

Improving the Energy Efficiency of Cognitive Radio Wireless Network Using Coverset Prediction

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Abstract: The cognitive radio-based unique range access instrument is unrest in wireless correspondence, which eases the range use issue. Cognitive Radio Networks (CRNs) furnishes portable clients with high transfer speeds through effective range access plans and heterogeneous wireless designs. Nonetheless, these networks accompany different difficulties ascribable to the accessible range's fluctuating attributes, alongside the changed Quality of Service (QoS) prerequisites of the various applications. Cognitive Radio-Wireless Sensor Networks (CR-WSNs) are approaching as basically one of the most guaranteeing wireless innovations of things to come for range access. This paper presents a calculated plan of CR-based WSNs, improving the energy efficiency utilizing coverset prediction with EECSP Algorithm, and recommend potential solutions for beat the difficulties. As an illustration, we study the presentation of CR-based WSN utilized for energy efficiency with the packets' directing are made utilizing the coverset.

Key words: CRWSN, QoS, Energy Efficiency, Coverset, Data Transmission, EECSP.

1. Introduction

The field of wireless correspondence has noticed enormous development throughout the years [2]. A sensor is an electro-mechanical gadget with a battery and contraption to quantify changes in the climate/environmental factors. Instances of sensors incorporate the thermometers that specialists use to gauge a patient's internal heat level, seismic tremor sensors to quantify the quake's force, and fire sensors to quantify the chance of fire by detecting smoke. Different sensors are accessible, relying upon the usefulness like sensors utilized in transportation, synthetic industry, estimating light and properties, warm sensors, and nearness sensors. As sensors are situated in thick woodland zones/distantly, it isn't easy to have a wired association, so the sensors are associated through a wireless medium [1]. Cognitive radio innovation [1], [3] has been created to improve radio asset utilization of the wireless network climate. As of late, different wireless administrations have been broadly sent. The measure of portable traffic is ceaselessly and quickly expanding. To fulfill quite an appeal for portable interchanges, versatile wireless networks' ability must be improved, requiring different radio-recurrence groups [3]. Late networks comprise numerous heterogeneous clients, nodes, and connections working under powerful conditions. Taking care of and proficiently dealing with all these expanding complexities is unimaginable with manual administration and requires programmed administration without human intercession [9].

The contrast between WSN and CWSN has appeared in Figure. 1. The Figure shows that the CWSN has an extra operation where the WSN has no operations.

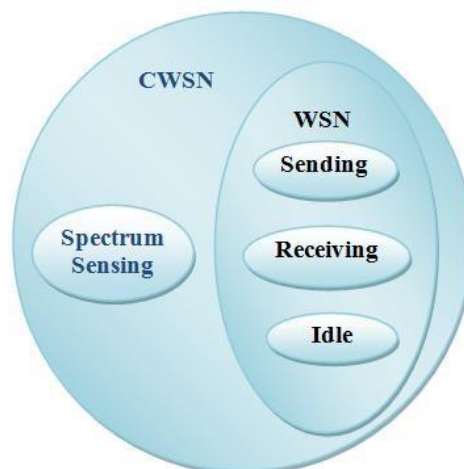


Figure 1: WSNs.CWSN

Lately, CWSN has arisen as another innovation that has been the time of cognitive networking popularity. The specialists concentrate on prior degrees of advancement [10]. As a general rule, real WSN faces different issues leads from the compelled transfer speed designated to them that are typically unlicensed groups. The design of Cognitive WSN is meant in Figure 2.

As sensors are situated in thick woods zones/distantly, it isn't easy to have a wired association, so the sensors are associated through a wireless medium. A wireless sensor network (WSN) is an assortment of sensor nodes arranged in distant/far off territories. Its motivation is to screen, record changes in the climate, send the data to the worker, or once in a while, start the correct conceivable activity [6]. When wireless sensors are furnished with cognitive abilities, another network method arose called Cognitive Radio Wireless Sensor Network (CR-WSNs) [11] [5]. In these networks, nodes are furnished with CRs that can detect, learn, and respond to network conditions changes.

