

Proportional Fairness Based Energy Efficient Routing in Wireless Sensor Network

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Received: 05 July 2021; Accepted: 10 August 2021

Abstract: Wireless Sensor Network (WSN) is an independent device that comprises a discrete collection of Sensor Nodes (SN) to sense environmental positions, device monitoring, and collection of information. Due to limited energy resources available at SN, the primary issue is to present an energy-efficient framework and conserve the energy while constructing a route path along with each sensor node. However, many energy-efficient techniques focused drastically on energy harvesting and reduced energy consumption but failed to support energy-efficient routing with minimal energy consumption in WSN. This paper presents an energy-efficient routing system called Energy-aware Proportional Fairness Multi-user Routing (EPFMR) framework in WSN. EPFMR is deployed in the WSN environment using the instance time. The request time sent for the route discovery is the foremost step designed in the EPFMR framework to reduce the energy consumption rate. The proportional fairness routing in WSN selects the best route path for the packet flow based on the relationship between the periods of requests between different SN. Route path discovered for packet flow also measure energy on multi-user route path using the Greedy Instance Fair Method (GIFM). The GIFM in EPFMR develops node dependent energy-efficient localized route path, improving the throughput. The energy-aware framework maximizes the throughput rate and performs experimental evaluation on factors such as energy consumption rate during routing, Throughput, RST, node density and average energy per packet in WSN. The Route Searching Time (RST) is reduced using the Boltzmann Distribution (BD), and as a result, the energy is minimized on multi-user WSN. Finally, GIFM applies an instance time difference-based route searching on WSN to attain an optimal energy minimization system. Experimental analysis shows that the EPFMR framework can reduce the RST by 23.47% and improve the throughput by 6.79% compared with the *state-of-the-art* works.

Keywords: Wireless sensor network; proportional fairness; multi-user routing; energy consumption; greedy instance fair; Boltzmann distribution



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1 Introduction

Efficient route identification is one of the most promising approaches in WSN. For example, if a sensor node wants to send data packets to the destination_node (DESTN_Node) through the sink, efficient route identification reduces the energy consumption of nodes and improves the network's lifetime. Many papers have been published to address the route identification problem. In [1,2], balancing energy during sensing and communication was performed based on the distributed algorithm to balance energy allocation efficiently. A Store Carry and Forward (SCF) mechanism was designed by applying the mathematical formulation to identify the shortest route to Base Station (BS). In [3], a method was presented to achieve high energy savings using optimization mechanisms that improved the battery lifetime with cooperative game theory. The cooperative theory applied in [4] has also been shown to consume less energy during packet forwarding than the non-cooperative principle. A quorum-based energy efficient protocol to minimize collision and improve transmission rate was presented in [5].

In recent years, improving the network lifetime by providing efficient routing has generated interest in WSN [6]. The methods involved harvesting energy using dedicated radio frequency and explored the possibility of increasing the network lifetime, resulting in energy fairness. This has motivated researchers to develop energy fairness mechanisms [7], resulting in an improved energy consumption rate but compromising transmission power. The ways to improve data transmission have been done through energy-optimized mechanisms [8], wherein transmission is performed based on open and closed-loop parameters. However, the delay factor for data transmission was not concentrated.

One of the most popular techniques in WSN for efficient routing is the design of an effective means for clustering. Cluster-based routing was introduced in [9] based on the cluster's energy density to reduce delay. A representative path was determined [10] by generating an algorithm for clusters to improve node lifetime. A trust aware routing protocol was designed in [11] using a Bayesian network. In [12], opportunistic routing was designed to improve the number of transmissions using link correlation. Data integrity [13] was measured using a multipath dynamic routing algorithm to reduce the delay involved during routing. In [14], a Secure Quality of Service Routing and communication using Hierarchical Key Ranking in Mobile *Ad hoc* Network was presented. In [15], security groups have been formed, and the communication is done based on key mechanisms. In [16], Blockchain technology with its essential characteristics and applications are discussed. Particle Swarm Optimization algorithm [17] helps reduce SN's energy and prevent energy utilization by Low-energy adaptive clustering hierarchy protocol while routing the data sent to the destination or BS and results in data loss [18].

