

Maximization Energy Efficiency by Novel LEACH-Based Protocol and Optimal Wireless Sensor Networks

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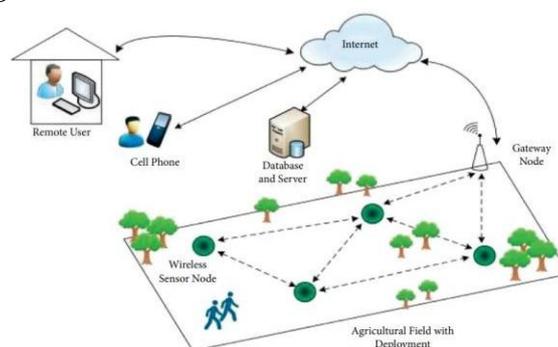
Abstract-Wireless sensor networks attract so much attention in current IoT-enabled industrial and domestic applications having either homogeneous or heterogeneous sensors deployed to acquire information of intent. WSNs are intended to work involving self-fueled sensor hubs as their decision of use is geographic basic. Such hubs should uphold energy effectiveness so that the network life span turns out to be high. Group head choice plays an essential stage in a WSN design which mostly centers around the minimization of organization energy utilization. It bunches sensor hubs so that a modern organization group is framed to have an improved lifetime other than a low powerutilization. A famous grouping procedure, known as LEACH and its variations, is viewed as energy effective contrasted with its partners. A novel fully connected energy-efficient clustering (FCEEC) mechanism using the electrostatic discharge algorithm to establish a fully connected network with shortest path routing from sensor nodes (SNs) to cluster head (CH) in a multihop environment. The proposedelectrostaticdischargealgorithm(ESDA)enhancesnetwork lifetimewhileattainingenergy-efficient full connectivity between sensor nodes. As a result of ESD, the dead node count is reduced significantly so that the network longevity is increased. In the end, simulation results exhibited improved performance metrics such as energy efficiency, dead node count, packet delivery, and network latency compared to a certain conventional CH selection approach.

I. Introduction

Wireless sensor networks (WSNs) showcase highly significant contributions across several applications such as environment monitoring, seismic control, agriculture management, security surveillance, furthermore, numerous other comparative regions. Lately, research on WSN has been fast developing for analysts given its one-of-a-kind qualities. WSN contains a few completely associated sensor hubs connected remotely. Each sensor hub is associated with gathering occasions of expectation, obtaining information, and helps in directing back to a base station (BS). Remote correspondence happens between various sensor hubs with the guide of a sensor hub the board framework to empower network checking and information assortment for every single explicit assignment, and afterward, information transmission is performed by interfacing all elaborate hubs through an expert hub to a close-by RF terminal [1]. WSN correspondence among BS and organization bunch is laid out utilizing the customary CSMA convention; information move across sensor hubs (SNs) and BS happens to utilize a group head (CH) as circuitous access [2]. A keystone in any of the sensor hub models is the lifetime augmentation of the WSN. At the point when the hub is depleted, as a rule, it can't be re-energized or supplanted by batteries. Situating of hubs in

conventional WSN proficiency is lesser because of the power interest and intricacy. The synchronization process among hubs and CH is limited by a few organization requirements that make the group head choice interaction dreary on account of all predominant bunching calculations. Henceforth, different methodologies are associated with choices for the CH. One of the navigations is completed by the assessment of the energy. As a general rule, a CH is picked given setting up a base energy measure utilizing energy edges to guarantee full network and dependability of the given WSN. WSNs are like the regular specially appointed sort of organizations containing many hubs relying on the organization scale, interconnected with separate CHs to detect the boundaries of expectation, and associated with information procurement and sending to BS which further transmits to resulting RF focuses. Fundamental difficulties in a network the executives are tied in with outlining an organization region alongside specific quality worries like adaptability, dependability, and asset the board [3]. Limitations are taken from specially appointed networks and are remembered for the fruitful WSN geography of the executives. From the new exploration discoveries, it is apparent that organization bunching is the generally well-known geography procedure working using the gathering of hubs to impel the CH and hubs for additional administration of planned assignments. Bunching strategies do primarily focus on proficient energy utilization to achieve sturdy and solid organizations. Information accumulation and grouping processes really do principally zero in on decreasing generally speaking organization energy utilization by wiping out a specific measure of transmission information and henceforth expanding network versatility and lifetime. Different calculations, for example, counterfeit brain organizations, support learning, and a multitude of knowledge help in the decrease of send information size utilizing a portion of the distributive qualities of the organization. In light of a legitimate concern for holding command over the unique idea of organizations, effective calculations should be utilized to convey a dependable and productive sensor organization. A large number of the examination works fused AI strategies to dispose of excess information being sent. These strategies deliver different functional answers for exploit assets used for delaying the lifetime of the whole organization. In area-driven approaches, the CH choice is ordinarily performed by picking an advantageous hub near a beneficial area. CH choice in light of a regular area includes computational intricacy while finding a reasonable sensor hub, consequently prompting unfortunate choice exactness and copied hub determination.