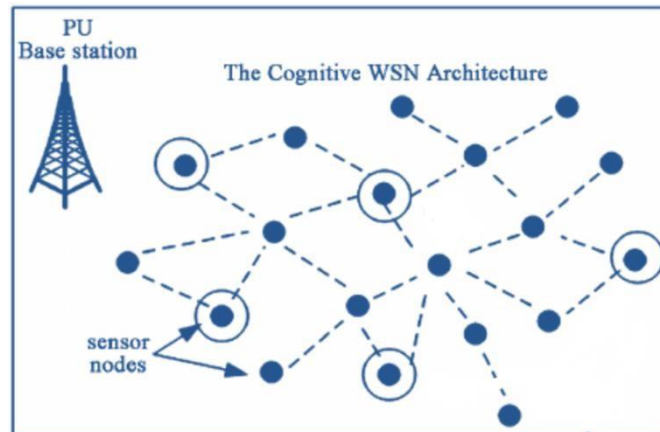


Figure 2: Architecture of CWSN.

Hence, a versatile transmission power control for energy reaping CRWSN is proposed in this work. The focal idea is to utilize the coverset prediction to build start to finish network execution. Accordingly, this new methodology decreases the traffic load on the low force nodes and develops on the nodes with higher accessible force. To accomplish this, nodes with higher force increment their transmission power, bringing about changes in the network's coherent geography that decline the nodes' traffic load with lower energy. Realizing that the dead nodes make a few broken connections inside the web keeps nodes from running out of intensity; along these lines, it keeps up network availability.

This paper's remaining part is coordinated as follows: Section 2 talks about the connected work in energy efficiency discovery in CRWSN. The proposed technique is presented in Section 3, and the outcomes are introduced in Section 4. At long last, the paper is closed in Section 5.

2. Background Study

Arched, J., et al. [1] the cognitive radio distinguishing is a rising field that utilizes cognitive radio to detect, learn, and react to framework condition changes. When incorporated with WSN, it deals with essentialness capability, information move limit partition, and steering. A batching plan for improving essentialness profitability will be completed.

Supraja, P. et al. [2] Spectrum fluctuates starting with one spot then onto the next because of its irregular nature; there is a huge interest for this scant asset, so it's smarter to comprehend the conduct of the range. Range detection is a vital thought for the improvement of cognitive radio usage. The cognitive radio ought to foresee the range with practically zero blunder to dodge impedance. The point is to build up a model fit for foreseeing this conduct.

Tumuluru, V. K., et al.[4] Channel status prediction is imperative to cognitive radio networks. It can enormously spare the detecting energy and help the optional clients misuse the range openings all the more proficiently. A dependable channel status prediction instrument ought to guarantee a lower likelihood of wrong channel status predictions.

Joshi, N. et al. [5] The cognitive radio recognizing is a rising field that utilizes cognitive radio to detect, learn, and react to framework condition changes. When incorporated with WSN, it deals with imperativeness capability, information move limit partition, and steering. A clustering plan for improving essentialness efficiency will be realized.

Yoon Hyun Kim et al. [7] proposed the serious sign recognition strategy utilizing the computerized watermarking grouping for cognitive radio-based wireless impromptu networks in a co-channel obstruction climate. We show that the proposed plot has lower multifaceted nature and higher location likelihood with

common framework SNR and low watermarking succession level from the framework model. Recreation results show the recognition likelihood with - 27dB advanced watermarking. Additionally, reproduction results show that signal recognition conspire based computerized watermarking succession has effective sign location execution for CR based wireless impromptu network.