This paper develops an energy-efficient routing framework to support energy-efficient routing with minimal energy consumption in WSN. The following are the main contributions of this research: (a) A PFMR is developed with preference value to reduce the energy consumption during routing, and the results are supported by Network Simulation (NS-2); (b) the significance of considering interference rate in GIFM is demonstrated through simulation results with different node density; (c) an efficient design of BD considering the traversing time to reduce the RST between SN in WSN [19].

The remainder of this article is organized as follows. Section 2 presents the formulation of an energy-efficient routing framework called EPFMR in WSN and explains it with the help of the block diagram. Experimental results with parametric definitions are provided in Section 3. In Section 4, our proposal is compared to the other routing mechanisms. The paper concludes with a summary of our contributions in Section 5.

2 Energy-Aware Proportional Fairness Multi-User Routing Framework

In this section, we first described the principles of the PFMR, then the GIFM to improve the instantaneous throughput rate. Finally, the construction of BD is made efficiently to reduce the RST considerably in WSN. Fig. 1 shows the block diagram of the EPFMR method [20].

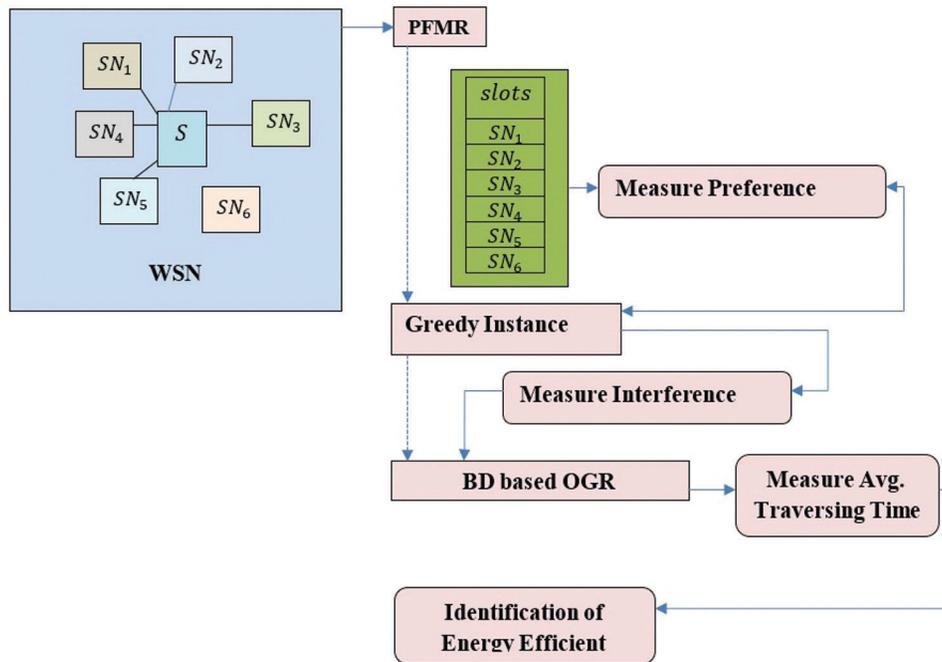


Figure 1: Proposed model of EPFMR

The EPFMR approach is divided into 3 sections to minimize the energy consumption levels during route discovery and improves throughput by minimizing interference during route discovery and RST [21]. PFMR is designed with slots for each SN, and the next SN is assigned to a specific slot based on the preference value [22]. This process is repeated until all the SNs are assigned to a slot. Next, the interference rate is identified using the GIFM. Finally, according to the traffic (i.e., node density) and average traversing time, the Optimum Global Route (OGR) is measured, resulting in an energy-efficient route. A detailed description of each part is described in the forthcoming sections [23].

2.1 Design of Proportional Fairness Multi-User Routing

PFMR is presented in the conceptual methodology to minimize the energy consumption levels in WSN. The PFMR is deployed in WSN with the help of request time and serves as the first step in the EPFMR framework. Packet flow is performed in an efficient manner using the relationship between the periods of requests. Fig. 2 shows the design of PFMR using the design of the EPFMR framework in WSN [24].

