

## A combined spectrum sensing method based DCT for cognitive radio system

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### ABSTRACT

In this paper a new hybrid blind spectrum sensing method is proposed. The method is designed to enhance the detection performance of Conventional Energy Detector (CED) through combining it with a proposed sensing module based on Discrete Cosine Transform (DCT) coefficient's relationship as operation mode at low Signal to Noise Ratio (SNR) values. In the proposed sensing module a certain factor called Average Ratio (AR) represent the ratio of energy in DCT coefficients is utilized to identify the presence of the Primary User (PU) signal. The simulation results show that the proposed method improves PU detection especially at low SNR values.

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## 1. INTRODUCTION

The modern requirements on wireless devices services such as smart phones, Tablets, and new applications that allow multimedia exchanging becomes massive and bandwidth consumer services. This, in turn, leads to scarcity in spectrum availability. Cognitive Radio (CR) is an emerging technology demanding to achieve Dynamic Spectrum Allocation (DSA) that aims to compensate the spectrum stinginess. According to the Federal Communication Commission (FCC) survey revealed that the daily use of licensed spectrum by the Primary User (PU) is lower than 30% in some geographic area [1-4]. By using CR, these opportunities can be utilized in spectrum such as in TV band and 4G cellular networks to overcome the spectrum problem and optimize resources usage. CR as a system consists of the life cycle the basic and more sensitive step is spectrum sensing cycle. Spectrum sensing is the process that achieved by Secondary User (SU) side to know the PU status which either idle or busy with a purpose to detect the spectrum holes [5-9]. There are many narrow-band spectrum sensing methods such as Matched Filter [10], Cyclostationary [11], Energy Detector [12], features based [13] and eigen value based [14]. On the other hand, there are many spectrum sensing methods specialized for wide-band channels such as compressive sensing [15, 16] and edge based detection methods [17]. The main challenge of spectrum sensing is achieve a spectrum sensing method that mixes the sufficient detection accuracy at low SNR values with acceptable processing complexity. Solving this challenge will act to enhance the overall spectrum utilization and increasing the total network throughput. In this paper, a hybrid energy detection method mixes the Conventional Energy Detector (CED) with an assisting sensing module based on the Discrete Cosine Transform (DCT) is proposed.

## 2. RELATED WORKS

Many efforts are proposed in the literature to improve the detection performance of energy detection based spectrum sensing using a statistical measure of PU and noise signals. For example, in [18] the authors introduced a sensing method use a statistical parameter which represents the ratio between variance and mean values of energy as an indicator whether the received signal is a PU or noise alone. In [19], the authors proposed a spectrum sensing method based on using noise uncertainty factor to update the threshold value of the energy detector. In [20], the authors proposed an adaptive spectrum sensing method based on higher order moment and cumulate with a multilevel threshold for decision stage. There are another efforts to developing CED by utilizing another transformation approaches for example In [21] the authors designed a new energy based spectrum sensing using DCT instead of Discrete Fourier Transform (DFT), this was a good method but not take into account the noise signal recognition from PU signal from DCT coefficient's relationship.

According to the related works above, it can be seen that most of these efforts focused on utilizing complex approaches to achieve high detection accuracy and did not take into account the recognition performance of PU and noisy signals based on variation resulting from transformation of these signals. Another most important note can be seen, is that the all above works did not utilize the traditional energy detection method as assisting module to simplify the detection complexity when the received signal falls in medium SNR range, which in turn can provide acceptable sensing accuracy with low overhead that gives short detection time and increased network throughput. In this paper, a new approach to distinguish between PU and noise only signals by utilizing DCT transform features is proposed. The distribution of energy in DCT sub-bands, fortunately, show a wide variance between PU and noise alone signal cases especially at lower sub-bands. This measure is used to support CED detector when it fails to produce the right decisions at low SNR values