3. System Model

The coverset network is shaped by recognizing the nodes in the coverage zone. Directing of the packets is made utilizing this coverset. It recognizes the course by distinguishing channel accessibility. The base station's coverage zone is taken, and a node inside this coverage zone is set, and its neighboring nodes are found, and packets are sent to these nodes. The distance of the packets to the objective is recognized by alluding to the nodes' query table. This distance decided is shipped off the neighboring nodes, which distinguishes the course to the objective, refreshed in the relative multitude of nodes' query tables. The likelihood of the transmission force and force picked up is set, which decides the coverset nodes when moving out of the coverage region. As the distance between the nodes changes, the energy likewise differs. Given the node's length and the coverset territory, the following best neighboring node, the next gadget, the following neighbor to be utilized for steering are recognized. This helps distinguish the channel's energy efficiency, which is free and productive utilizing hand-off node transmission.

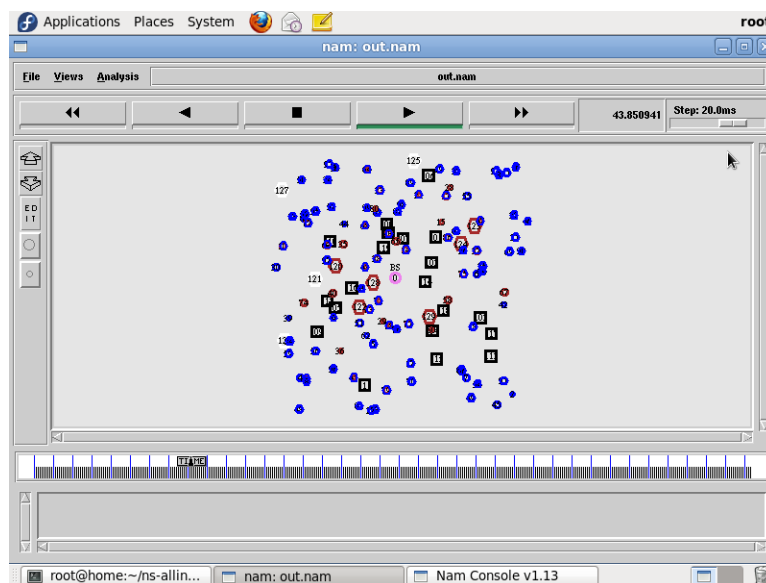


Figure 3: Construction of Nodes

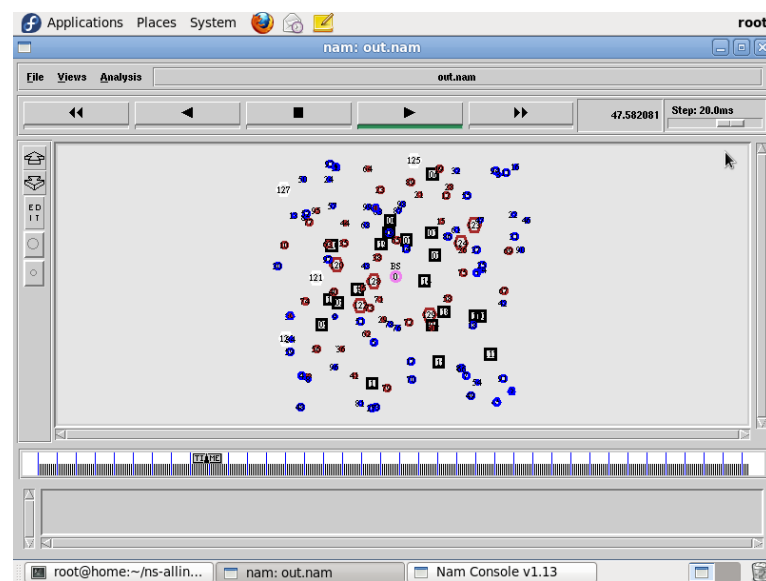


Figure 4: Efficient Data Transmission using EECSP Algorithm.

Geography Changes

Geography straightly impacts the network lifetime in CWSN. In light of the utilization of nature, sensor nodes go through the organization statically or progressively. Much of the time, the node disappointment in CWSN exists because of equipment disappointment of energy weariness. The geographies of the CWSN and WSN are equivalent. Versatile self-configuration geography is basic for CWSN to get adaptability, diminished energy use, and accomplish better outcomes. The dynamic self-configuration geography offers static geography, even though it is a requesting issue to plan and actualize. This field should be investigated more.

