# Spectrum Sensing for Cognitive Radio Users using Constant Threshold Range in Energy Detector

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#### Abstract

Spectrum Sensing has an important role in Cognitive Radio Network (CRN). Different methodologies have been proposed in the literature for spectrum sensing. Energy detector method is more practical and less in complexity. This paper presents analysis of energy detection algorithm in different scenarios with different parameters to detect presence or absence of Primary User (PU). The energy detection threshold has been used as Constant Threshold Range (CTR) to reduce the trade-off between Probability of False Alarm Pfa and Probability of miss detection P<sub>md</sub>. The Receiver Operating Curves (ROC) has been presented by simulating Spectrum Sensing scenarios under fading and non-fading environments. It is also shown that theses curves can be further improved by Cooperative Spectrum Sensing (CSS) technique. The hardware implementation has also been presented by receiving real time signals using Universal Software Radio Peripheral (USRP) and GNU Radio, which has verified the simulated results.

**Keywords:** Cognitive Radio, Spectrum Sensing, Energy Detection, Constant False Alarm Rate, Constant Threshold Range.

# 1. Introduction

In wireless communication, CR has gained significant importance for spectrum management to avoid the inescapable spectrum scarcity. Cognitive Radio is a smart radio which acquires knowledge about available spectrum through process called Spectrum Sensing. Any licensed user known as Primary User (PU) do not utilize the spectrum all the time. Thus policy of Opportunistic Spectrum Access (OSA) has emerged in CR, so that any secondary user known as Cognitive Radio User (CRU) can access the spare band opportunistically without causing any interference with PU [1]. Ideally a CRU should not cause any interference with PU but wireless channels are always random in nature due to which a CRU at times make wrong decision about absence or presence of PU [2]. A reliable wireless

communication can be described in terms of ROC which includes Probability of detection  $P_d$ , Probability of false alarm  $P_{fa}$  and Probability of miss detection  $P_{md}$  [3]. The optimization of these stochastic parameters is the work being targeted.

Transmitter spectrum sensing techniques can be categorized as coherent detection and non-coherent detection. The widely used methods of coherent detection are Matched Filter detection [4] and Cyclostationary feature detection [5] whereas Energy Detection [6] is a non-coherent detection technique. The basic key points for an appropriate selection of spectrum sensing techniques are speed of estimation, accuracy and complexity. Matched Filter and Cyclostationary detection are more accurate methods than energy detection. But these methods require priori knowledge about the signal so are complex too. The least complex and more practical technique for spectrum sensing is Energy Detection which works as a blind detector [7].

The credibility of energy detector depends on signal to noise ratio (SNR  $\delta)$  and the value of threshold ( $\tau$ ) which is used to make decision about the presence or absence of PU. The combination of these techniques with different threshold values has been implemented in the literature [8]. The Constant False Alarm Rate (CFAR) has been used by the researchers over the ancient past to derive  $\tau$  which requires trade-off between  $P_{\rm fa}$  and  $P_{\rm md}$ . In order to reduce this trade-off and probability of error [9], different threshold values are used in this paper.

However spectrum sensing result of a single CRU is not reliable due to shadowing and fading environments. So by cooperation among different CRU we can get even better results. In CSS a CRU makes its local decision and sends it to the CR Controller (CRC) known as Fusion Center (FC) where data from every CRU is aggregated for final and accurate decision. But this cooperation brought the additional communication overhead by increasing the  $P_d$  [10].

A comparative analysis has been carried out in this paper using simulations in terms of  $P_{\rm fa}$ ,  $P_{\rm d}$  and  $P_{\rm md}$  by changing different parameters of energy detector under both AWGN and Rayleigh fading wireless communication models. We have taken results by considering CFAR and CTR for comparison using simulations. The improved results by CSS have also been



modulated in this paper using simulations. To verify the simulation results a data set has been prepared and demonstrated in a Hardware environment using USRP and the comparison results have been analyzed.