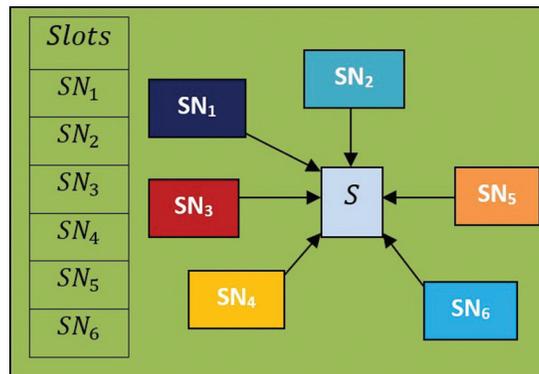


Figure 2: Design of PFMR in WSN

Let us consider a scenario with 'n' SN (i.e., multi-users). 'SN_i' in the network is placed randomly within the sink 'S' section. Therefore, the EPFMR framework invokes multi-user as given below:

$$S \rightarrow SN = \sum_{i=1}^n SN_i \quad (1)$$

Eq. (1) assumes that 'n' SN requirements forward data packets from the sink node 'S' to 'n' DESTN_Node with varying transmission rates. Then, the EPFMR framework uses slots 'Slot_i', through which the time is split into trivial gaps. With the help of 'Slots' in the EPFMR framework, only one SN transmits data packets. During the next slot assignment, in 'Slot_{i+1}', the network evaluates the slot rates 'SR' by measuring Signal to Interference Plus Noise Ratio. The slot rates are obtained by deriving equilibrium between the prevailing slot rates and fairness [25].

The PFMR [26] in the proposed framework is performed by comparing the feasible slot rate for each SN to its average slot rate called the preference SN. The SN in WSN with a higher preference value is then selected for transmission during the next slot and is mathematically formulated as given below.

$$PV_{SN_i}(n+1) = \frac{SN_i(n+1)}{AVG(SN_{i \rightarrow i+1}) * n} \quad (2)$$

From Eq. (2), the preference value 'PV' is assigned for each SN. 'SN_i' is obtained using the overall slot rate $SN_i(n+1)$ and average slot rate ' $AVG(SN_{i \rightarrow i+1}) * n$ '. The EPFMR framework selects the best route path for data packet transmission based on each SN request period. Therefore, by applying preference value-based slot data packet transmission, the EPFMR reduces the energy consumption during routing in WSN [27].

2.2 Preference Value Multi-User Routing Algorithm

Initialize: 'SN_i = 'SN₁, SN₂, SN₃... SN_i' Sink, 'S'

Output: Energy Efficient Routing

Step 1. Begin

Step 2. For Each SN_i and Sink 'S'

Step 3. Evaluates the SN ready for data packet transmission

Step 4. Evaluate preference value to decide upon the next SN 'n' the slot

Step 5. End For

Step 6. End

The preference value multi-user routing algorithm improves the energy consumption during routing in WSN. The algorithm involves two parts. The first part senses the SN that are ready for data packet transmission. Upon successful identification of SN, the preference value is evaluated for each SN to decide upon the sensor node to be selected as the node for packet transmission. With this, the energy consumption during routing is significantly reduced.

2.3 Design of Greedy Instance Fair Method

With the efficient route path discovery using the PFMR, GIFM is applied to the EPFMR framework to minimize interference during routing. The GIFM in EPFMR develops the SN dependent energy-efficient localized route path searching probability. Upon successful identification of route path discovery, GIFM is

applied in the EPFMR framework. Each link with the SN and BS is allocated with an implicit demand according to the interference-free data packet. Next, the greedy instance is applied for each link between the SN and the BS. Then, each SN is assigned a slot. Based on the interference rate, the links are decreased. This process is performed until a fair route with reduced interference is evolved. As a result, the throughput rate of data packet transmission between SN is increased. Tab. 1 given below shows the throughput-improved GIFM using interference rate.

Table 1: GIFM using interference rate

Source (SRC)_node	DESTN_node	Hop node	Interference rate (kbps)
SN ₁	SN ₅	SN ₂ , SN ₃ , SN ₄	140
SN ₁	SN ₅	'SN ₃ , SN ₄ '	125
SN ₁	SN ₅	SN ₄ , SN ₅ , SN ₆ , SN ₇ '	110
SN ₁	SN ₅	'SN ₃ , SN ₄ , SN ₅ '	105

The analysis of a SRC_Node 'SN₁' with the DESTN_Node 'SN₅', there are four routes (*i.e.*, hop nodes) identified as 'SN₂, SN₃, SN₄'; 'SN₃, SN₄'; 'SN₄, SN₅, SN₆, SN₇' and 'SN₃, SN₄, SN₅' each having a different inference rate. By applying GIFM, the hop nodes, 'SN₃, SN₄, SN₅' with minimum interference rate of '105', is finally selected, thus improving the throughput rate.