Energy Efficiency

In light of the packet loss, there is colossal energy squander for retransmission of packets in WSN. To change the channel's state, CR wireless sensors may be equipped for adjusting their working ascribes. Henceforth, energy usage in light of the retransmission and impact of a packet can be decreased.

Algorithm: EECSP: Energy Efficiency using Coverset Prediction

Step 1: Create a network with BS and CR, Primary User (PU).

Step 2: BS computes the channel gain based on the transmission and receiving power.

Step 3: Location, Channel Gain, Signal Strength (RSS), Tx Power, and the Transmitter node id are maintained by BS.

Step 4: BS performs the scheduling process based on the energy and power information.

Step 5: To compute the coverset formation, the tmp_list, covered_list, uncovered_list should be calculated.

Step 6: Compute the lifetime of a node.

lifetime = CURRENT_TIME * node_ ->energy_model()->initialenergy() / consumed;

Step 7: BS computes the energy harvesting factor as the energy conversion efficiency, transmission power, slot period, powered channel gain, and the non-linear energy harvesting parameter.

avg_slot = slot_duration * actual_slot;

Step 8: Interference temperature for each device is initiated.

Step 9: Calculate the secondary user's instantaneous achievable rate concerning the interfered energy harvesting sensor.

Step 10: Estimate Maximal Ratio Combining (MRC) for communication.

mrc = max(ene_harvesting) + TX power

Step 11: CR initiate learning decision process

Step 12: end

Step 13: Compute updated lifetime of covers

lifetime[i] = lifetime[i] + wi[i];

Step 14: compute probability of current target based coverset

probability[i] = etotal * cmw * cws * cea;

probability[i] = probability[i] / sum_prob;

Step 15: Compute the final coverset by sending the coverset announcement to update the node and location list.

Send CoverSet Announcement (coverset.nodeid,coverset.count);

Step 16: Predicate the energy efficiency of a node.

Step 17: Allocate Channel for PU based on efficient energy of a node.

Power Consumption

With a compelled energy source, CR wireless sensors are energy-restricted gadgets. The force needed for gathering and transmitting information packets, information handling, range detecting, course discovery, chill out, station arrangement; notwithstanding the entirety of the abovementioned, CR wireless sensors need power for regular range hand-off. A CR wireless sensor requires keen PU activity over the channel. To PU activity checking, various applications need various receiving wires; along these lines, high force is used. For energy recovery, there exist numerous kinds of examination, and these strategies contain their detriments. Supplanting the energy assets or energy recovery isn't plausible in a couple of utilizations. In those hindrances are given. Force usage is not a huge consideration in impromptu CRWSNs; in any case, it is a critical plan factor. In any case, it is one of the essential exhibitions quantifies in CRWSN, which straightly influences network life expectancy.

Quality of Service (QoS)

The QoS is normally isolated through four ascribes in customary WSNs: Jitter, data transfer capacity, unwavering quality, and Delay. WSN requires dealing with an adequate QoS level to forestall perilous outcomes in huge applications. In WSN, memory, power assets, and preparing power are the asset restrictions that the QoS uphold is a troublesome issue. It likewise experiences different issues like safeguard the PU rights to get

occupant spectra. With SU (secondary users), PU (primary users) correspondence will be without obstruction. This isn't simply that it is mind-boggling to PU appearance prediction over the channel. Bogus alert and direct sign miss-discovery can add additional trouble.

4. Discussion

CR-WSNs vary from regular WSNs in numerous angles. When we execute the cognitive radio in NS2 programming, we get some recreation results that portray the improvement of CRWSN from the conventional WSN; we run the reenactment. If a way is gotten, at that point, information is sent. Else backup way to go is picked as appeared in depictions underneath. Figure 3, 4 shows a course of action of nodes and transmission of information in a network. Figure 5-11 shows the packet delivery ratio, which is high contrasted with past conventions. Figure 5 shows the energy consumption diagram, which offers less energy is burned-through than past works.