This paper is organized in different sections. Our system model with an energy detector is explained in section II. The description of CSS is given in section III. Then section IV will show the simulations and analysis. The Hardware implementation using USRP has been demonstrated in section V, which also contains analysis of its comparison with the simulations. The conclusion and future work of paper is in section VI.

# 2. System Model

The system model consists of a single CRU, sensing the presence of PU with hypothesis  $\mathcal{H}0$  and  $\mathcal{H}1$  defined as,

$$x(t) = n(t)$$
  $\mathcal{H}0$  defining signal is absent (1)

$$x(t) = s(t) + n(t)$$
  $\mathcal{H}_1$  defining signal is present (2)

Where s(t) source signal, x(t) is received signal, and n(t) is an AWGN. The energy of the signal is compared with threshold  $\tau$  values as,

$$\gamma < \tau$$
 Signal is absent (3)

$$\gamma > \tau$$
 Signal is present (4)

The approximated formulae of  $P_d$ ,  $P_{fa}$  and  $P_{md}$  for energy detector can be given by [11].

$$P_{fa} = \Gamma(N, \tau/2) / \Gamma(\mathbf{u}) \tag{5}$$

$$P_{d} = Q(\sqrt{2\delta}, \sqrt{2\tau}) \tag{6}$$

$$P_{md} = 1 - P_d \tag{7}$$

Where the  $\delta$  represent the SNR, time bandwidth product is given by u,  $\Gamma$  (.,.) and  $\Gamma(.)$  are incomplete and complete gamma functions respectively. When we set a  $\tau$  according to the CFAR, a trade-off between  $P_{md}$  and  $P_{fa}$  is required [12] which is shown by Fig. 1. The targeted task of our system model is to reduce this trade-off, thus we set the  $\tau$  range from  $\tau(1)$  to  $\tau(N)$  as Constant Threshold Range (CTR) which is shown in Fig. 2, to reduce both  $P_{md}$  and  $P_{fa}$ .

#### Probability

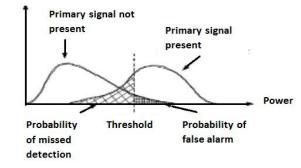


Fig. 1 Tradeoff between  $P_{\text{fa}}$  and  $P_{\text{md}}$  with CFAR

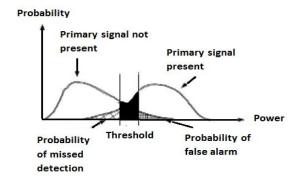


Fig. 2 Tradeoff between Pfa and Pmd with CTR

$$P_{d} = e^{-(\tau/2)} \sum_{k=0}^{u-2} (1/k!) (\tau/2)^{k} + \{(1+\delta_{a})/\delta_{a}\}^{u-1} \{e^{-(\tau/2(1+\delta_{a}))} - e^{-(\tau/2)} \sum_{k=0}^{u-2} (1/k!) (\tau\delta_{a}/2(1+\delta_{a}))\}$$

Although an Energy Detector does not detect a signal below certain limit of  $\delta$  which is known as  $\delta$  wall, yet it is more practical, as energy of signal can easily be detected. In any communication system model we make an assumption that the noise is Additive White Gaussian. However, practically different kinds of noises added in the signal during its path [13]. In this paper we have taken results under more practical fading channel i.e, Rayleigh Fading.

# 3. Cooperative Spectrum Sensing

The first objective of CRU is to detect the presence of PU in a particular spectrum band, to be used by CRU, in case found vacant. Any CRU, in isolation does not sense the presence of PU. Therefore CSS is adopted as a reliable and efficient method of spectrum sensing. Cooperative Spectrum Sensing has two stages; sensing and reporting. A CRU makes its local decision in sensing stage and forward that to the FC where final decision has been made using different rules [14].



