An Enhanced Distributed Energy-Efficient Clustering (DEEC) Protocol for Wireless Sensor Networks

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Abstract

Wireless Sensor Networks (WSNs) is a warm area of research. WSNs consist of large number of sensor nodes deployed randomly in the sensor field. But, it suffers with several shortcomings such as energy imbalance, processing power, storage, transmission range and mitigates energy hole problems. Therefore it’s a tough task to design and develop an optimized routing protocol for WSNs. Hence, this research paper mainly focuses on the energy consumption and network lifespan issues of WSNs and presents an improved version of Distributed Energy-Efficient Clustering (DEEC) protocol. To enhance the network lifetime and reliability of DEEC protocol, three improvements are incorporated in the algorithm and the proposed algorithm named as improved DEEC (IDEEC). From experiment section, it is stated that proposed improvements make protocol more robust and efficient in terms of lifetime of nodes, stability and energy consumption in comparison to other algorithm being compared.

Keywords: DEEC, LEACH, WSNs and Clustering

1. Introduction

For past several years, there is a need for a medium through which entire source of information can be transmitted from the place where it is needed and the received information should be precise to give an absolute and accurate picture of the real world. Hence, a new category of networks has appeared and it is known as wireless sensor networks (WSNs). WSN contains large number of nodes, which collects the information from its surrounding environment. It can also describe as a network of sensing nodes with power facility. These nodes have capability to sense atmospheric or physical parameters such as humidity, temperature, pressure from the surroundings and send all collected information to the central node. These nodes work together to achieve a common goal. In WSN, sensor nodes are used for sensing, processing and communication and it can be either fixed or mobile. There are number of constraint which affect the performance of WSNs, these are energy, processing power, storage and transmission range. Among these, energy is one of main constraint for WSNs. The energy of deployed sensor nodes gradually decreases in terms of data and distance. Lot of research work is going on to address the pitfalls of WSNs. Some of the recent research work in this direction is presented in [1]. In WSNs, sensors are densely deployed for collecting the data and information and it can be widely applicable to different domains such as agriculture, forests, coal mines, monitoring of rail tunnels, monitoring of solar photovoltaic cell in a grid, etc., and collect the data from all locations using a centralized Base Station (BS) with the help of cluster heads (CHs) [2–3]. In such networks, data is collected periodically by the BS.

In Literature, it is found that clustering with hierarchical topology is suitable for continuous monitoring of networks [4]. Further, it is also noticed that clustering improves
the lifetime of data within network as compared to direct data transmission and it can be improved by a factor of nearly 2 to 3 times [5]. Clustering protocols have proven its advantage in data gathering networks. These protocols also enable data aggregation at cluster head (CH), it decreases transmitted data packets and reduces consumption of energy at sensor nodes [6]. These protocols are communicated in two ways, intra-cluster communication and inter-cluster communication. In WSN, communication can be either single hop or multi-hop [7]. Most of clustering protocols use single hop communication for communicating within clusters such as LEACH [8], LEACH-DT [9], HEED [10], etc. It is also reported that multi-hop communication is more energy-efficient (communication b/w sensor node to CH) than single hop communication. For example, in dense network, propagation loss exponent is high due to close deployment of sensor nodes and it results in inference and collision problems. Therefore, in such situation, multi-hop communication is more preferable or better than single hop as it successfully overcomes signal propagation difficulties [7]. Another thing in favor of multi-hop is the limitation of direct transmission. For a certain threshold distance, it is preferred to use direct transmission [11] but beyond the threshold, it consumes more energy like to be fourth power of distance [12]. Hence, to improve network scalability, multi-hop communication is more preferable [13]. If BS is located far away then, it is suggested that to use multi-hop communication [11]. Some of these are multi-hop LEACH [14], EADC [15] and EDUC [16].

In present scenario, research community focus on the energy imbalance and mitigate energy hole problem of WSNs. A lot of research has been carried out to handle these popular constraints. Hence, the motive of this research work is to improve network lifespan of DEEC protocol [17]. In DEEC, CHs are selected on the basis of probability. It can be computed as the ratio between residual energy of each node and the average energy of network. The main contribution of this research work is as follows:

- To introduce an improved scaling factor parameter to minimize the energy of nodes within cluster.
- To propose a new modified Equation to compute threshold probability more efficiently.
- To extend network life span by using neighborhood information concept.