Performance Metrics

Performance metrics manage and controls network resources by setting priorities for specific types of data on the network. The various performance metrics include Delay, average energy consumption, goodput, Jitter, packet loss rate, packet delivery ratio, and throughput.

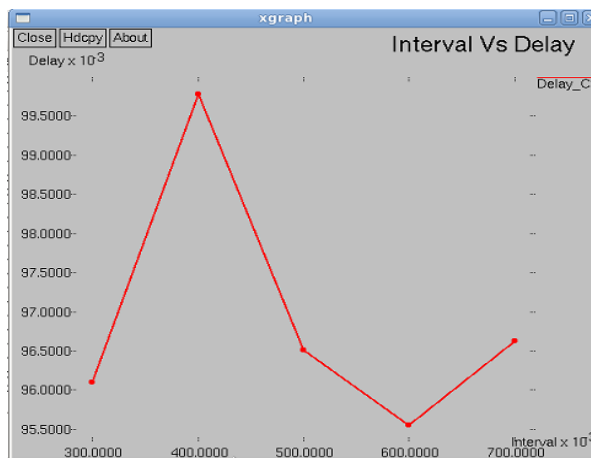


Figure 5: Data Transmission Delay Chart.

Figure 5 shows the interval vs. Delay in the Data transmission chart. In X-Axis denotes the Interval time, Y-Axis denotes the Transmission Delay

It refers to the data packet's standard time to be broadcasted from source to destination across a network. It also includes the Delay caused by the route detection process and the queue in data packet broadcast. Only the data packets that effectively transported to destinations that calculated and given by the formula

$$\text{End to End Delay} = \sum_i (\text{Arrive time} - \text{Send time}) / \sum_i \text{Number of Connections}$$

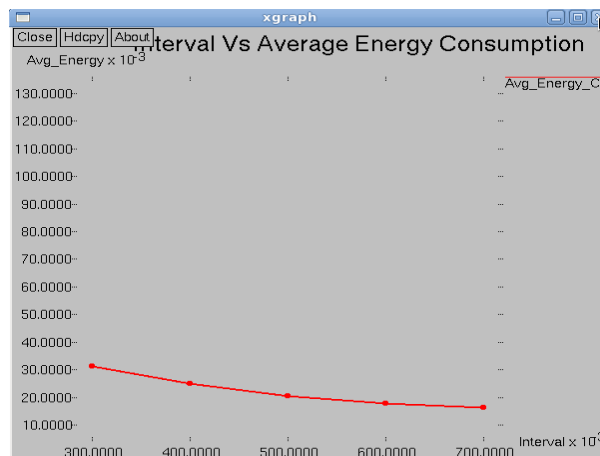


Figure 6: Average Energy Consumption Chart.

Figure 6 denotes the Energy Computation chart when using the proposed system. The X-Axis represents the Interval time, and Y-Axis indicates the Average Energy Level.

Energy consumption is an important concern for wireless networks as it has an inevitable impact on the network's performance. The energy required to send data depends on the distance between the nodes and the number of bits transmitted. The life needed for receiving also depends on the number of bits being received.

$$\text{Average Energy Consumed (AEC)} = \sum_{i=1}^N [E_i - E_r] \text{ Number of Nodes}$$

Where N be the set of all nodes present in the network $N = \{n_1, n_2, n_3, \dots\}$, E_i = Initial Energy of a node and E_r = Residual Energy of a node