Figure 1: Structure of wireless sensor network



The energy expenses in a WSN comprises of firstly sensing the intended parameter and secondly connecting the data to BS. The fact is that WSNs consume maximum energy during data transmission over the sensing and processing stages which lead to draining of power sources

rapidly, leaving many nodes to die at a fast pace, and as a result, reducing the network life expectancy. However, the network fails when even a single node in the chain fails to involve in the data transmission process with its head node. Hence, it is firmly ascertained that WSNs are hypersensitive and highly vulnerable to working energy than any of the standard wireless networks. In the case of direct transmission from the sensor to BS, the sensor nodes get easily exhausted and become dead soon which is not a good sign in a WSN. The life span of a network shall be enhanced by employing power-efficient clustering architecture while carefully choosing sensor quality, network area, and the number of WSN nodes. It is concluded that WSN clusters do support careful power consumption and efficient energy utility with an added advantage of network longevity and packet delivery. Altogether, this approach helps to enable the remote data acquisition of physical processes from a global geometry and make that available in the Internet cloud through popular IoT technology to take deep leaps of wireless connectivity across men and machines as artificial intelligence (AI) initiative.

II. System Model

It is clear from different writing sources that remote sensor network engineering centers around accomplishing beneficial energy productivity while cautious administration is completed on limitations, specifically, heterogeneity, versatility, energy saving, life length, and so on. A very much positioned gathering of hubs is algorithmically coordinated to shape a bunch to raise the energy effectiveness of remote sensor organizations. Here, the key concluding boundaries are the nature of administration, load adjusting, and energy minimization. Every one of the hubs associates with a group head which then, at that point, gathers and passes sensors' information on to the objective. CHS introduces transmission through different hubs in the organization are individuals that immediate information from the hub to BS remotely. In the bunching approach, any data sent through asset obliged hubs straightforwardly causes energy consumption, failure, and impedance. A portion of the examination works is as of now completed in light of the study of WSN grouping. A few bunching strategies like HEED, BEENISH, FLOC, and LEACH are proposed with expected assortments and expansions thinking about equivalent and inconsistent groups. A superior LEACH convention [7] known as LEACH-Impt is accounted for in the writing which shows preferable execution over customary strategies. Drain Impt contains different disjoint ways inside the bunch hubs as a constituent of the directing geography. An ideal way is chosen to dispense with power utilization and lingering energy while the steering jumps across hubs. The proposed LEACH-Impt has less energy utilization than its standard partners. Arbitrary CH determination is one disadvantage of this approach causing misfortunes in gathered data and thus communicated. A productive fluffy rationale for CH determination [8] has been proposed to pick the best group head (CH) which uses capable coordination between the hubs in WSN by joining the fluffy deduction framework. In the proposed FBECS, the likelihood is conveyed to each hub of the organization concerning the distance-related utilizing fluffy rationale. The execution proportions of FBECS show better burden adjusting, dependability, and expanded lifetime. Another CH choice is performed by a bunch chain weight measurements (CCWM) approach given the positioning system [9]. The hub with a higher worth of position measurements inside its organization range is viewed as CH. It is presumed that the proposed CCWM technique offered adaptable weight factor changing qualities which upgraded the organization throughput.