2. Related Work

This section describes related work in the direction of energy efficient protocols especially designed for WSNs. To improve the scalability and energy saving in heterogeneous wireless sensor networks, Heinzelman et al. [8] have developed a clustering based protocol for WSNs known as LEACH protocol. The LEACH was developed to select CHs randomly for transmission of the energy load from CHs to BS. It uses the concept of single-hop communication between the sensor nodes and BS. However, the main shortcoming of this protocol is that it is not suitable for large-scale networks. Hence, lot of improvements has been made in LEACH to make it more efficient and powerful, such as LEACH-DT [9], M-LEACH [1]. Younis and Fahmy [10] have introduced a distributed energy efficient protocol for Ad-hoc sensor networks, known as HEED protocol. In this protocol, residual energy is used to elect the CHs and selection of these CHs is done periodically. This protocol uses multi-hop communication. The performance of the HEED protocol is compared with LEACH protocol and claimed that HEED protocol provides better results in terms of network lifetime. To reduce the energy consumption and to improve the scalability and lifetime of the network, Qing et al. [17], have presented another protocol for WSNs known as DEEC. In this protocol, CHs are selected based on residual energy and average energy of the network. It is also stated that simulation results of DEEC is more significant than any other algorithms being compared with. To prolong the network life time, Ye et al. [18] have presented an energy
efficient protocol called as EECH. In this algorithm CHs are selected by using the localized competition. Simulation results of the protocol compared with LEACH and HEED it was observed that the EECH protocol consumes energy more efficiently and significantly improves the lifetime. To improve network lifetime and reliability, Kumar et al. [19] have introduced an improved version of EECH. In this protocol, CHs are elected on the behalf of weighted probabilities of initial energy nodes. The results of the proposed protocol are compared with LEACH and EECH protocols. The author claimed that the protocol significantly handle the issues of Network life time and reliability. In continuation of their work, Kumar et al. [20] have developed another protocol to address the issues of network lifetime and energy efficiency known as EECDA. This protocol is based on the concept of energy efficient clustering routing and data aggregation. A new CHs selection and data communication mechanism were adopted for this algorithm. The experimental results were compared with LEACH, EECH and EDGA. From results, it is stated that EECDA protocol achieves better results both in terms of network lifetime and energy efficiency.

3. Distributed Energy- Efficient Clustering (DEEC) Algorithm

DEEC is a well-known energy efficient protocol for heterogeneous WSNs. In DEEC, the field is divided into different clusters. Each cluster contains some sensor nodes and has a CH. The work of the CH is to receive the information from sensor nodes within a cluster and send this information to BS. To make the node as Cluster Head (CH), a probability function is computed. This function is defined in terms of residual energy and average energy of networks. This function computes the ratio between residual energy of each node and the average energy of the network and it is computed for each node of a cluster. The node having high computation value in comparison to other nodes have better chance to be selected as a CH. In WSN, CHs are chosen periodically. A cluster head selection algorithm is adopted for this. The main work of CH is to collect the data within a cluster and send it to BS. It is also assumed that all nodes of WSN having different amount of energy initially and newly added nodes or energy-harvested nodes have more energy than the old ones. It is also noted that probability of CHs are also influenced by the ratio of residual energy of each node and the average energy of the network.

3.1. Heterogeneous Network Model

A network model of M×M (100 × 100) square region is considered for experimental purpose. It is noted that ‘n’ number of stationary nodes are uniformly dispersed in above-mentioned region. The BS is located in the center of square region and nodes consist of data, which is sent to the BS. Among these nodes, some nodes are selected as CHs and CHs are responsible to transmit the aggregated data to BS. The sensor nodes are categorized into two types-advanced nodes and normal nodes. The initial energy of normal nodes is denoted by \(E_0\) and ‘m’ denotes the fraction of advanced nodes. So in a WSN, total advanced nodes are \(mN\) and the energy associated with these nodes is \(E_a\) \((1-m)\). Whereas, total number of normal nodes is \((1-m)N\) and the energy associated with these nodes is \(E_n\). Hence, the total initial energy of the two-level heterogeneous networks is computed using Equation 1.

\[
E_{\text{total}} = N(1-m)E_0 + mNE_0(1+a) = NE_0(1+am)
\]  