2.4 Construction of Boltzmann Distribution

In this section, an OGR for SRC_DESTN pairs is designed using the BD to reduce the RST. The objective of the design of BD for multi-user routing in WSN is to update the request time based on the volume of the traffic. Based on the traversing time and volume of the traffic, BD for multi-user routing is designed in the EPFMR framework, as shown in Fig. 3. BD for multi-user routing involves an iterative 'O' value updating to search the route optimality and time optimality for a given source-destination pair. The 'O Value' represented as 'O(i, j)', is the minimum traversing time to the DESTN_Node DN_d with the assumption that a SN traverses from 'i' to 'j'. The BD for multi-user routing with minimum RST is then formulated as given below:

$$O(SN_i, DN_d) \rightarrow t(SN_i, DN_d) + \min O(SN_i, DN_d) \tag{3}$$

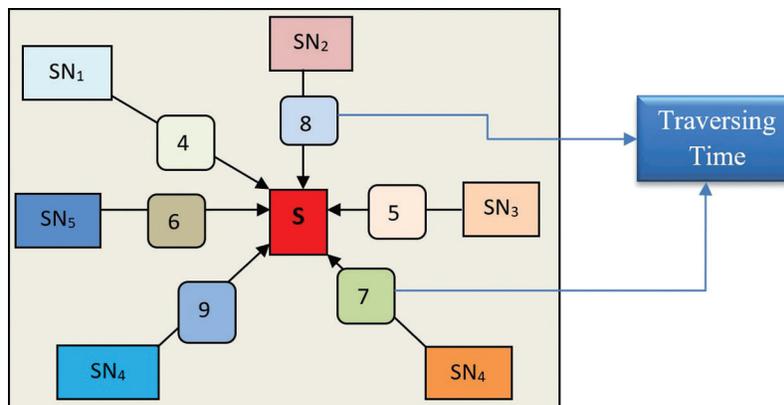


Figure 3: Optimum route search for S-D pairs

Eq. (3), ' $t(i, j)$ ' refers to the traversing time from SN_i to DN_d for ' d ' DESTN_Node. The basic idea of BD for multi-user routing is to measure the average traversing time for the source-destination pairs. The fundamental rule of the traffic assignment strategy is to calculate the average travelling time for the overall Origin-DESTN pairs using Eq. (3). GIFM and BD for multi-user routing follow an instance time-based route searching on WSN to reduce the RST.

The instance time-based route searching on WSN uses a threshold function ' α '. A significant value for ' α ' is assigned when the node density is high, whereas less value for ' α ' is assigned when the node density is low. Therefore, the value for ' α ' is set based on the node density. However, only a tradeoff is obtained. So, to reduce the RST, the EPFMR framework using BD for multi-user routing with delay factor is designed with the aid of a threshold function ' α ' and is formulated as given below.

$$\alpha_i(SN_d) = \frac{O(SN_i, SN_d)}{ND} \quad (4)$$

From Eq. (4), the threshold function ' $\alpha_i(SN_d)$ ' is derived by obtaining the ratio of multi-user routing with minimum RST ' $O(SN_i, DN_d)$ ' and the number of nodes in the network ' ND '. As a result, the delay factor in identifying the OGR for SRC-DESTN pairs is obtained by reducing the RST.

3 Experimental Settings

In this section, we simulated the effectiveness of EPFMR's new framework. In WSN, the EPFMR model was evaluated by comparing it to Traditional Underwater acoustic sensor networks with Learning Automata-cluster (TULA) and Store Carry and Forward (SCF). The simulation parameters used to execute the experimental studies are declared below in Tab. 2. The SN in the EPFMR system has a consistent topology.

