

# Statistic Cyclostationary Feature Detection and Improving Channel Estimation Accuracy in Cognitive Radio

<sup>1</sup>Shefali Chaudhary(Student), <sup>2</sup>Er. Harnek Singh(Assistant Professor), <sup>3</sup>Dr. R Kashyap(HOD)

<sup>1</sup>Rayat Institute of Engineering and Information Technology, Railmajra

<sup>2</sup>Rayat Institute of Engineering and Information Technology, Railmajra

<sup>3</sup> Rayat Institute of Engineering and Information Technology, Railmajra

**Abstract:** In future the basic problem is to find suitable spectrum band to fulfill the demand of services. While observations shows that the radio spectrum is unutilized. Cognitive radio is an important solution to improve the spectrum utilization. Cognitive users (CU) continuously sense the spectrum to utilize the spectrum without disturbing licensed users (LU). There are various spectrum sensing techniques but the ideal one should be fast, accurate and efficient. In this paper cyclostationary feature detection (CFD) technique is discussed and channel estimation accuracy in the systems has been improved. First of all an algorithm is proposed based on CFD and Hilbert Transformation. Algorithm is named as statistic CFD. Comparing with the conventional CFD algorithm, proposed algorithm is more flexible. Second objective is to improve channel estimation accuracy in OFDM system using Least Square (LS) method and minimum mean square error (MMSE) for initial channel estimation and then add LMS iterative algorithm.

**Keywords:** Cognitive radio, cyclostationary feature detection, Hilberts Transformation, Channel Estimation, LS, MMSE and BER.

## I. INTRODUCTION

The concept of cognitive radio was first introduced by Sir Joseph Mitola III at Royal Institute of Technology Stockholm in 1998. The need of the technology is to improve the utilization of radio spectrum. In cognitive radio network, cognitive user (CU) continuously sense the spectrum and came to know that when the primary user is using it or not. And it intelligently use the band when primary user is absent. There are many sensing techniques like

matched filter, energy detection and cyclostationary feature detection. Out of these techniques CFD is fast, accurate and efficient and it works well in low SNR situations. Spectrum sensing and channel estimation is the first step to implement cognitive radio system. Process of spectrum sensing is shown in the figure below:

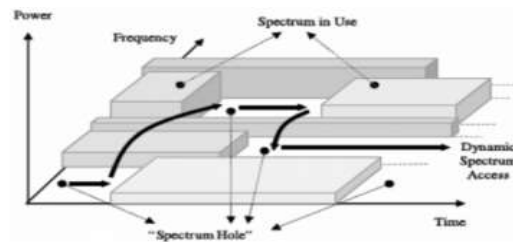


Figure 1: Spectrum hole or White Space

### A. Cyclostationary Feature Detection using Hilbert Transformation

The properties of cyclostationary process vary periodically. Cyclostationary feature detection method deals with the feature of the signal that have a periodic statistics. CFD method is more immune to noise as it exploits this periodicity in the received primary signal to identify PU. In conventional CFD method the parameters which is used for detecting primary signal is cyclic autocorrelation function. But in statistic CFD algorithm we don't need to know about CAF of signal. Factor is  $e^{-j2\pi\alpha t}$  which is related

to cyclic frequency  $\alpha$ . After adding the factor we will calculate its average on a certain cyclic frequency.

Let us assume that  $r(t)$  is received signal.  $r(t)$  may contain both signal  $s(t)$  and  $n(t)$  because of environment. Two hypothesis of  $r(t)$  :

$$\begin{aligned} r(t) &= n(t) & H_0 \\ &= s(t)+n(t) & H_1 \end{aligned}$$

Where  $H_0$  represents absence of LU and  $H_1$  represents presence of LU means signal and noise both are present.