(1)

3.2. Cluster Heads Selection Algorithm

This subsection describes the process of CHs selection. Assume the number of nodes disseminated in the region is “S”. A probability function is associated with each node in each round whether a node is elected as CH or not. The value of \(n_i\) differs in each node
because every node has different energy associated with it. Hence, the probability of a node to be a CH after \( n \) rounds can be calculated using Equation 2.

\[
P_i = \frac{1}{n_i} \quad (2)
\]

If nodes consist of same energy level in each round then \( P_i = P_{\text{opt}} \). If nodes consists of different amount of energy, then the value of \( P_i \) is larger for the nodes having more energy in comparison to nodes with low energy \( i.e. \ P_i > P_{\text{opt}} \). Then the average energy of network can be computed using Equation 3.

\[
\bar{E}(r) = \frac{1}{N} \sum_{i=1}^{N} E_i(r) \quad (3)
\]

Using it as reference energy, probability of \( i \)th node can be described using Equation 4 and the average number of clusters per epoch is identified using Equation 5.

\[
P_i = P_{\text{opt}} \left[ 1 - \frac{\bar{E}(r) - E_i(r)}{E(r)} \right] = \frac{P_{\text{opt}} E_i(r)}{E(r)} \quad (4)
\]

\[
\sum_{i=1}^{N} P_i = \sum_{i=1}^{N} \frac{P_{\text{opt}} E_i(r)}{E(r)} = \frac{P_{\text{opt}} \sum_{i=1}^{N} E_i(r)}{E(r)} = P_{\text{opt}} N \quad (5)
\]

The threshold probability \( T(s_i) \) decides whether \( S_i \) can become a CH or not, is computed using Equation 6.

\[
T(s_i) = \begin{cases} 
\frac{P_i}{1 - P_i \left( r \mod \frac{1}{E_i(r)} \right)} & \text{if } S_i \in G \\
0 & \text{otherwise} 
\end{cases} \quad (6)
\]

In each round, if \( S_i \) finds itself eligible for CH, a random number is selected in between 0 to 1. The selected number is compared with threshold value, if it is less, then the node acts as CH in that particular round. \( n_i \) is inverse of \( P_i \) and it is chosen on the basis of residual energy \( E_i(r) \) which is computed using Equation 7.

\[
n_i = \frac{1}{P_i} = \frac{P_{\text{opt}} E_i(r)}{E(r)} = \frac{E_i(r)}{E(r)} \quad (7)
\]

According to Equation 7, nodes with high residual energy have more chance to be elected as a CH. It also observes that at every epoch every node is busy in checking its candidature for cluster head. The energy dissipation can be minimized by reducing the number of calculations and it can be possible only by restricting the number of nodes. The nodes having more energy than a threshold value are considered to participate. Hence, a node \( S_i \) can compete for cluster head selection if \( E_i(r) > E_{\text{threshold}} \). In WSNs, initially the nodes have different amount of initial energy. The weighted probability for advanced (\( P_{\text{adv}} \)) and normal (\( P_{\text{norm}} \)) is computed using Equation 8.

\[
P_{\text{adv}} = \frac{P_{\text{opt}}}{1 + \alpha m}, \quad P_{\text{norm}} = \frac{P_{\text{opt}}(1 + \alpha)}{1 + \alpha m} \quad (8)
\]

Thus the \( P_i \) can be described using Equation 9.

\[
P_i = \begin{cases} 
P_{\text{opt}} E_i(r) & \text{if } S_i \text{is the normal node} \\
\frac{P_{\text{opt}}(1 + \alpha)E_i(r)}{(1 + \alpha m)E(r)} & \text{if } S_i \text{is advanced node} 
\end{cases} \quad (9)
\]

Further, it can extend to multi-level heterogeneous networks and the weighted probability is calculated using Equation 10, whereas, \( P_i \) can be computed using Equation 11.

\[
P(s_i) = \frac{P_{\text{opt}} N(1 + \alpha_i)}{N + \sum_{i+1}^{N} \alpha_i} \quad (10)
\]


\[ P_i = \left( \frac{P_{opt} N (1 + \alpha) E(r)}{(N + \sum_{i=1}^{N} \alpha_i E(r))} \right) \]

(11)