## 3. CONVENTIONAL ENERGY DETECTOR (CED)

Conventional Energy Detector (CED) is the simple context of ED method in which the Secondary User (SU) accumulates the received energy with respect to a certain number of samples ( $N_s$ ). The binary hypothesis used in the detection criteria is as follows [22]:

$$y(n) = \begin{cases} w(n) & ; H_0 \\ s(n) + w(n) & ; H_1 \end{cases} \quad (1)$$

Where  $y(n)$  is the received sample,  $w(n)$  is the Additive White Gaussian Noise AWGN sample collected from the transmission channel,  $s(n)$  is the PU signal sample,  $H_0$  and  $H_1$  are null and power hypothesis respectively. The test statistic of the received sample is defined as follows:

$$T = \frac{1}{N_s} \sum_{n=1}^{N_s} |y(n)|^2 \quad (2)$$

Where  $N_s$  is the total number of received samples. The distribution of  $T$  is chi-squared with  $2N_s$  degrees of freedom [23]. The received samples are compared with predefined threshold computed based on Constant False Alarm probability (CPFA) [24, 25]. The performance of the detection in energy detector schemes is evaluated based on Receiver Operation Characteristics (ROC) that showing detection probability ( $P_d$ ), False Alarm Probability (PFA), Signal to Noise Ratio (SNR), and threshold ( $\lambda$ ). ROC parameters are computed in AWGN as follows [26].

$$P_{D=} P[T > \lambda |_{H_1}] Q\left(\frac{\lambda - N_s(1+SNR)}{\sqrt{2N_s(1+SNR)}}\right) \quad (3)$$

$$P_{FA=} P[T > \lambda |_{H_0}] Q\left(\frac{\lambda - N_s}{\sqrt{2N_s}}\right) \quad (4)$$

Where here  $Q$  is complementary Q-function (Marcum),  $P_d$  represents the correct detection when PU is present, PFA represents false detection alarm that indicates the presence of PU but, actually it is not there. PD and PFA in (3) and (4) are calculated in case of unity noise variance. From (4)  $\lambda$  is calculated as follows:

$$\lambda = \sqrt{2N_s} Q^{-1}(P_{FA}) + N_s \quad (5)$$

where  $Q^{-1}$  is inverse Q-function. The system model of CED method is shown in Figure 1 [7].

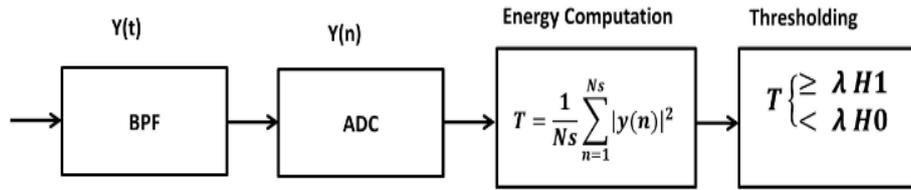


Figure 1. CED block diagram

**4. THE PROPOSED METHOD**

The method mixes the CED with an enhancement sensing module based on DCT statistics as shown in Figure 2. This method has two operation modes. The first mode is performing spectrum sensing using CED method when the received energy is greater or equal to the predefined threshold. When the received energy is less than the predefined threshold, the second mode is initiated where enhancing module based on DCT statistics is activated. The enhancement module will act to make further verification about the hypothesis decision produced by CED energy detector. The verification mechanism can be explained as follows: first DCT transform is computed to received samples using (6) and (7) [27]:

$$A[0] = \frac{1}{\sqrt{N}} \sum_{n=0}^{N-1} a[n], M=0 \tag{6}$$

$$A[M] = \frac{2}{N} \sum_{n=0}^{N-1} a[n] \cos\left(\frac{\pi K(2n+1)}{2N} M\right), 1 \leq M \leq N-1 \tag{7}$$

Where  $a[n]$  and  $A[M]$  are the original and DCT  $N$ -point signals. After that, the whole DCT spectrum band (range) of length  $N$  is divided into  $K$  non-overlapped sub-bands of length  $N/K$  as shown in Figure 3 such that the  $i$ th sub-band has local energy  $E_i$  ( $i=1,2,\dots, K$ ). The energy of first and last sub-bands ( $E_1, E_K$ ) are calculated using (8) and (9) then Average Ratio (AR) between  $E_1$  and  $E_K$  is calculated using (10).