III. Proposed LEACH Technique

The acronym LEACH denotes low energy adaptive clustering hierarchy. Currently, one of the well-proven network routing and clustering protocols operates with the time division multiple access (TDMA) protocol. The fundamental objective of this protocol is to conserve energy during the active performance of a sensor network. Generally, WSN is a collection of sensor nodes that are interconnected in a particular manner, providing a strong impact on many monitoring applications in daily life. The disparities found in energy consumption in an ongoing transmission certainly drain the sensor battery. LEACH uses the distributed algorithm to organize the sensor nodes into a cluster. Every cluster identifies its CH which establishes a transmission link to BS (sink node). The basic architecture of LEACH is shown in Figure 2. In LEACH, the CH nodes aggregated data received from clustered nodes, accumulate, and forward to the sink. It is proficient in the self-organizing capability of the cluster and by its adaptability in nature. The LEACH protocol is recognized by a round concept. For each round, there will be a new CH to initialize data transmission. Each round of the LEACH protocol involves two main phases, namely, the setup and steady-state phases, respectively. The setup phase contains CH selection and formation of cluster stages whereas only the transmission of data takes place in the steady-state phase. To select a typical head node, all nodes in the network broadcast their probability at the beginning of every round. The formula to calculate CH to BS distance is shown as follows:

$$d_{BS} = \sqrt{(x - 100)^2 + (y - 100)^2}$$

where x and y form the BS position in the search space. The optimum number of clusters is calculated using cluster optimization techniques as follows:

$$k_{opt} = \sqrt{\frac{T_n}{2\pi} \times \frac{\sqrt{\text{free space energy}}}{\text{multipath energy}} \times \frac{L}{d_{BS}^2 \rho}}$$

where T_n is the total number of nodes, and L is the network length/width. Initially, CH is selected based on the available energy level. CH will be sending the advertising message to all its member nodes on the CSMA mode. Upon receiving signal strength information from one CH node, the remaining nodes start to determine a new CH to lead the upcoming iteration. Then, the CH node schedules TDMA slots for data packet transmission and coordination within the cluster. As per the TDMA slots, sensor nodes update concerned data to the destined CH which, in turn, aggregates all nodes' information. In the uplink process, CH to BS communication is carried out using spread spectrum modulation. A unique spreading code is used by every cluster to communicate with BS to avoid interclass interference.

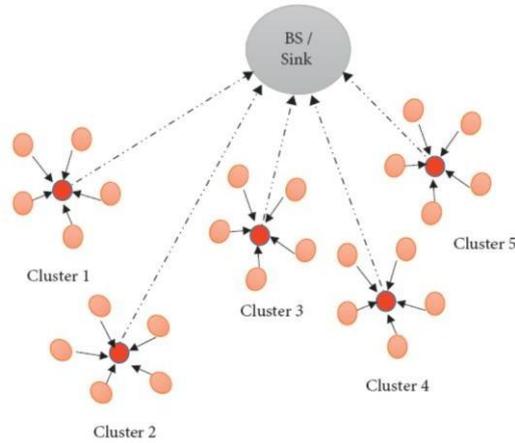


Figure2: The basic architecture of LEACH.

After completion of data transfer in round one, the network enters into the setup phase initializing CH for the next round and following its routine iterations. Figure 3. illustrates the LEACH process flow. Apart from its advantages such as energy efficiency and network life, it maximizes network coverage with minimum latency. The success rate of LEACH is determined by the following criteria:

- (i) Distance between CH to BS decides the number of hops and energy dissipation
- (ii) Energy must be low for nearer nodes and high for farther nodes delivered across the BS from even the farthest node. Here are the sequential flows of the ESD algorithm.