3.3. Energy Consumption

In DEEC, all nodes share the information about total energy and life time of network. The average probability \( P_i \) is computed using average energy \( \bar{E}(r) \) and total energy \( E_{\text{total}} \) of network. The life time value \( R \) is given by the BS. The average energy of network is calculated using Equation 12.

\[ \bar{E}(r) = \frac{1}{N} E_{\text{total}} \left( 1 - \frac{r}{R} \right) \]

(12)

Here, \( R \) represents the network life time, and it can be computed using Equation 13.

\[ R = \frac{E_{\text{total}}}{E_{\text{round}}} \]

(13)

The energy expansion is computed using Equation 14 and it is defined in terms of radio transmission of 1-bit message and distance \( d \).

\[ E_{\text{tx}}(l, d) = \begin{cases} E_{\text{elec}} + l e_{fs} d^2, & d < d_0 \\ E_{\text{elec}} + l e_{mp} d^4, & d \geq d_0 \end{cases} \]

(14)

Where \( E_{\text{elec}} \) denotes energy dissipated to run the transmitter on the receiver circuit, and \( e_{fs} d^2 \) and \( e_{mp} d^4 \) is energy dissipated in free environment and multipath path loss. Each node sends \( L \) bits of data to CH in a single iteration. Hence energy dissipated in a network is computed using Equation 15.

\[ E_{\text{round}} = L \left( 2N E_{\text{elec}} + N E_{\text{DA}} + k e_{mp} d^4_{\text{CHtoBS}} + N e_{fs} d^2_{\text{NoCH}} \right) \]

(15)

Where, \( k \) defines no. of clusters, \( E_{\text{DA}} \) denotes data aggregation cost, \( d_{\text{CHtoBS}} \) denotes average distance of CH and BS and \( d_{\text{NoCH}} \) is the average distance of nodes and CH. \( d_{\text{CHtoBS}} \) and \( d_{\text{NoCH}} \) is computed using Equation 16.

\[ d_{\text{toCH}} = \frac{M}{\sqrt{2} \pi k}, \quad d_{\text{trans}} = 0.765 \frac{M}{2} \]

(16)

4. Proposed Improvements in DEEC Algorithm

This section describes the proposed improvements in the DEEC algorithm. In this work, three improvements are incorporated in DEEC protocol for enhancing its performance in terms of energy and packet lifetime. These improvements are discussed as follows.

4.1. Scaling Factor

To minimize the energy of nodes within a cluster, a scaling factor parameter is introduced. It assumes that the nodes are capable of directly communicating to the base station located a farther distance, thus when intra-cluster communication is to be conducted it should be done at a reduced power level. For example if the network field is 100 x 100 in size and there are 10 clusters this means the size of one cluster is 10 x 10, thus the maximum power of intra cluster communication may be limited to a little more than 10 x 10 size. In other words the amplification power \( e_{fs} \) could be reduced by a factor of 10. Thus the scaling factor can be defined using Equation 17.

\[ \text{Scaling factor} = \left\{ \text{rand}() \times \frac{\text{Area of Network field}}{\text{area of the cluster} \times \text{no. of nodes in a cluster}} \right\} \]

(17)

4.2. Modified Threshold Probability

The threshold probability has significant impact on the selection of CHs. It determines whether a node can become a CH or not. A node can be acted, as a CH after nth epoch, is calculated using Equation 2. This Equation is used to determine number of nodes live
after nth epoch, but it cannot consider the energy of nodes that are live after nth epoch. Hence, a new modified Equation is proposed to determine the probability, whether a node becomes a CH or not which is described as follows.

\[
P_i = \frac{\text{Energy of the } i\text{th node}}{\sum_{i=1}^{n} \text{Total energy of all nodes within a cluster}} \quad (18)
\]

After computing the probability of \(i\)th node, it is the threshold probability, which recognize whether a node is capable to become CH or not. Hence, to improve the efficiency of the DEEC algorithm for CHs selection, total energy of all nodes within a cluster is considered for computing threshold probability and it can be defined using Equation 19.

\[
T(s_i) = \begin{cases} 
\frac{P_i}{1 - P_i \left( \frac{P_i}{\sum_{i=1}^{n} P_i} \right) r \mod n} & \text{if } S_i \in G \\
0 & \text{otherwise}
\end{cases} \quad (19)
\]

4.3. Neighborhood Information

The concept of neighborhood information is used to identify the next possible CHs. Therefore to compute the location of next CH, a distance-based measure is applied on the current CH location. So a threshold distance (\(d_0\)) is taken into account to find the neighbors of current CH. If energy of nodes within threshold distance is less than current CH energy, all these nodes are rejected, otherwise those nodes are selected as neighbor of current CH and CH stores all its neighbors. The objective of neighborhood information is to choose a CH with maximum energy among its neighbors. Hence, the nodes having greater energy than its CH become neighbors of current CH.