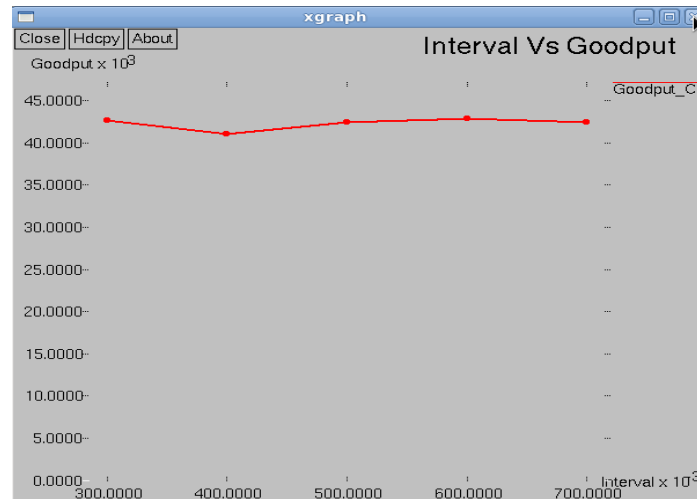


Figure 7: Goodput Chart.

Figure 7 denotes the number of useful information bits delivered by the network to a certain destination per unit of time. The X-axis represents the interval time; the y-axis indicates the goodput.

Goodput is calculated by dividing the size of a transmitted data by the time it takes to transfer the data.

$$\text{Goodput} = \frac{\text{size of the transmitted data}}{\text{total transfer time}}$$

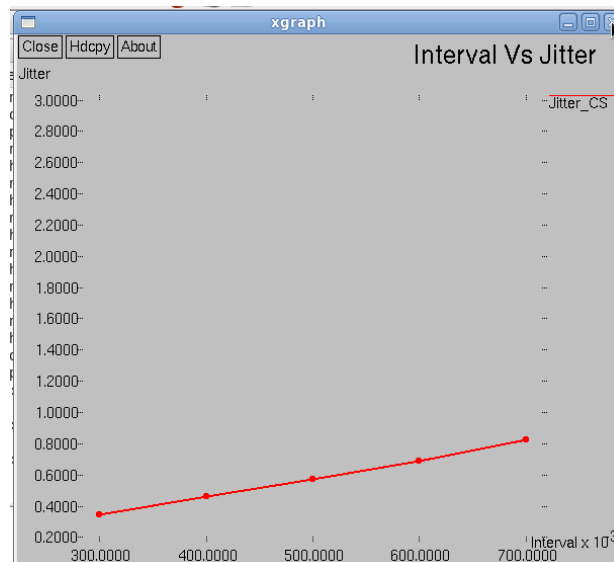


Figure 8: Jitter outcome Chart.

Figure 8 denotes the Jitter as the varied Delay between received packets usually measured in milliseconds (ms). It is the average of the time difference between each packet sequence. Jitter = Average of the time difference between packets in the sequence

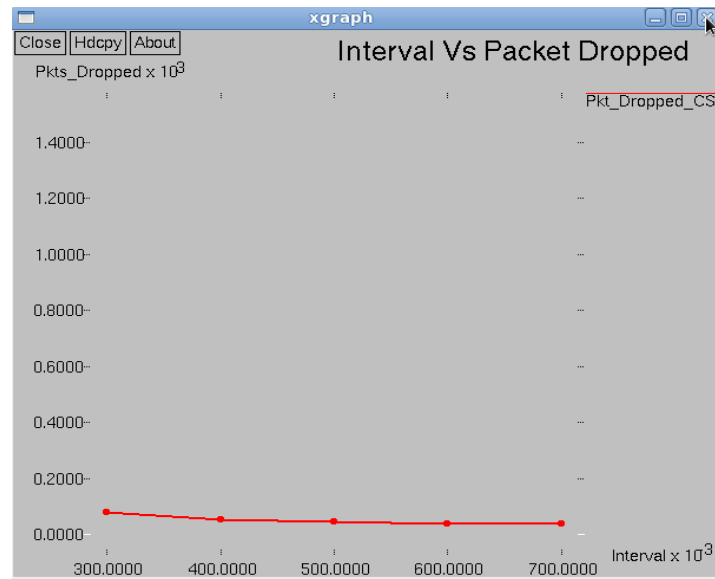


Figure 9: Packet Dropped Result.