$$E_{i=\frac{N}{K}} = \sum_{i=1}^K \sum_{i=1}^{N/K} |y(i)|^2 \tag{8}$$

$$E_K = \sum_{i=N-\frac{N}{K}}^N |y(i)|^2. \tag{9}$$

$$AR = \frac{E_1}{E_K} \tag{10}$$

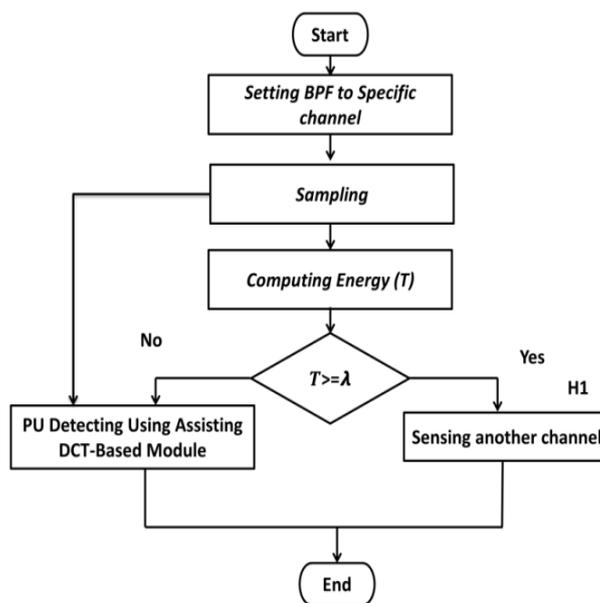


Figure 2. The proposed algorithm flowchart

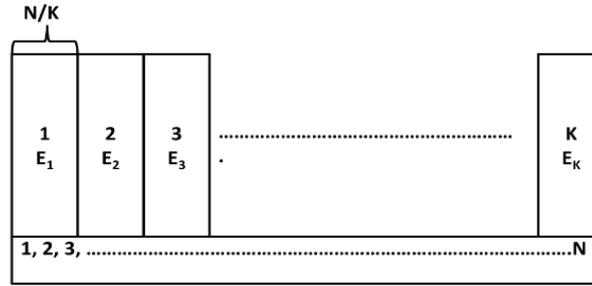


Figure 3. DCT's Sub-bands division

Based on the value of AR, the signal type can be identified which either PU (PU+Noise) or noise only signal. In the case of PU presence, the AR value will be greater than one depending on the SNR value of the transmitted signal. This is related to DCT feature that the lower bands contain the most of signal energy (information), in opposition to the higher bands that contain less energy. In the case of noise-only signal the AR will be approximately unity because there is no source of information except the noise and each sub-band approximately collects the same amount of energy. Figure 4 and Figure 5 show the testing of AR factor in two scenarios, the first scenario assumes PU is present and the second assumes PU is absent with SNR as a parameter respectively.