When we add factor  $e^{-j2\pi\alpha t}$  to  $r(t)$  we get

$$x(\alpha, t) = r(t) * e^{-j2\pi\alpha t}$$

Calculate average of  $x(\alpha, t)$  then a defined variable  $M(\alpha, t_0)$  for certain cyclic frequency is given as follow and also length of detecting time is assume to be  $t_0$

$$\begin{aligned} M(\alpha, t_0) &= \mathcal{N}_0 \left| \int_0^{t_0} x(\alpha, t) dt \right| \\ &= \mathcal{N}_0 \left| \int_0^{t_0} r(t) * e^{-j2\pi\alpha t} dt \right| \end{aligned}$$

However, influence caused by noise present in channel can affect the result of detection. So to reduce this influence we use received signal's complex form and we can get complex signal from its real component  $x_r(n)$  by using Hilbert Transformation as shown below:

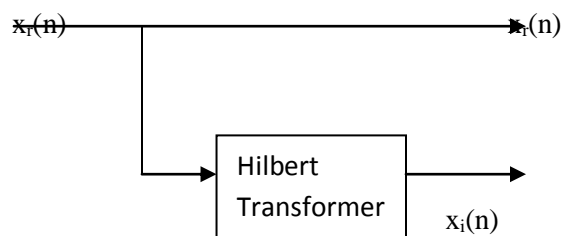


Figure 2: Hilbert Transformer

In this statistic cyclostationary algorithm two parameters will affect the amount of calculations, one is sampling time  $N$  and step size of cyclic frequency

ALPHA. Large amount of calculations takes more time and energy for detection. On calculating average of several sampling results can reduce the influence of noise as AWGN is wide sense stationary.

### B. Channel Estimation

Modulation technique used in cognitive radio is OFDM which divides a high data rate streams over subcarriers. Subcarriers save bandwidth in frequency domain and in time domain they are orthogonal and also cyclic prefix will avoid ISI. So OFDM is a technique that enhance the bandwidth efficiency by removing the frequency selective fading channels. During detection of the signal at the receiver end channel information is required. Pilot aided approach is chosen from the different methods of channel estimation as it is less complex.

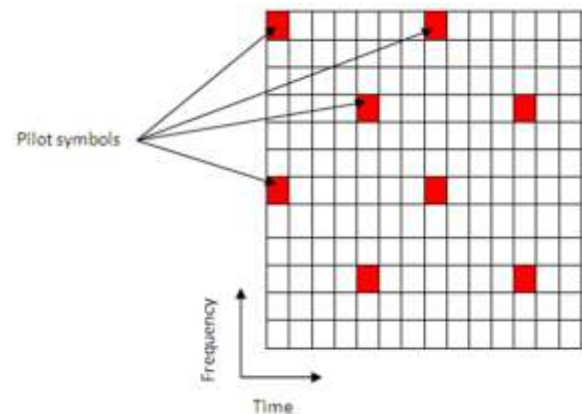


Figure 3: Insertion of Pilot Symbols

At the transmitter end we introduce pilots among subcarriers with equally spaced in frequency domain and at the receiver end they are extracted to estimate channel and interpolation is done in other subcarrier for channel estimation. In this analysis, channel is estimated with Least Square (LS) method and Minimum Mean Square Error (MMSE). Linear interpolation is done at the end to complete the estimation and then LMS iterative algorithm is used to improve the quality of channel estimation. Although channel estimation can be improved by using different iterative algorithm. Output of one iteration can be used as feedback for the next iteration. So in this way result is improved in each iteration. It can be used for SISO as well as for MIMO system. In this study pilots are transmitted along with data and LMS iterative algorithm is used.

In this algorithm channel estimation of last iteration is used. Iteration process is continued until desired result is not obtained.