Figure 9 denotes. Packet loss occurs when one or more data packets traveling across a computer network fail to reach their destination. The packet loss ratio represents the number of lost packets' ratio to the total number of sent packets.

Each packet has a deadline before it must be executed, and if this is not possible, the scheduler tries to minimize the number of lost packets due to deadline expiry.

$$\text{packet loss ratio} = \frac{\text{number of lost packets}}{\text{total number of packets sent}}$$

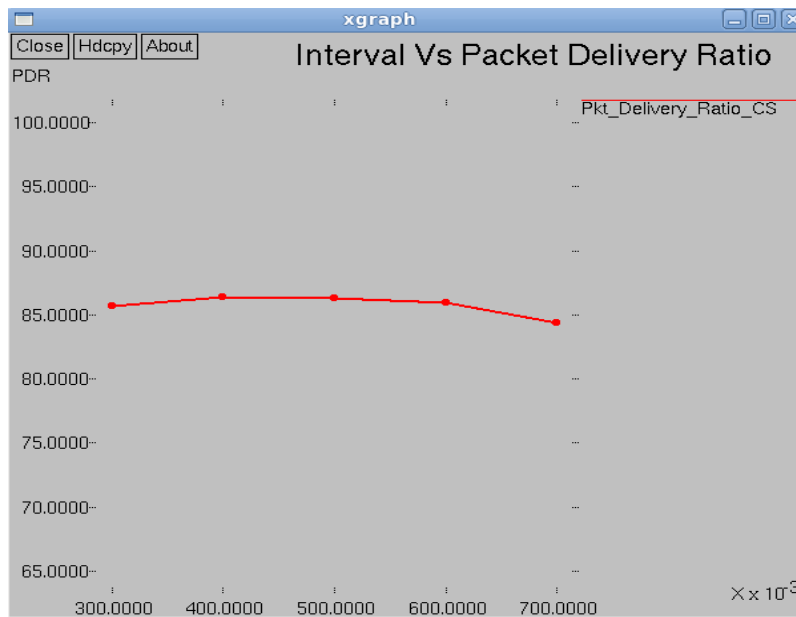


Figure 10: Packet Delivery Ratio Chart.

Figure 10 illustrates the packet delivery ratio obtained from the total number of data packets arrived at destinations divided by the total data packets sent from sources.

In other words, the Packet delivery ratio is the ratio of the number of packets received at the destination to the number of packets sent from the source.

$$\text{packet delivery ratio} = \frac{\text{total number of packets received}}{\text{total number of packets sent}}$$

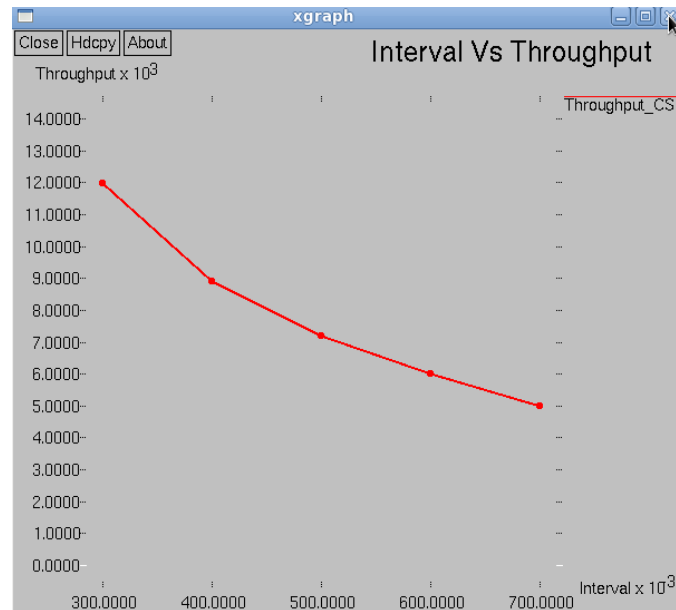


Figure 11: Throughput Chart.