Step 1: firstly, we initialize with a random size of the object denoted as “Obj_Size,” i.e., the total number of electrical equipment in the design space

- (i) Position of nodes decides the fitness value. If the fitness value is more, the equipment is safe from ESD
- (ii) Besides, each piece of equipment contains a counter to account for the maximum number of attacks. This is called the initialization stage

Step 2: Secondly, the initialization process is repeated “Max_Iter” times to find a solution for the identified optimization problem

- (i) Three objects (source, load, and victim) are randomly identified in every iteration, and the best is kept at first
- (ii) “n1” random number is generated. If $n1 > 0.5$, only two objects are involved; otherwise, all the three do take part
- (iii) In the case of two objects, if the least fitness object moves towards the best fitness one (object 2 to object 1), it is represented as

$$p_{3_update} = p_2 + 2\alpha_1 p_1 - p_2 \quad (3)$$

in which

- (iv) p_{2_update} is the updated position of object 2. p_1 and p_2 are the past locations of both objects. α_1 is an arbitrary number with mean and SD values 0.7 and 0.2, respectively
- (v) In this case, as object 2 gets closer to object 1, an ESD affects object 2 (victim) is called direct ESD incidence in cases where $n1 < 0.5$, three elements participate to cause an ESD. Assuming that the third object moves towards the other two elements, then

$$p_{3_update} = p_3 + 2\alpha_2 p_1 - p_3 + 2\alpha_3 p_2 p_3 \quad (4)$$

IV. Results and Discussion

A typical WSN is constructed using the FCEEC algorithm to validate the energy efficiency and packet delivery across CH and BS. The simulation results are compared with LEACH, LEACH-C, BO-LEACH, and ESD algorithms. The authors have considered the following design parameters which are listed in Table 1. Simulation results are described concerning energy retention, dead node count, packet delivery, and network latency. The variation in the total energy of all the nodes after every round is shown in Figure 3. It is observed that after 500.

Table 1: Design parameters.

Number of nodes	100
Network size	100×100 square units
Max. energy of each node	50 mJ
Network topology	Arbitrary
Energy of transmit bit (Et)	50 nJ
Energy of receive bit (Er)	50 nJ
Energy at free space	10 pJ

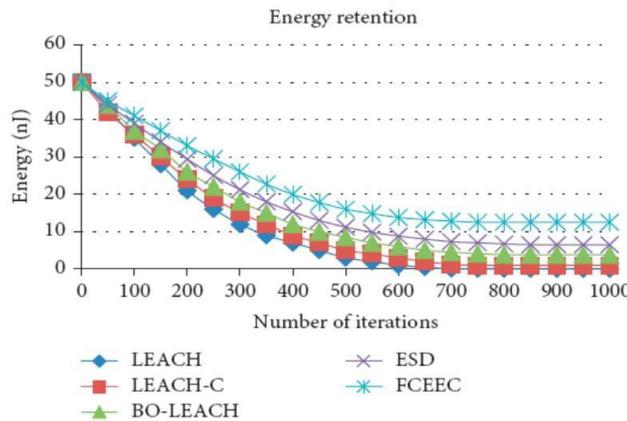


Figure 3: Energy retention.

rounds of execution, the proposed FCEEC increases total energy savings by 81.25% than LEACH, 68.75% than C-LEACH, 46.87% than BO-LEACH, and 30% than ESD. The energy retention further rises to a maximum of 96% compared with LEACH, 92% than C-LEACH, 69.6% than BO-LEACH, and 48% than ESD when the number of rounds reaches 1,000. Energy retention analysis after 500 and 1,000 rounds of iterations for various algorithms is shown in Figure 4. The dead node count after every round of iteration is shown in Figure 5. It is identified that after 500 rounds of execution, the proposed FCEEC reduces the dead node count by 55.5% from LEACH, 46.15% from C-LEACH, 33.3% from BO-LEACH, and 20% from ESD. The FCC reduces the dead node count by 25.8% from LEACH, 20.69% from C-LEACH, 16.86% from BO-LEACH, and 6.75% from ESD after 1,000 rounds of execution. The number of dead nodes after 500 and 1,000 rounds of iterations for various algorithms is shown in Figure 8. The data packet delivery from CH to BS after every round of execution is shown in Figure 9. After 500 rounds of execution in the proposed FCEEC, the packet delivery is raised by 33.52% than LEACH, 28.25% than C-LEACH, 23.65% than BO-LEACH, and 13.78% than ESD. The proposed FCEEC improves the packet delivery by 32.28% than LEACH,