We have considered the Cooperative Spectrum Sensing to guarantee the higher probability of detection. The performance analysis of Spectrum sensing can be made by Receiver Operating Curves (ROC). We use AND rule for making final decision at FU. Now the probabilities, detection and false alarm can be given by [15].

$$P_d = \prod_{i=1}^{k_i} (P_d) \tag{9}$$

$$P_{d} = \prod_{1}^{k_{i}} (P_{d})$$

$$P_{fa} = \prod_{1}^{k_{i}} (P_{fa})$$
(10)

In CSS all the CRUs will send their local decisions, '1' representing presence of PU and '0' representing absence of PU, then final decision will be taken by FC using AND rule. In AND rule, if all CRUs give their local decision as '1' then final decision will be '1'. This whole process is shown in Fig. 3.

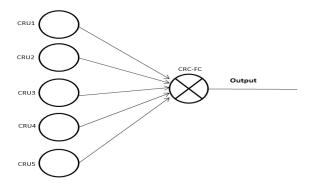


Fig. 3 Model of Cooperative Spectrum Sensing

The final decision will come back to every CRU about presence or absence of PU. Clearly, this will overcome any wrong decision made by any of CRU due to environmental conditions but causes additional overhead.

## 4. Simulation and Analysis

The detection performance of single CRU and CSS will be evaluated in this section using ROC. The most important parameter contributing in accurate results is SNR ( $\delta$ ). The Fig. 4 shows that detection performance of energy detector is improved by increasing the  $\delta$  under both AWGN and Rayleigh fading environments with CFAR. The resultant decrease in P<sub>md</sub> is shown in Fig. 5.

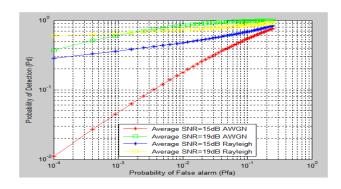


Fig. 4 P<sub>d</sub> with different SNR using CFAR

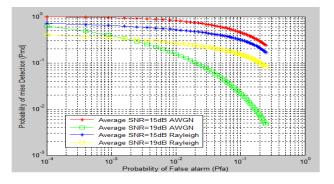
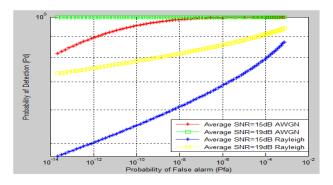


Fig. 5 Pmd with different SNR using CFAR

It is evident from the result shown in Fig. 4 and Fig. 5 that decrease in P<sub>md</sub> with higher δ, P<sub>fa</sub> increases which warrants the trade-off between  $P_{\text{fa}}$  and  $P_{\text{md}}$ . Infect, increase in  $P_{\text{md}}$  means CRU is causing interference for a PU and increasing Pfa means a CRU is not utilizing the free spectrum efficiently. However, it is required to avoid interference for a CRN. We prefer low P<sub>md</sub> over low P<sub>fa</sub>. P<sub>fa</sub> approaches to 1 even at very low P<sub>md</sub>. So to improve the Pfa if we avoid CFAR and use CTR in (5), better results of Pfa are expected.

An ideal sensing case i.e, Pd become 1 in AWGN channel at 19dB ( $\delta$ ) is shown in Fig. 6 and P<sub>fa</sub> decreases by taking CTR with an acceptable increase in P<sub>md</sub> too as shown in Fig. 7. Actually, we have to set our  $\tau$  such as to get both  $P_{md}$  and  $P_{fa}$  as low as possible by the applied algorithm





With Cooperation AWGN

No Cooperation AWGN

With Cooperation AWGN

With Cooperation AWGN

No Cooperation Rayleigh

10-7

10-4

10-7

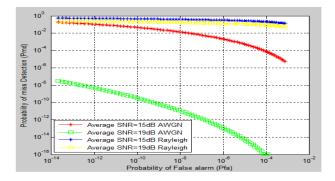
10-4

10-3

Probability of False alarm (Pfa)

Fig. 6 P<sub>d</sub> with different SNR using CTR

Fig. 9 P<sub>md</sub> with Cooperative Sensing



CTR instead of using CFAR provides  $P_{\text{fa}}$  as low as  $10^{-2}$  as compared to almost 1 for CFAR.