Figure 11 denotes throughput is the key quality-of-service metric in wireless networks. It is the average speed of data transfer during a data connection. It is defined as the average number of bits sent or received per data request to the average duration of data transfer

$$\text{throughput} = \frac{\text{average number of bits sent or recived per data request}}{\text{average duration of data transfer per data request}}$$

5. Conclusions

This paper examines the impact of energy-productive information transmission systems on the presentation of energy-efficiency in CRWSN. The thought is to diminish the traffic load on the nodes with low energy by expanding the nodes' information transmission with higher public power. Reenactment results demonstrate that utilizing the proposed coverset prediction can improve the network execution when an EECSP algorithm is applied to a level network and energy-based grouping as it has appeared, transmission power control helps to improve the packet delivery ratio by over 15% while somewhat accomplishing lower bounce tally from source to the objective. To improve the data transmission power control approach considering distinctive spread models for additional dissecting more realistic situations, better network end-to-end performance is achieved.

References

1. Agarkhed, J., & Joshi, N. (2016). A framework to integrate cognitive radio sensing with a wireless sensor network. 2016 Thirteenth International Conference on Wireless and Optical Communications Networks (WOCN). doi:10.1109/wocn.2016.7759895
2. Supraja, P., Gayathri, V. M., & Pitchai, R. (2018). Optimized neural network for spectrum prediction using genetic algorithm in cognitive radio networks. Cluster Computing. doi:10.1007/s10586-018-1978-5
3. Hasegawa, M., Hirai, H., Nagano, K., Harada, H., & Aihara, K. (2014). Optimization for Centralized and Decentralized Cognitive Radio Networks. Proceedings of the IEEE, 102(4), 574–584. doi:10.1109/jproc.2014.2306255
4. Tumuluru, V. K., Wang, P., & Niyato, D. (2010). Channel status prediction for cognitive radio networks. Wireless Communications and Mobile Computing, 12(10), 862–874. doi:10.1002/wcm.1017
5. Joshi, N., & Agarkhed, J. (2016). Path construction using cognitive radio sensing in wireless sensor network. 2016 International Conference on Circuit, Power and Computing Technologies (ICCPCT). doi:10.1109/iccpct.2016.7530316
6. LI, Z., & WANG, R. (2010). Secure coverage-preserving node scheduling scheme using energy prediction for wireless sensor networks. The Journal of China Universities of Posts and Telecommunications, 17(5), 100–108. doi:10.1016/s1005-8885(09)60514-8

7. Yoon Hyun Kim, Kyung Sun Lee, & Kim, J. Y. (2011). Efficient signal detection of cognitive radio systems for wireless ad-hoc networks. 2011 IEEE MTT-S International Microwave Workshop Series on Intelligent Radio for Future Personal Terminals. doi:10.1109/imws2.2011.6027192
8. Yang Liu, Zhiyong Feng, & Ping Zhang. (2010). A novel ARQ scheme based on Network Coding Theory in Cognitive Radio networks. 2010 IEEE International Conference on Wireless Information Technology and Systems. doi:10.1109/icwits.2010.5611991
9. Zareei, M., Vargas-Rosales, C., Hernandez, R. V., & Azpilicueta, El. (2019). Efficient Transmission Power Control for Energy-harvesting Cognitive Radio Sensor Network. 2019 IEEE 30th International Symposium on Personal, Indoor and Mobile Radio Communications (PIMRC Workshops). doi:10.1109/pimrcw.2019.8880825
10. Seema Dev Aksatha, D., Pugazendi, R., & Arul Pon Daniel, D., "Enhancing the performance of Efficient Energy Using Cognitive Radio for Wireless Networks", 2019, 2nd EAI International Conference on Bigdata Innovation for Sustainable Cognitive Computing, 143-153.
11. Seema Dev Aksatha, D., & Pugazendi, R., "A Fuzzy Logic System For Improving The Performance Of Qos In Energy Efficiency Using Cognitive Radio", International Journal of Management (IJM), Volume 11, Issue 10, October 2020, pp. 1325-1336