28.24% than C-LEACH, 24.71% than BO-LEACH, and 17.13% than ESD after 1,000 rounds of execution. The number of data packets generated will be reduced if the number of dead nodes increases. In LEACH, more than 80 nodes will not be alive after 650 rounds of iteration. Hence, the number of packets generated will be very less compared

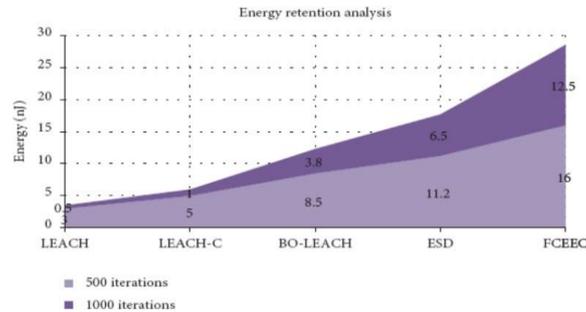


Figure4:Energyretention analysis.

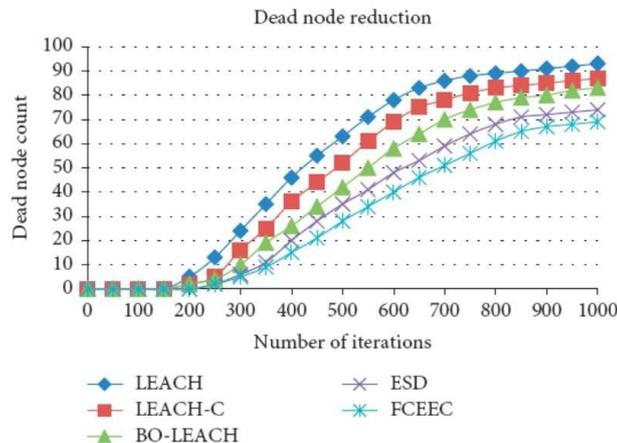


Figure5:Deadnode reduction.

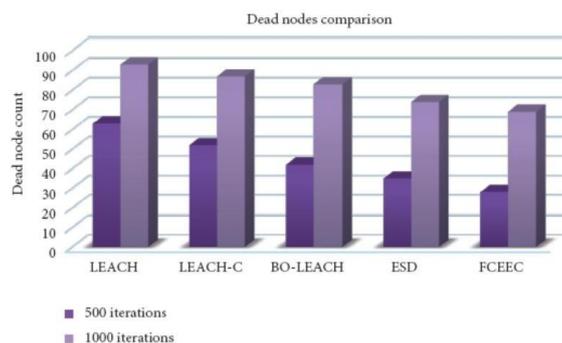


Figure6:Deadnode comparison.

with earlier rounds. The same scenario occurs when using other algorithms also. The packet delivery comparison after 500 and 1,000 rounds of iterations.

V. Conclusion

Thus, it is observed from the above results that the ESDA-based FCEEC algorithm facilitates optimum CH-BS placement and the shortest path discovery for full connectivity of nodes. The proposed method improves the packet delivery rate, and most importantly, the energy efficiency of nodes is increased significantly when compared with the generic LEACH and other conventional methods. Hence, it is concluded that the newly inducted FCEEC results in the optimization of WSN output parameters in terms of reduction in node energy by 96%, reduction of dead nodes by 25.8%, increase in the packet delivery rate by 32.28%, and the network latency by 66.46%, respectively.