5. Experimental Results

This section describes the experimental results of proposed protocol. The proposed algorithm is implemented in MATLAB 2014 using corei5 processor and 4 GB RAM on window operating system. The performance parameters to test the effectiveness and efficiency of proposed protocol are stability period, lifetime and data packets. The results are also compared with other well-known existing algorithms like LEACH and DEEC. In our experiment, the WSNs consist of hundred nodes (N). The nodes are randomly scattered in 100x100m dimension field. It is assumed that all nodes are stationary. It is also considered that BS is placed at center of network field. For heterogeneous WSNs, the settings of radio parameters are shown in Table 1.

![Table 1. Parameter Setting of IDEEC Protocol](image)

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
<th>Parameter</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Network field</td>
<td>100m x 100m</td>
<td>(E_{DA})</td>
<td>5 nJ/ bit/ Message</td>
</tr>
<tr>
<td>Number of nodes</td>
<td>100</td>
<td>(d_0)</td>
<td>70 m</td>
</tr>
<tr>
<td>(E_{elec})</td>
<td>5 nJ/bit</td>
<td>Message Size</td>
<td>4000 bits</td>
</tr>
<tr>
<td>(E_{fs})</td>
<td>10 pJ/ bit/ m(^2)</td>
<td>(P_{opt})</td>
<td>0.1</td>
</tr>
<tr>
<td>(E_{amp})</td>
<td>0.0013 pJ/ bit/ m(^2)</td>
<td>(E_{Threshold})</td>
<td>(E_0/4)</td>
</tr>
<tr>
<td>(E_0)</td>
<td>0.5 J</td>
<td>Scale Factor</td>
<td>10</td>
</tr>
</tbody>
</table>

Book made by this file is ILLEGAL.
Figure 1. Uniform Distribution of Sensor Node in 100m x 100m Sensor Field with BS

Figure 2. Comparison of Dead Nodes of IDEEC, DEEC and LEACH Protocols

Figure 2 show the comparison of dead nodes in each iteration of IDEEC, DEEC and LEACH protocols over 4000 iterations. In first 1000 iterations, all nodes are live nodes. In DEEC and LEACH protocols, after 1200 iteration nodes become dead as entire energy is consumed. Whereas, in proposed IDEEC algorithm nodes become dead after 1500 iterations. In proposed IDEEC protocol, only 48 nodes are converted into dead nodes throughout its entire iterations. Whereas, in LEACH all nodes become dead nearly after 3100 iterations and in DEEC, only 8 nodes are live after 4000 iterations. From this, it can be concluded that the proposed protocol is more effective and efficient. Hence the proposed improvements make the DEEC protocol prolong in the direction of network lifetime. Figure 3 shows the comparison of packets sent to the BS using IDEEC, DEEC and LEACH protocols. From Figure 3, it is clearly observed that there is significant difference between the packets sent by IDEEC protocol and other protocol being compared.
Figure 3. Comparison of Packet Sent to BS of IDEEC, DEEC and LEACH Protocols

6. Conclusion

In this work, three improvements are integrated in DEEC protocol to make it more effective, efficient and stable for WSNs. Basically DEEC is adaptive energy efficient protocol to design the change in probability of sensor nodes for becoming a CH. To make the protocol more robust, capable and reliable, few modifications are proposed in this work. These modifications target the packet lifetime and energy consumption issues of DEEC. Hence, the concept of scaling parameter and neighborhood information are integrated in existing DEEC protocol. Among these, a modified threshold probability Equation is also proposed to select the CH in efficient way. The experimental results of the proposed protocol are compared with LEACH and DEEC protocols. Among these protocols, proposed protocol gives better and more efficient results.

References


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