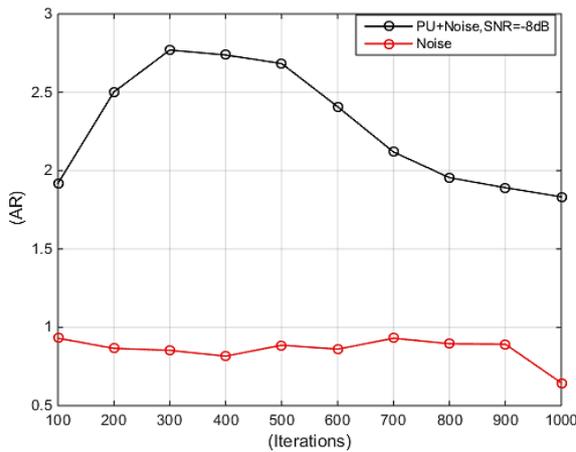


Figure 4. AR factor testing for PU and noise signals at SNR=-8dB

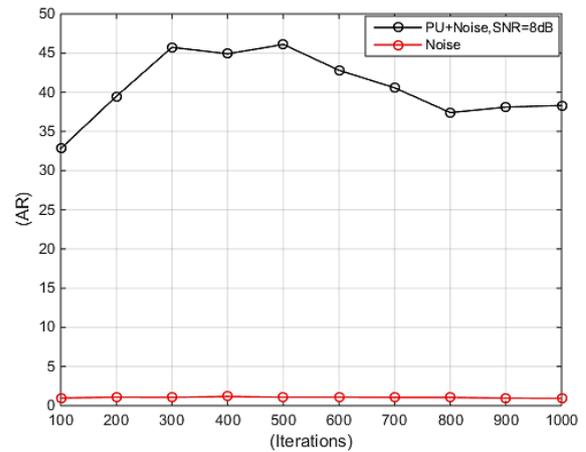


Figure 5. AR factor testing for PU and noise signals at SNR=8dB

From Figure 4 and Figure 5 it can be seen that the value of AR provides an excellent measure to the presence or absence of the PU signal. The rounding of AR will be to unity in case of noise only signal and greater than one in case of PU presence; meanwhile, it can be used as a new threshold for the use at the detection in CED assisting module. The detection mechanism is achieved by comparing the value AR with value of “1”, if the value of AR less or equal to 1 the hypothesis is H0; otherwise it is H1. The flowchart of the CED assisting sensing module is shown in Figure 6. The use of this two- mode sensing scheme will provide a good compromise between low complexity provided by CED at high SNR values and accuracy provided by DCT at low SNR values.

The detection probability of the proposed method (PDPRO) is a composite detection probability includes detection probability of CED and the DCT based assisting module as in (11).

$$PD_{PRO} = P_{INCED} * PD_{CED} + (1 - P_{INCED}) * PD_{DCT} \tag{11}$$

where  $P_{INCED}$  is the probability of falling in CED sensing event,  $PD_{CED}$  is the detection probability of the CED method, and  $PD_{DCT}$  is the detection probability of DCT based assisting module.

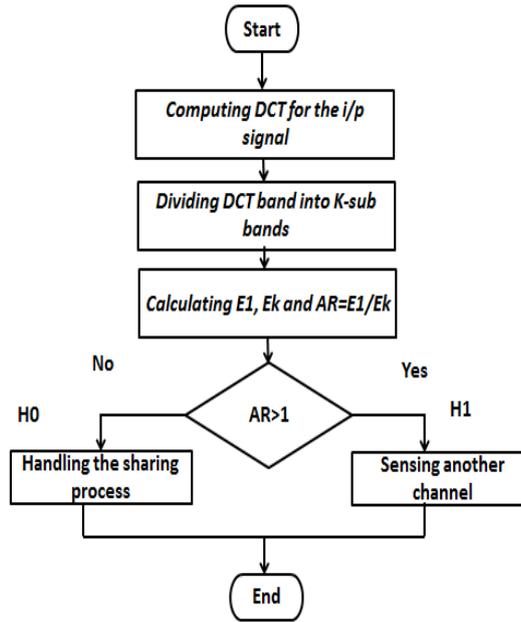


Figure 6. Flowchart of CED assisting sensing module based on DCT

## 5. RESULTS

The proposed method is simulated using MATLAB. The evaluation has been tested for Local Sensing (LS) non-cooperative scenario showing the ROC with different values of  $N_s$ ,  $P_{FA}$  and channels types. The simulation parameters used are given in Table 1. Figure 7 and Figure 8 show the performance of the proposed method versus CED in AWGN channel at  $N_s=1000$ ,  $P_{FA}=0.001$  and  $P_{FA}=0.01$  respectively. It can be seen from these figures that the proposed method significantly improve the detection probability compared to CED method and this improvement increases as  $P_{FA}$  decreases. For instance, at  $SNR=-9$  dB, the detection probability is improved by 95% and 90% using the proposed method when  $P_{FA}$  is 0.001 and 0.001 respectively. Also, we notice from the performance curves that the detection probability of the proposed method reaches its optimum value of one in about 4 dB SNR earlier than that of CED method.

