded "Masters Developing Solution" scholarship and "Faculty of Engineering Research Excellence for

PhD" scholarship to pursue his Master of Research (M.Res.) and Ph.D. in Electrical and Electronic Engineering from George Green Institute for Electromagnetics Research-GGIEMR, The University of Nottingham, UK.



Steeve Greedy was born in Cardiff, UK in 1966. He received the MEng and Ph.D. degrees in electrical and electronic engineering from the University of Nottingham, UK, in 1998 and 2002 respectively. In 2003 he was appointed as a Senior Experimental Officer within the George Green Institute for Electromagnetics Research at the University of Nottingham, UK. His areas of research interests are computational electromagnetics and communications. He has recently focused on Cognitive Radio, SDR-based Short Range Communication Systems.



Naeem A. Mahoto received his M.E. from MUET, Pakistan. He was awarded Ph.D. scholarship under UESTPs Project for higher studies and received his Ph.D. from Politecnico Di Torino, Italy. He is currently working an Assistant Professor at MUET Pakistan. Naeem is a co-author of several research articles published in national as well as international journals. He works in the field of data mining and bioinformatics and his research interests are focused on pattern extraction, classification of electronic records in the medical domain and software engineering.



Mukhtiar A. Unar is Meritorious Professor (BPS-22) in the Department of Computer Systems Engineering and the Director of the Institute of Information & Communication Technologies (IICT). He received the B.E in Electronic Engineering from Mehran University of Engineering and Technology, Jamshoro in 1986, the Postgraduate Diploma in Communication Engineering in 1989 from Mehran University, and the MSc and Ph.D. degrees from University of Glassgow, Scotland, UK. Dr. Mukhtiar Ali Unar is a member of IEEE (USA), an affiliate of International Federation of Automatic Control, a member of Pakistan Institute of Engineers and a member of Pakistan Engineering Council.