Optimization of Collaborative Spectrum Sensing Mechanism for Cognitive Radio

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Abstract: Cognitive radio (CR) is an enabling technology that helps efficiently use spectrum. In this paper, a collaborative spectrum sensing mechanism is proposed which combines passive detection and sensing service compensation. Theoretical analysis and simulations show that this method not only reduces the energy cost and the passive sensing time, but also maintains fairness in the CR communication systems.

Keywords: Cognitive radio, collaborative spectrum sensing, detection time, sensing mechanism.

1. INTRODUCTION

With the development of wireless communication technology, the contradiction between performance and cost rises. And the development is blocked by the scarcity of spectrum resources and the inefficiency of static spectrum allocation strategy. To improve the spectrum utilization, Dr. Joseph Mitola proposed “cognitive radio” [1]. Its core idea is that the wireless communication device can find the “spectrum hole” in the licensed spectrum band, and take advantage of it with no interference to the primary user (PU). This technology can significantly improve the utilization of spectrum resources and effectively alleviate the problem of spectrum scarcity.

Before the secondary user (SU) can send data on the licensed spectrum, it has to perceive the spectrum accurately, so spectrum sensing is one of the key technologies in cognitive radio. At present, the spectrum detect precision of single-node will be effected by the multipath fading and shadow because of the location of node, besides the computational capability of single-node is limited. As a result, its detection performance is not good. The cooperative spectrum sensing technology can overcome the above shortcomings, but there are a lot of contradictions between performance and efficiency. How to improve the efficiency of spectrum sensing and spectrum utilization is the key to the collaborative spectrum sensing. There are several researches and optimizations for cooperation spectrum sensing in [2-9]. Distributed collaboration network structure is researched in [2-4] without considering cognitive users’ computing burden, perception time and throughput is optimized in [5], how to lighten the cognitive networks burden is researched in [6,7] the author maximizes the spectrum sensing efficiency in Gaussian noise channel and Rayleigh fading channel in [8]. All of these studies have adopted active spectrum sensing mode without considering the energy cost of cognitive users. In [9], active spectrum sensing is set a cycle, but it is unrealistic to force all users to periodically percept to maintain the whole networks. In response to these problems, a collaborative passive spectrum sensing mode combined with sensing service compensation is proposed. Passive spectrum sensing mode can reduce the average energy cost of the whole cognitive networks, and the sensing service compensation can shorten the spectrum sensing time to make up for the shortcoming of passive mode.

The rest of paper is organized as follows: section 2 introduces the proposed model which combines passive spectrum sensing mode with sensing service compensation, section 3 analysis the feasibility of proposed model and addresses the optimization for the performance of spectrum sensing from simulation results, section 4 describes the advantages of proposed model in energy cost, finally section 5 draws conclusions.

2. SYSTEM MODEL AND PROBLEM STATEMENT

The passive spectrum sensing mode is known as its long perception time. And the shortcoming will lead to missing the spectrum opportunities and making the channel switching delay longer. Before trying to find a new mechanism to overcome the shortcoming, let’s see the working mechanism of the traditional passive spectrum sensing. Fig. (1) illustrates a SU’s complete communication process. There are four time periods between “demand” and “start transmission”, but it is unrealistic to force all users to periodically percept to maintain the whole networks. In response to these problems, a collaborative passive spectrum sensing mode combined with sensing service compensation is proposed. Passive spectrum sensing mode can reduce the average energy cost of the whole cognitive networks, and the sensing service compensation can shorten the spectrum sensing time to make up for the shortcoming of passive mode.

There is less difference between the two processes in total time. In the proposed program, periods of sensing and data integration are removed before transporting, spectrum sensing will be carried out after the SU completes its transmission and data integration runs asynchronously along with the entire process. In the cooperative spectrum sensing
scheme, one SU can take advantage of others’ sensing result for its own service as long as results of other SUs are reliable enough. If one SU uses others’ spectrum sensing result ignoring its own result, there is no need for the FC to fuse data synchronously. In this way, the period before data transmission has been shortened, finally the performance of spectrum detection will be better.

Fig. (1). The flow chart of traditional model.

The theoretical challenge faced by the proposed program is that SU must obey the principle of “listen before transmitting” through traditional spectrum sensing theory [10]. It is necessary for the single cognitive user, because it’s the only way that it gets available channel. However in the practical situation, single user affected by multipath fading and shadow performances bad on detection results. Cooperative spectrum sensing technology has been applied to solve this problem. In collaborative networks, several SUs send their local sensing results to FC where data will be fused based on appropriate data fusion rules(such as “AND” “OR” “K/N”, etc.). The final judgment will be given by FC, and this result has been improved compared to that of single users.

Assume that there are N cognitive users which are independent and identically distributed in the network, the average SNR of each user is the same. Therefore the probability of detection (Pd) for each user is the same when energy detection method is carried out. No matter what kind of data integration is used, the quality of detection (Qd) in FC will meet the detection precision (R) when N is large enough. With “OR” fusion rule, for example, N is required to meet

$$Q_d = 1 - (1 - P_d)^N \geq R$$  \hspace{1cm} (1)

Simulation results in Fig. (3) show that with the Pd different, Qd requires different N to reach detection precision (R). As N increases, Qd always can meet the requirements. Meanwhile Qd almost increases no longer after N reaches to a certain point which can be set when Qd reaches to 0.95. Then the point is the optimal number (Nop) for coordination users, larger waste energy and smaller lead to bad performance.

Fig. (3). the relationship between N, Pd and Qd

In the collaborative cognitive radio networks, there will be several SUs to sense spectrum including the SU itself which has data transmission demand. If the number of these SUs is larger than Nop+1, the SU which needs transmission can skip its sensing period and use others’ sensing result, so that, the sensing delay will be shortened and the perception precision will be guaranteed.