From the simulations we can see that the proposed technique of

Fig. 7  $P_{md}$  with different SNR using CTR

# 5. Experimental Setup

As explained earlier, with cooperation among nodes we improve the detection performance which is shown in Fig. 8 and Fig. 9. Here also the  $P_d$  is nearly equal to 1 by applying CSS in AWGN channel.

To verify and validate the simulation results, an experimental setup was arranged. The experimental setup consists of a signal generator, USRP and a Personal Computer (PC). The daughter board of USRP is XCVR2450 which covers the dual band frequency ranges from 2.3GHz to 2.4GHz and 4.9 to 5.9GHz. The whole work is done on 2.4GHz frequency band. A SME06 Signal Generator is connected with an antenna via RF cable for signal transmission and USRP connected with PC using Ethernet cable is our receiver. A GNU Radio software toolkit is used for signal processing of received signal on PC. This is shown in Fig. 10.

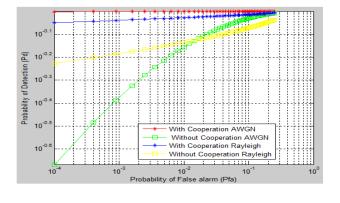




Fig. 8 P<sub>d</sub> with Cooperative Sensing

Fig. 10 Experimental test bed



A sine wave with PSK modulation is transmitted by the signal generator using an antenna to USRP, which is seen by GNU Radio on PC. We have taken a USRP source block on GNU Radio at 2.4GHz frequency with 32k sampling rate and makes a FFT plot of received signal with GUI FFT Sink from GNU Radio signal processing blocks. Later the real time input signals and noise are evaluated using Matlab. Similar results are obtained by real time data using USRP.

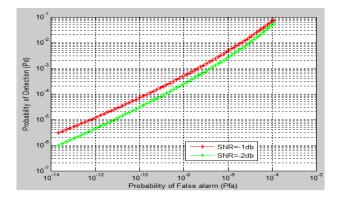


Fig. 11 Pd with different SNR using USRP

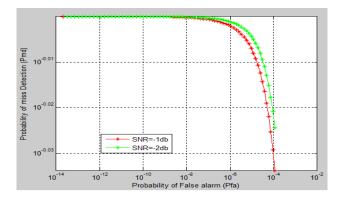


Fig. 11 P<sub>md</sub> with different SNR using USRP

The  $\delta$  of received signal is improved by increasing the power level of input signal from transmitter, by which the detection performance increases [16] under AWGN and Rayleigh channel as shown in Fig. 11 which effects decrease in  $P_{md}$  in Fig. 12.

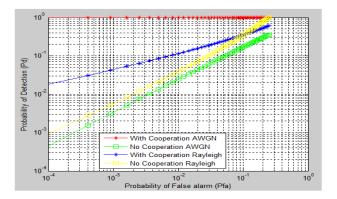


Fig. 13 Pd with CSS using USRP

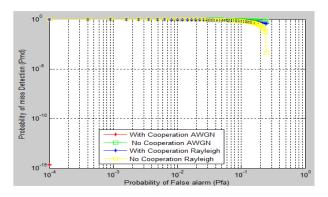


Fig. 14  $P_{md}$  with CSS using USRP

By applying the CSS on real time date, the  $P_d$  increases and becomes ideal in AWGN channel as shown in Fig. 13. The  $P_{md}$  decreases too, which is shown in Fig. 14.

## 6. Conclusions and Future Work

The performance of CRN has been evaluated for Spectrum Sensing using energy detector in different environments. It is shown that if  $\delta$  of a signal is improved then detection performance also enhances but at the same time  $P_{\rm fa}$  increases. Then that  $P_{\rm fa}$  can be reduced in an ideal environment with filtering and no temperature change at receiver end with CTR. However, practically it is difficult and requires a long sensing time. Thus, CSS has been implemented which provide good results as shown by ROC in our paper. All this comparative analysis has shown that trade-off between  $P_{md}$  and  $P_{\rm fa}$  is made according to a required practical environment but we can get a reasonable decrease in  $P_{\rm fa}$  with an acceptable increase in  $P_{md}$  by avoiding the CFAR method.