Table 2: Specifications for simulation

Limits	Values
Network simulation area	1500 m × 100 m
Number of SN	10, 20, 30, 40, 50, 60, 70
Data packets	8, 16, 24, 32, 40, 48, 56
Communication range	30 m
Node speed	0–10 ms
Test time	1500 s
Test counts	7

The network is comprised of 70 nodes that are distributed randomly throughout the WSN and increases traffic every 10 ms. The sink node receives network packets from various SN ranging from 8 to 56 and sends the data to the sink node. Each data packet size ranges from 100 to 512 kb. The simulation time ranges from 500 to 1500 ms, and the following performance measures are tested in WSN: power consumption during route discovery, average energy per packet, throughput, and RST.

The device of the energy consumed by a single SN and the total SN in a WSN is the energy consumed during route discovery.

$$EC = Energy_{SN} * Energy_{S \rightarrow DN_d} \quad (5)$$

In Eq. (5), 'EC' is the energy consumption during which 'Energy_{SN}' represents the energy consumed by each SN to reach sink node 'S' and 'Energy_{S→DN_d}' represents the energy consumed from the sink node 'S' to DESTN_Node 'DN_d'.

Throughput is the ratio of packets sent by the SRC Node to the sink and packets appropriately collected by the respective DESTN Node *via* the sink node.

$$T = \frac{DP_S}{DP_{DN_d}} * 100 \quad (6)$$

From Eq. (6), the throughput 'T' is measured using data packets sent by the SRC_Node. 'DP_s' and data packets are received at the destination end 'DP_{dn_d}'. Time taken for route search is the difference between the end time and start time for route search by the sink in WSN.

$$RS_t = (End\ time_{RS} - Start\ time_{RS}) * ND \quad (7)$$

From Eq. (7), the time for route search is measured using 'RS_t', whereas 'End_Time_{RS}' represents the end time for route search by the sink and 'Start_Time_{RS}' represents the start time for route search by the sink in WSN. The average energy packet is measured using the total energy required for each packet to the packet size and are formulated below in Eq. (8).

$$AEPP = \frac{Total\ energy\ required}{Packet\ size} \quad (8)$$

4 Result and Discussion

In this section, the EPFMR's new framework results were evaluated and compared with two previous TULA and SCF techniques in WSN. The EPFMR framework's nodes are linked in a consistent topology. The following parameters, such as energy consumption during route discovery, throughput, RST, and energy consumed per packet in WSN, are evaluated to determine the effectiveness of the EPFMR protocol.

4.1 Impact of Energy Consumption During Routing

The values obtained from Eq. (5) are presented in Tab. 3 for various node densities using the presented EPFMR framework and extensively compared with TULA and SCF. Significant experimental findings were summarized and compared with the conventional TULA and SCF, respectively, to observe the efficiency of the EPFMR approach.

Table 3: Energy consumption during routing

Node density	Energy consumption during routing (J)		
	EPFMR	TULA	SCF
10	35.32	56.12	61.35
20	41.30	60.33	65.33
30	44.53	69.56	74.56
40	42.13	65.16	70.16
50	47.21	67.24	72.24
60	49.14	69.17	74.17
70	45.22	50.25	55.25

The results presented above Fig. 4 accept that as node density in WSN increases, so does energy usage during routing, which is significantly smaller while using EPFMR. So, the framework EPFMR is said to be highly energy-efficient compared with TULA and SCF. With node density 30, the energy consumption during routing was 44.53 J using the EPFMR framework, whereas 69.56 and 74.56 J were consumed using TULA and SCF. With the use of PFMR, the energy consumption during route discovery is significantly decreased. To reduce the energy consumption during routing, Proportional Fairness assigns the SN in WSN with slots using preference value by comparing the feasible slot rate and average slot rate. As a result, the energy consumption during routing using the EPFMR framework is improved by 35.05% compared to TULA and 47.93% compared to SCF.