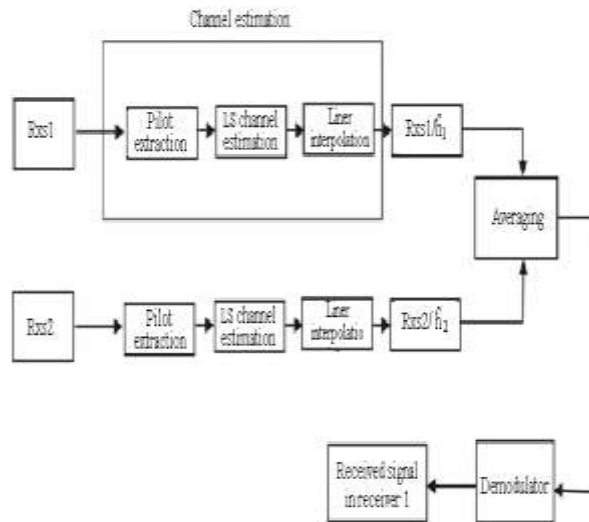


Figure 4 : Proposed Receiver for Channel Estimation

## II. RESULTS & DISCUSSIONS

### A. Simulation of cyclostationary based detection

In first case, statistic cyclostationary detection algorithm is used to detect modulated binary phase shift keying (BPSK) signal. Let us assume that modulated BPSK signal is transmitted in AWGN channel and SNR of the detector is set to be -10db. The parameters used in simulation are sampling time, sampling frequency ( $F_s$ ) and step size of cyclic frequency. The unit step size of cyclic frequency is set to be 1MHz and its sampling time ( $N$ ) is equals to 100. The simulation consequence is normalized to [0,1]. From result of simulation, the possible spectrum of BPSK is shown below :

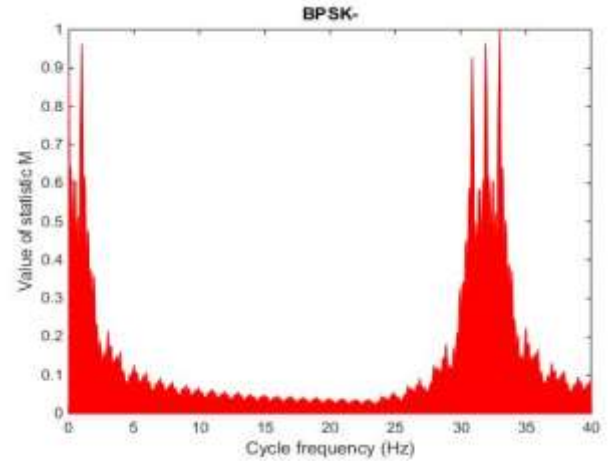


Figure 5: Spectrum of BPSK Signal

Also to form correlation between sensitivity ( $P_d$ ) and specificity ( $P_f$ ) of statistic cyclostationary algorithm, the ROC that is Receiver Operating Characteristics curve is shown in figure below. Also the ROC curve of standard energy detection is also shown and a comparison is made with theoretical values. On comparing we find that CFD algorithm is improved than that of energy detection

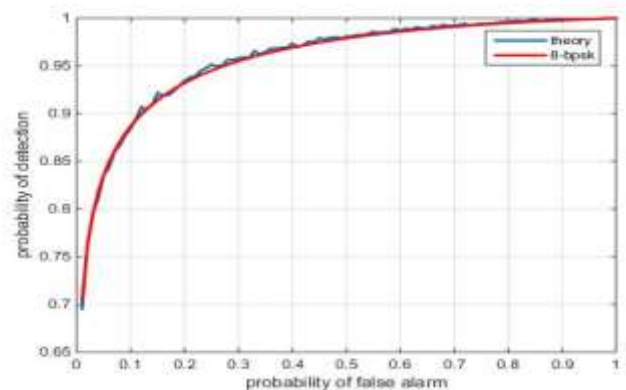


Figure 6: ROC Curve for BPSK signal

In second case detection of QPSK modulated signal is done. SNR is assumed to be -5db and AWGN channel is used to transfer data. Simulation parameters are sampling time, sampling frequency and step size of cyclic frequency. Now the step size of cyclic frequency is set to be 2MHz and sampling time is 100. Simulation outcome is normalized to [0,1] and resulted spectrum of QPSK is shown below:

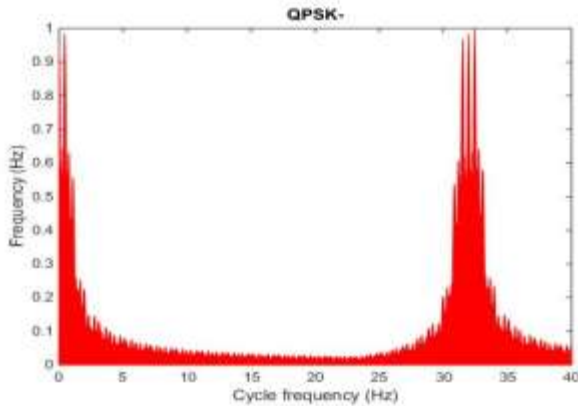


Figure 7: Spectrum of QPSK Signal

Relation between probability of detection and probability of false alarm for QPSK signal is shown below. As the no. of sample increases, the detection probability also increases.