3. PERFORMANCE SIMULATION

The performance of the proposed scheme is simulated in MATLAB. A collaborative cognitive radio network is considered, with 10 licensed channels changing randomly and uniformly, spectrum detection based on energy and the number of SUs in line with “Poisson Process”. We did a survey of the average number of success detection in a period of time between traditional passive spectrum sensing model and proposed model under AWGN channel and Rayleigh channel. Fig. (4) shows the proposed scheme has improved the detection performance than the traditional passive detection scheme whether in AWGN channel or Rayleigh channel. And with the improvement of the SNR, the improved performance of proposed scheme is more pronounced and stabilized. As a result of the cooperative sensing strategy, the detecting precision in Rayleigh channel is close to that in AWGN channel. In the certain situation, the detection performance of the proposed scheme under Rayleigh channel is better than that of the traditional scheme under AWGN channel.
The change rate of licensed channel is another important factor to affect the detection performance of the two schemes. Fig. (5) shows that how much the proposed scheme can improve in detection number than the traditional scheme both in AWGN channel and in Rayleigh channel. X-axis represents channel numbers that the proposed scheme detected more than the traditional scheme, and Y-axis represents how often licensed channels change. Obviously, the faster detection scheme has more obvious advantage than the slower one in the case of channel changing fast. So the proposed scheme is more applicable in the networks where channel states change fast.

Fig. (5). Detection difference with channel change.

4. OPTIMIZATION OF ENERGY CONSUMPTION

Saving energy is another prominent advantage of the proposed scheme. As is shown in Fig. (6), there are different SUs divided into active SU and sleep SU in the networks. A SU will be active if it has data transmission demand, and goes through the process shown in Fig. (2), otherwise it will sleep. Compared with traditional scheme which requires all SUs to sense spectrum, the proposed scheme will obviously save a lot of energy.

A specific example will be given to illustrate the proposed scheme on energy saving. Assume in a multimode and heterogeneous cognitive radio network, a large number of nodes are allowed to communicate at the same time. If there are some SUs (the number is equal or larger than Nop) completing their communication and spectrum sensing in a short period, FC can get the last information about spectrum in this network. Any other SUs that have communication demand at this moment can ask the FC for channel state directly, and don’t need to wait for the results of all SUs in this network. If the Pd of single SU is 0.7, the FC just needs 4 SUs to reach high precision (R>0.95). According to the U.S. Federal Communications Commission (FCC), the utilization of the licensed spectrum band is only 15% to 85% [11]. In the case that the idle spectrum is taken full advantage of, the above condition is easily satisfied.

Fig. (6). Network state diagram.

If someone still think the conditions mentioned above are some harsh, we can relax the requirements and prove its feasibility using experiments. Now there is no necessary for some SUs to send their sensing results when a SU has communication demand. FC just needs to fuse the sensing results in a period before the present moment. In order to simplify the problem, we still assume that channels change uniformly and the number of SUs in line with “Poisson Process”. When we fix the parameters of SUs, test results affected by the channels’ change cycle and FC’s fusion period are shown in Fig. (7). As is shown, the detection results perform best when the fusion period in FC is shorter than the changing cycle of channels, if there are enough sensing results.

Fig. (7). Detection results affected by channel cycle.
Now we fix the channel change cycle for 1s, and adjust the “t”, a parameter of “Poisson Process”, which will influence the number of SUs’ sensing results in a period of time. We can get Fig. (8) and draw a conclusion that with the sensing results reduce, the optimal fusion period is not always equal to channel change cycle.

Fig. (8). Detection results affected by parameter “t”.

Since the parameter “t” can influence the last detection result, the average number of SUs influenced directly by it should be researched. We set another parameter $\lambda = 0.2$ in order to make the number change significantly. Fig. (9) shows that when the parameter “t” is greater than 0.4, there will be no more than two SUs’ sensing results sent to FC to finish data fusion. It can’t meet the basic requirements of cooperative perception.

Fig. (9). Average number of SUs affected by parameter “t”.

Now we can draw the conclusion. If we know channels’ average change cycle according to the statistics law and the average number of SUs is larger than 2.5 which will send their sensing results to FC in one channel change cycle, the proposed scheme is applicable.

Since we know the feasibility of the proposed scheme, assume that the number of SU is $N (N >> N_r)$ in the cognitive radio network and each SU’s energy consumption for spectrum sensing is $E$. In a unit of time, there are $N_a$ active SUs and $N_s$ sleep SUs ($N = N_a + N_s$). So the energy consumption of the proposed scheme is $(E \times N)$, compared to $(E \times N)$ of the traditional scheme. It can save a lot of energy especially in a large network.

This scheme can also ensure the fairness of communication when saving energy. Some SUs will be crazy to find their mobile terminals’ batteries died quickly because the traditional scheme asks all SUs to sense spectrum whether they need to communicate or not. It will hinder the development of cognitive radio technology. The proposed scheme inherits the advantages of the technology of passive spectrum sensing and makes the cognitive radio technology more realistic.

**CONCLUSION**

The paper studied and optimized the collaborative spectrum sensing scheme for cognitive radio networks, including the perception time and the energy consumption. A passive spectrum sensing mode combined with perception service compensation was proposed in a centralized collaboration network structure. With the help of the advantages of centralized network and the flexible working of collaborative spectrum sensing, spectrum sensing time was shortened along with saving energy consumption. Theoretical analysis and simulation show that the proposed scheme not only meets the requirements of collaborative spectrum sensing in performance, but also saves energy. It is significant for the mobile wireless communication which aspires to flexible mobile and lasting work.

**CONFLICT OF INTEREST**

The authors confirm that this article content has no conflict of interest.

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**REFERENCES**


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