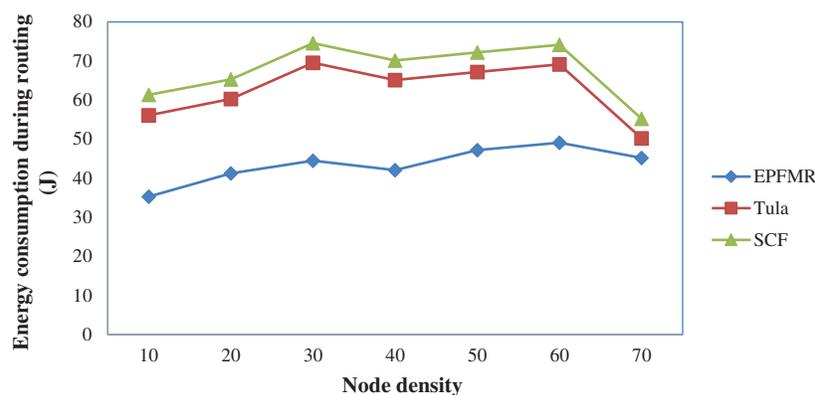


Figure 4: Measure of energy consumption

4.2 Impact of Throughput

Throughput for various data packets in the range of 8 to 56 is computed using Eq. (6). The number of SN in the testing method ranges from 10 to 70. Tab. 4 shows the result of seven model testings used to determine the network throughput. The throughput for three different methods is expressed as a percentage. The throughput rates obtained in our approach are similar to those obtained with *state-of-the-art* techniques.

Table 4: Throughput analysis

Data packets	Throughput (%)		
	EPFMR	TULA	SCF
8	75.23	63.45	51.16
16	69.13	60.11	50.08
24	76.19	71.17	67.14
32	78.24	73.22	69.19
40	72.48	67.46	63.43
48	80.23	75.21	71.18
56	79.14	74.12	70.09

A comparison was made between the targeting results of throughput rate using EPFMR framework with two advanced methods TULA and SCF, and Fig. 5 represents the visual comparison based on the pertinent

data provided. The throughput rate for data packets is given, with each SN sending different data packets to the sink node.

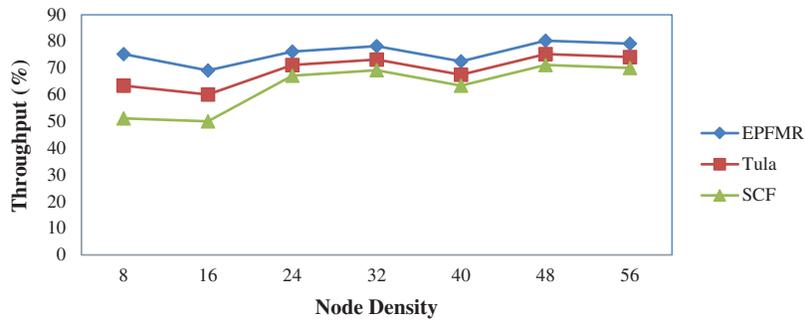


Figure 5: Measure of throughput

If the sink node receives 24 data packets, the throughput rate using the EPFMR framework will improve to 6.58% compared to TULA and 11.87% compared to SCF. This improvement is observed in the EPFMR framework by incorporating GIFM. The benefit of using GIFM in the EPFMR framework is that it performs data packet transmission based on the interference rate rather than directly. For effective transmission, the SN data packet with the lowest interference rate is chosen. Later, the data packet is transmitted between sensors and sink nodes using interference rate thus, increasing throughput by 10.95% vs TULA and 2.63% vs. SCF.

4.3 Impact of Route Searching Time

Tab. 5 further compares the RST to node density at different time intervals in WSN. 70 SNs were used in the experimental studies, and the RST is measured in milliseconds (*ms*).

Table 5: Route searching time

Node density	Route searching time (ms)		
	EPFMR	TULA	SCF
10	1010	1255	1458
20	1080	1330	1455
30	1100	1350	1475
40	1085	1330	1445
50	1120	1370	1485
60	1115	1365	1480
70	1128	1378	1493

Fig. 6 depicts the RST for the EPFMR framework, TULA, and SCF vs. 70 different SN. Compared to the other two methods, the RST returned by the EPFMR framework has increased slightly but not linearly with node density.

The RST for SRC-DESTN pairs using the EPFMR framework has been reduced. With the application of BD based on the instance time-based route searching, the traversing time and volume of the traffic are considered resulting in the minimized RST using the EPFMR framework. The BD with the aid of the ‘O’

value measures the average traversing time for the SRC_DESTN pairs, and therefore OGR is obtained for SRC_DESTN pairs. After identifying the OGR, the route is provided to the SRC_SN, reducing the RST by 15.82% and 31.12% compared to TULA and SCF, respectively.