References

- [1] D. Puccinelli and M. Haenggi, "Wireless sensor networks: applications and challenges of ubiquitous sensing," *IEEE Circuits and Systems Magazine*, vol. 5, no. 3, pp. 19–31, 2005.
- [2] M. Li, Z. Li, and A. V. Vasilakos, "A survey on topology control in wireless sensor networks: taxonomy, comparative study, and open issues," *Proceedings of the IEEE*, vol. 101, no. 12, pp. 2538–2557, 2013.
- [3] C. Li, J. Hailing, M. Yong, L. Tian, L. Wei, and Z. Ze, "Overview of wireless sensor networks," *Journal of Computer Research and Development*, vol. 42, no. 1, 2005.
- [4] M. M. Afsar and M.-H. Tayarani, "Clustering in sensor networks: a literature survey," *Journal of Network and Computer Applications*, vol. 46, pp. 198–226, 2014.
- [5] B. P. Deosarkar, N. S. Yadav, and R. P. Yadav, "Clusterhead selection in clustering algorithms for wireless sensor networks: a survey," in *Proceedings of the International Conference on Computing, Communication, and Networking*, pp. 1–8, Kaur, India, 2008.
- [6] L. Buttyan and P. Schaffer, "Position-based aggregator node election in wireless sensor networks," *International Journal of Distributed Sensor Networks*, vol. 6, no. 1, Article ID 679205, 2010.
- [7] Z. X. Wang, M. Zhang, X. Gao, W. Wang, and X. Li, "A clustering WSN routing protocol based on node energy and multipath," *Cluster Computing*, vol. 22, no. 3, pp. 5811–5823, 2019.
- [8] P. S. Mehra, M. N. Doja, and B. Alam, "Fuzzy based enhanced cluster head selection (FBECS) for WSN," *Journal of King Saud University Science*, vol. 32, no. 1, pp. 390–401, 2020.
- [9] S. Mahajan, J. Malhotra, and S. Sharma, "An energy balanced QoS based cluster head selection strategy for WSN," *Egyptian Informatics Journal*, vol. 15, no. 3, pp. 189–199, 2014.
- [10] M. A. Song and C. L. Zhao, "Unequal clustering algorithm for WSN based on fuzzy logic and improved ACO," *Journal of China Universities of Posts and Telecommunications*, vol. 18, no. 6, pp. 89–97, 2011.
- [11] S. Kaur and R. Mahajan, "Hybrid meta-heuristic optimization-based energy efficient protocol for wireless sensor networks," *Egyptian Informatics Journal*, vol. 19, no. 3, pp. 145–150, 2018.
- [12] P. Maheshwari, A. K. Sharma, and K. Verma, "Energy-efficient cluster-based routing protocol for WSN using butterfly optimization algorithm and ant colony optimization," *Ad Hoc Networks*, vol. 110, Article ID 102317, 2021.
- [13] G. K. Nigam and C. Dabas, "ESO-LEACH: PSO based energy-efficient clustering in LEACH," *Journal of King Saud University-Computer and Information Sciences*, vol. 33, no. 8, 2018.
- [14] A. Jari and A. Avokh, "PSO-based sink placement and load-balanced anycast routing in multi-sink WSNs considering compressive sensing theory," *Engineering Applications of Artificial Intelligence*, vol. 100, Article ID 104164, 2021.

- [15] B. Rambabu, A. V. Reddy, and S. Janakiraman, “Hybrid artificial bee colony and monarchy butterfly optimization algorithm (HABC-MBOA)-based cluster head selection for WSNs,” *Journal of King Saud University-Computer and Information Sciences*, 2019.
- [16] T. Sood and K. Sharma, “LUET: a novel lines-of-uniformity based clustering protocol for heterogeneous-WSN for multiple-applications,” *Journal of King Saud University-Computer and Information Sciences*, 2020.
- [17] P. Sivakumar and M. Radhika, “Performance analysis of leach-ga over leach and leach-c in the win,” *Procedia Computer Science*, vol. 125, pp. 248–256, 2018.
- [18] N. R. Maliseti and V. K. Pamula, “Performance of quasi oppositional butterfly optimization algorithm for cluster head selection in WSNs,” *Procedia Computer Science*, vol. 171, pp. 1953–1960, 2020.
- [19] R. Yarinezhad and S. N. Hashemi, “Solving the load-balanced clustering and routing problems in WSNs with an FPT-approximation algorithm and a grid structure,” *Pervasive and Mobile Computing*, vol. 58, Article ID 101033, 2019.
- [20] H. Yu and W. Xiaohui, “PSO-based energy-balanced double cluster-heads clustering routing for wireless sensor networks,” *Procedia Engineering*, vol. 15, pp. 3073–3077, 2011. [21] V. S. Devi, T. Ravi, and S. B. Priya, “Cluster-based data aggregation scheme for latency and packet loss reduction in WSN,” *Computer Communications*, vol. 149, pp. 36–43, 2020.