Figure 9 shows the performance of proposed and CED methods in Rayleigh fading channel at  $N_s=1000$  samples and  $P_{FA}=0.001$ . Figure 10 shows the performance of the proposed method at a different number of received samples. From Figure 9 it can be seen the proposed method achieves gain of 20% improvement in detection energy at  $SNR=-11$  dB, while in Figure 10 shows the flexibility of the proposed method to operate in different number of received sample with almost similar performance especially as  $N_s=500$  and  $N_s=250$  sample.

Figure 11 shows the performance of the proposed method compared with the method in [4]. From this figure, it can be seen that our proposed method outperform the method of [4] in detection probability. For instance, at  $SNR=-10$  dB,  $P_d$  is increased from 0.5 to 0.8 by improvement factor of 60%. Also, we notice that the detection probability of the proposed method reaches its optimum value of one in about 3.5 dB SNR earlier than that of the method in [4]. Finally, the complexity of the proposed method evaluated using sensing time as depicted in Figure 12. This figure shows the Simulation time comparison between proposed method and CED. From this figure it can be seen that proposed method time consumer than CED at low SNR values when it operate DCT based sensing module, then it return to normal state when CED is operating at high SNR values.

Table 1. Simulation Parameters

Parameter	Value
PU signal Type	QPSK
Channel Type	AWGN and Rayleigh fading
$P_{FA}$	$10^{-2}$ and $10^{-3}$
Number of Samples $N_s$	1000, 500 and 250
SNR Range	-15 to 0 dB

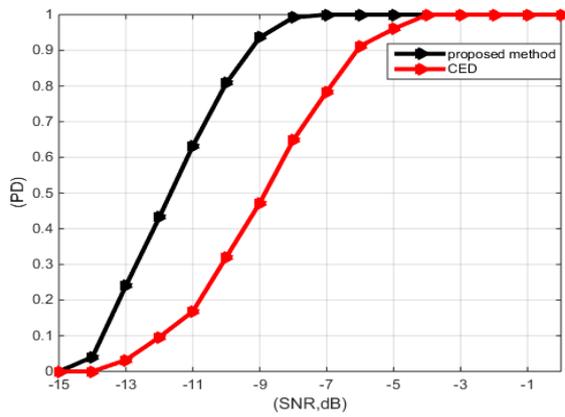


Figure 7. Performance of the proposed method with CED at PFA= 0.001 and NS=1000

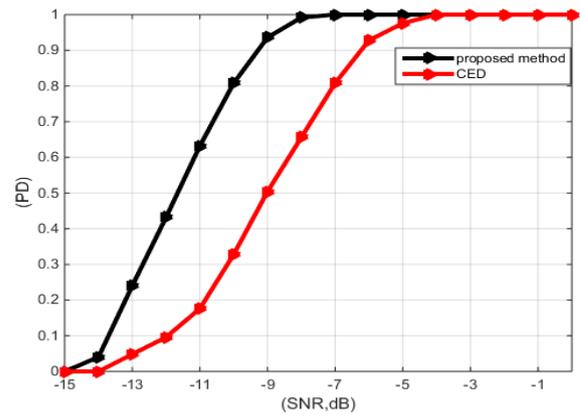


Figure 8. Performance of the proposed method with CED at PFA= 0.01 and NS=1000 Performance of the proposed method with CED at PFA= 0.001 and NS=1000