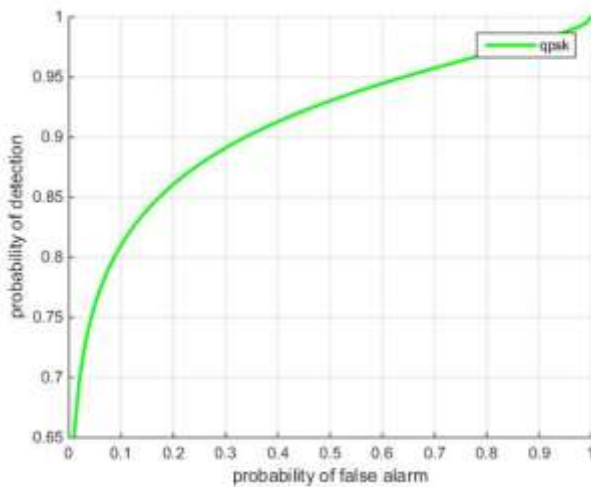


Figure 8: ROC Curve for QPSK signal

ROC curve of different modulation techniques are shown below when sampling rate  $N=200$  and SNR of detector is  $-5\text{db}$

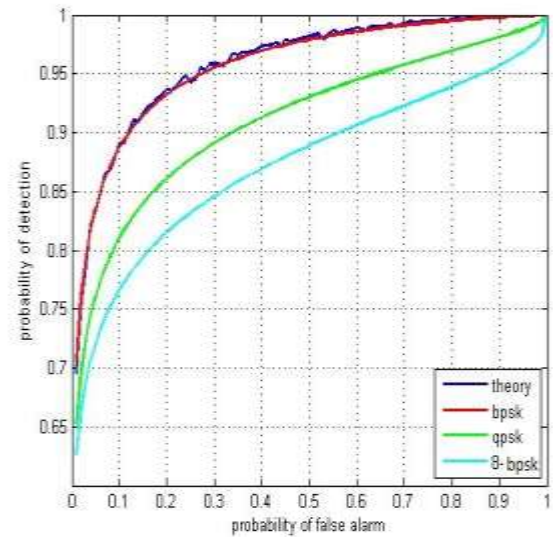


Figure 9: Comparison of ROC Curves

Statistic Cyclostationary Feature Detection is more flexible with better results and fast computation than that of conventional detection technique. The results can be improved by increasing sample size but computational time will be increased.

#### B. Simulation of channel estimation for the OFDM

Following figure shows the graph of Bit Error Rate (BER) v/s Signal to Noise Ratio (SNR) while applying no channel estimation and by applying Least Square (LS) method and Minimum Mean Square Error. It is clearly illustrated from the result that Ls method will produce better signal to noise ratio when the parameters are assumed to be  $BW=8.75\text{MHz}$ ,  $\text{FFT size} = 1024$ ,  $T_s = 1e-7$  whereas CP is fixed at  $1/8$ .

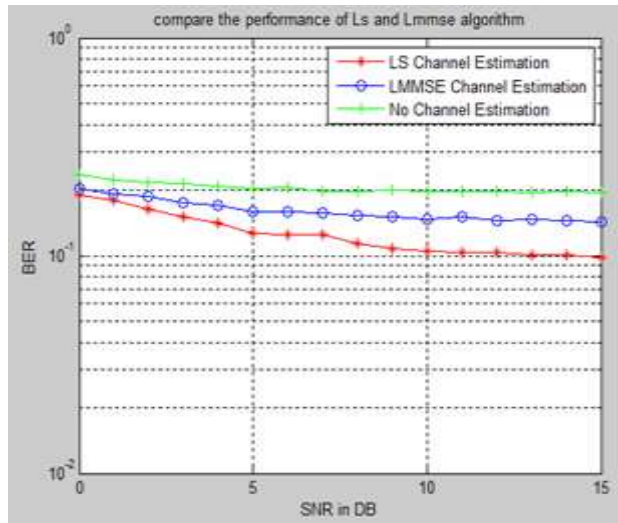


Figure 10: Performance of different Channel Estimation

### C. Simulation of LMS iterative algorithm

For initial LS channel estimation pilots are inserted among the data. Channel is modeled by multipath Rayleigh fading channel. Channel bandwidth is set to be 1.75 MHz. FFT size of 1024 and CP length of 256 was used. In which QPSK or BPSK modulation was chosen for pilots. Pilots are placed at the distance that satisfy the sampling theory. At the receiver end CP is removed. Pilots are extracted after FFT operation for LS estimation and linear interpolation.