In future the researchers may find a way to estimate the noise variance with low estimation error to validate the results. The



combination of different techniques may also be used for Spectrum Sensing in parallel with CTR technique and then use both the results in Spectrum Allocation process to specific users. We want to get an accurate and more reliable result with low sensing time in order to avoid interference with PU and efficiently utilize the available spectrum.

## References

- Ejaz.W, Hasan.N, Azam.A.M and Kim.S.H, "Improved local spectrum sensing for cognitive radio networks", EURASIP Journal on Advances in Signal Processing, 2012.
- [2] Bhatnagar.A.S.M.R and Mallik.R.K, "Cooperative Spectrum Sensing with an Improved Energy Detector with Cognitive Radio Network", IEEE National Conference on Communications, 2011.
- [3] Peh.E and Liang.Y.C, "Optimization for cooperative sensing in cognitive radio network", IEEE Wireless Communications and Networking Conference WCNC, 2007.
- [4] D. Cabric, S. M. Mishra, and R. W. Brodersen, "Implementation Issues in Spectrum Sensing for Cognitive Radios", 38th Asilomar Conference on Signals, Systems and Computers, 2004.
- [5] Chang.S.Y, "Analysis of Proposed Sensing Schemes", Conference IEEE Wireless RANs, 2006.
- [6] D. Cabric, A. Tkachenko and R.W. Brodersen, "Spectrum Sensing measurements of pilot, energy, and collaborative detection", IEEE MILCOM, 2007.
- [7] D. Cabric, A. Tkachenko, R.W. Brodersen, "Experimental Study of Spectrum Sensing based on Energy Detection and Network Cooperation", TAPAS'06 First International Workshop on Technology and Policy for Accessing Spectrum, 2006.
- [8] Ejaz.W, Hasan.N, Azam.A.M, Lee.S and Kim.S.H, "I3S: Intelligent spectrum sensing scheme for cognitive radio networks", EURASIP Journal on Wireless Communications and Networking, 2013.
- [9] Sadhukhan.P, Naveen Kumar.N and M.R. Bhatnagar, "Improved Energy Detector Based Spectrum Sensing for Cognitive Radio: An Experimental Study", Annual IEEE India Conference (INDICON), 2013.
- [10] Zhang.N, Chen.H and Xie.L, "A Communication-overhead-aware Cooperative Spectrum Sensing Scheme in Cognitive Radio Networks", International Conference on Wireless Communication and Signal Precessing, 2012.
- [11] S. Hussain X. Fernando, "Approach for cluster-based spectrum sensing over band-limited reporting channels", *IET Communications Journal*, 2011.
- [12] Nautiyal.G. and Kumar.R, "Spectrum Sensing in Cognitive Radio using Matlab", International Journal of Engineering and Advanced Technology (IJEAT), 2013. -Vol. 2.
- [13] Tandra.R.S and Anant, "SNR Walls for Signal Detection", IEEE Journal of Selected Topics in Signal Processing, 2008.Vol. 2.
- [14] Xie.Y.Z.X and Yang.L, "Cooperative Spectrum Sensing Based on SNR Comparison in Fusion Center for Cognitive Radio", International Conference on Advanced Computer Control IEEE, 2008.

- [15] Zeng.Y, Liang.Y.C, Hoang.A.T and Zhang.R, "A Review on Spectrum Sensing for Cognitive Radio: Challenges and Solutions", EURASIP Journal on Advances in Signal Processing, 2009. - Vol. 2010.
- [16] Uyanik.G.S, Cepheli.O.Z, Kurt.G.K and Oktug.S, "Implementation and Performance Evaluation of Dynamic Spectrum Access Using Software Defined Radios", First International Black Sea Conference on Communications and Networking (BlackSeaCom), 2013.

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