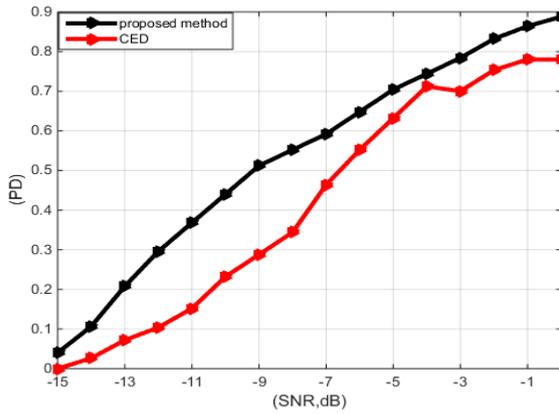


Figure 9. Performance of the proposed method and CED in Rayleigh fading channel at PFA= 0.001 and NS=100

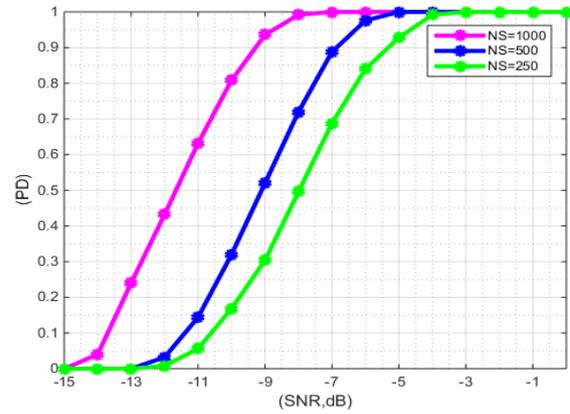


Figure 10. Performance of the proposed method at different NS in AWGN

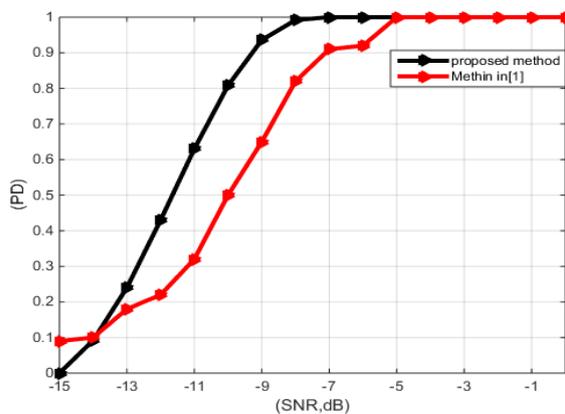


Figure 11. Performance comparison of the proposed method with method in [4]

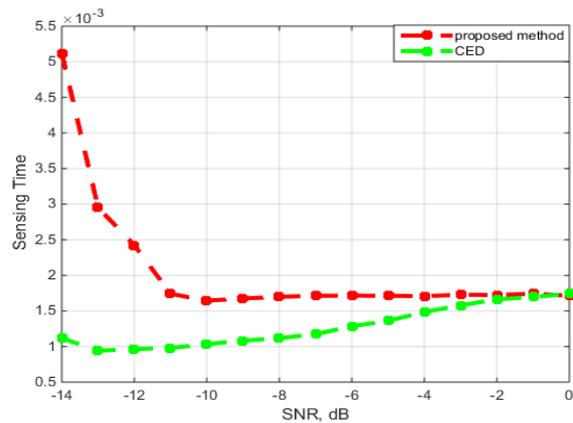


Figure 12. Complexity performance evaluation

## 6. CONCLUSION

In this paper, a hybrid blind spectrum sensing method has been presented. This method mixes CED with sensing assisting module based on DCT. This method operates in CED mode when received signal energy equal or greater than the threshold value, and operates the assisting sensing module mode when the energy falls below the threshold value. The result showed that the proposed method over perform CED in both types of tested channels, for an instant in AWGN the detection probability enhanced by 95% at SNR of -9dB. As well as in fading channel detection probability has been enhanced by 90% at SNR of -9 dB. According to the provided advantages the proposed spectrum sensing method, it represents a suitable method for sensing a low SNR for different signal type without synchronization due to it is blindness nature with energy saving facility that resulted from the switching between CED and the proposed sensing module with respect to the received signal strength.

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