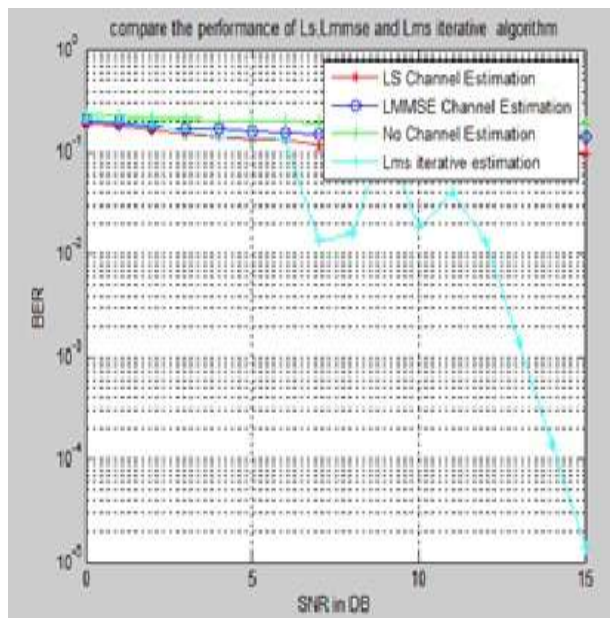


Figure 11: LMS Iterative

### III. Conclusions

In this paper statistic cyclostationary detection algorithm is proposed after analyzing the other detection technique and also LMS iterative algorithm is added to the system to improve channel estimation performance. First approach is based on cyclostationary feature and theory of Hilbert Transformation and on comparison we found out that this approach is more flexible than the conventional energy detection techniques. Also this approach works great where power is unknown to secondary or cognitive user. In second approach a receiver was proposed whose structure is less complex than LS method and linear interpolation is done. Then accuracy of channel estimation is done by using LMS iterative algorithm. It improves BER performance of the system closed to the ideal channel performance.

### REFERENCES

- [1] I. Mitola, J. and J. Maguire, G. Q., "Cognitive radio: making software radios more personal," IEEE Personal Communication Mag., vol. 6, no. 4, pp. 13–18, Aug. 1999.
- [2] D. Cabric, S. Mishra, and R. Brodersen, "Implementation issues in spectrum sensing for cognitive radios," in Proc. Asilomar Conf. on Signals, Systems and Computers, vol. 1, Pacific Grove, California, USA, Nov. 2004, pp. 772–776.
- [3] S. Haykin, "Cognitive radio: brain-empowered wireless communications," IEEE J. Select. Areas Communication, vol. 3, no. 2, pp. 201–220, Feb. 2005.
- [4] Yang Liu K. Maeda, A. Benjebbour, T. Asai, T. Furuno, and T. Ohya, "Recognition among OFDM-based systems utilizing cyclostationarity-inducing transmission," in Proc. IEEE Int. Symposium on New Frontiers in Dynamic Spectrum Access Networks, Dublin, Ireland, Apr. 2007, pp. 516–523.
- [5] Sid Ahmed Chouakri, Mohammed El Amine Slamati and Abdelmalik Taleb-Ahmed in "Implementation of a BPSK Modulation based Cognitive Radio using Energy Detection Technique" 7th International Symposium 2008".
- [6] Tefvik Yucek and Huseyin Arslan "A Survey of Spectrum Sensing Algorithms for Cognitive Radio Applications" IEEE Communications Surveys & Tutorials, Vol. 11, No. 1, First Quarter 2009.
- [7] Shengli Xie, Yi Liu "A Parallel Cooperative Spectrum Sensing in Cognitive Radio Networks" in IEEE Transactions on Vehicular Technology, Vol.59, NO.8, October 2010.
- [8] Abbas Taherpour, Masoumeh Nasiri-Kenari, and Saeed Gazor, "Multiple Antenna Spectrum Sensing in Cognitive Radios" IEEE Transactions on Vehicular Technology, Vol.59, NO.8,

February 2010.

- [9] Chien-Hwa Hwang, Guan-Long Lai, and Shih-Chang Chen in “Spectrum Sensing in Wideband OFDM Cognitive Radios” IEEE Transactions on Digital Signal Processing (Volume: 56, Issue: 2 )’ 2010.
- [10] Ashish Bagwari & Geetam Singh Tomar in “Adaptive double-threshold based energy detector for spectrum sensing in cognitive radio networks” International Journal of Electronics Letters 2012.
- [11] Fazlullah Khan in “Comparative Study of Spectrum Sensing Techniques in Cognitive Radio Networks” in in 7<sup>th</sup> International Conference on Wireless Communications, Networking and Mobile Computing (WiCOM), 2013 pp. 1 - 4.
- [12] S.Shobana, R.Saravanan, R.Muthaiah in “Matched Filter Based Spectrum Sensing on Cognitive Radio for OFDM WLANs” International Journal of Engineering and Technology 2013.
- [13] R.Gill, A.Kansal “Comparative Analysis of the Spectrum Sensing Techniques Energy Detection and Cyclostationary Feature Detection” ISAPE '14 in 7th International Symposium 2014”.
- [14] Won Mee Jang in “Blind Cyclostationary Spectrum Sensing in Cognitive Radios” IEEE Communication Letters, Vol. 18, No. 3, March 2014.
- [15] Bhowmick, Abhijit Das, Mrinal K. Biswas, Joydeep Roy, Sanjay in IEEE 2014 IEEE International Advance Computing Conference (IACC) - Gurgaon, India (2014.02.21-201).
- [16] Guolin Sun, Guisong Liu “SDN Architecture for Cognitive Radio Networks”2014 1st International Workshop on Cognitive Cellular Systems (CCS) - Germany (2014.9.2-2014.9.4) 2014 1st International.
- [17] Pratik R Parekh, Mehul B Shah “Spectrum Sensing in Wideband OFDM based Cognitive Radio” International Conference on Communication and Signal Processing, April 3-5, 2014, India.
- [18] Yang Mingchuan “Cyclostationary Feature Detection Based Spectrum Sensing Algorithm under Complicated Electromagnetic Environment in Cognitive Radio Networks” Synergetic Radio Cooperative and Collaborative Radio Sept 2015.
- [19] Jin Lai, Eryk Dutkiewicz, Ren Ping Liu, and Rein Vesilo “Opportunistic Spectrum Access with Two Channel Sensing in Cognitive Radio Networks” IEEE Transactions on Mobile Computing, Vol. 14, No. 1, January 2015.cha
- [20] S. Chatterjee P. Banerjee M. Nasipuri, “Enhancing Accuracy of Localization for Primary Users in Cognitive Radio Networks” in Vehicular Technology, IEEE Transactions on (Volume: 56, Issue: 2), 2015
- [21] “Cyclostationary Signal Processing and its Generalizations” IEEE Statistical Signal Processing Workshop (SSP 2014).
- [22] Hou, X., Li, S., Liu, D., Yin, C., and Yue, G., “On Twodimensional Adaptive Channel Estimation in OFDM Systems,” *60th IEEE Vehicular Technology Conference*, Los Angeles, Ca., vol. 1, pp. 498–502, Sept. 2004.
- [23] Sanzi, F., Sven, J., and Speidel, J., “A Comparative Study of Iterative Channel Estimators for Mobile OFDM Systems,” *IEEE Transactions on Wireless Communications*, vol.2, pp. 849–859, Sept. 2003.
- [24] Li, Y., “Simplified Channel Estimation for OFDM Systems with Multiple Transmit Antennas,” *IEEE Transactions on Communications*, vol. 1, pp. 67-75, January 2002.
- [25] Auer, G., “Channel Estimation in Two Dimensions for OFDM Systems with Multiple Transmit Antennas,” *GLOBECOM*, pp. 322–326, 2003.
- [26] IEEE P802.16 (Draft 8, May 2005), WiMAX Specification.
- [27] 1Mona Nasserri and 2Hamidreza Bakhshi 1Department of Electrical, Science and Research Branch, Islamic Azad University, Tehran, Iran 2Department of Electrical, Shahed University Tehran, Iran “Iterative Channel Estimation Algorithm in Multiple Input Multiple Output Orthogonal Frequency Division Multiplexing Systems” *Journal of Computer Science* 6 (2): 224-228, 2010 ISSN 1549-3636© 2010 Science Publications