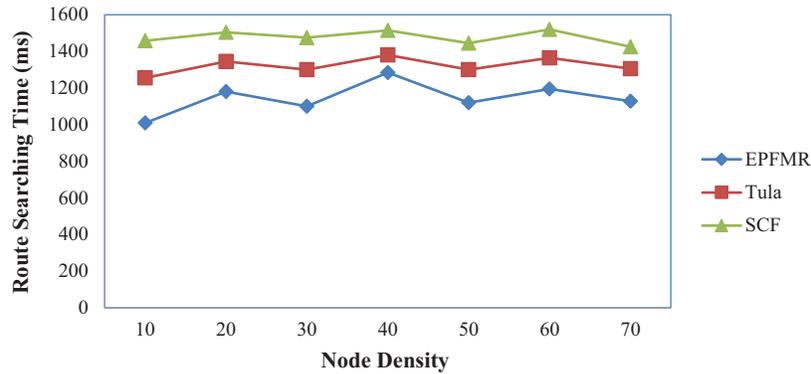


Figure 6: Measure of route searching time

4.4 Impact of Average Energy Per Packet

Eq. (8) is used to calculate the total energy per packet. Tab. 6 represents the average energy per packet for 56 data packets with node mobility of 15 ms. Significant experimental results showed the comparison between the existing TULA and SCF frameworks to demonstrate the efficiency of the EPFMR framework.

Table 6: Average energy per packet

Data packets	Average energy per packet (J)		
	EPFMR	TULA	SCF
8	4.25	5.25	6.35
16	6.35	8.40	11.45
24	10.14	12.19	15.24
32	9.33	11.38	14.43
40	14.59	16.64	19.69
48	17.23	19.28	22.33
56	15.21	17.26	20.29

The effect of average energy per packet on varying data packets ranging from 8 to 56, and the average energy using three methods varies with data packet size. The results presented above confirm that because the source SN sends increased data packets to the sink node for route identification, so does the average energy per packet.

The average energy per packet for the three methods varies with the number of data packets sent. As displayed in Fig. 7, the EPFMR framework outperforms the other two different methods, TULA and SCF. This is due to the use of the threshold function ' α ' with BD, which is determined based on node density. The multi-user route optimization ratio with the minimum RST is obtained, and the ratio of nodes is used for each packet sent by different nodes. This reduces the average energy per packet for route authentication using the EPFMR framework by 22.08% compared to TULA and 56.84% compared to SCF.

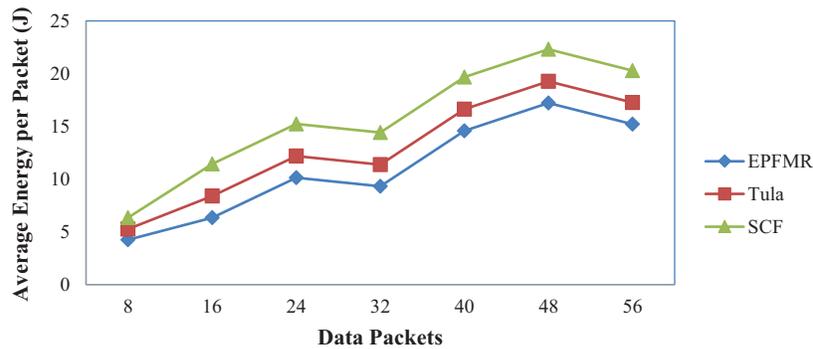


Figure 7: Measure of average energy per packet

5 Conclusion

This paper introduces a new EPFMR framework to reduce energy consumption during route discovery and RST with varying node density and packet size. This primary goal is met by identifying routes based on their periods and the percentage of interference observed by SN and transmitting network packets to the sink node accordingly. The goal implements a PFMR with BD, which saves energy and finding the optimal route for the various sensors with different packets in a WSN. As a result, throughput is enhanced, ensuring data packet transmission between the sender SRC and the DESTN Node *via* the sink node. An adaptation of the interference rate reduces the RST, ensuring the most negligible average energy per packet with the optimum solution. Experiments were performed out, and performance was compared with various process variables in terms of various performance measures, such as energy consumption during routing, throughput rate, average energy per packet, and RST. The findings demonstrate that the EPFMR framework outperforms TULA and SFC by optimizing energy consumption during routing by 41.49% and average energy per packet by 39.46%.

Funding Statement: The authors received no specific funding for this study.

Conflicts of Interest: The authors declare that they have no conflicts of interest to report regarding the present study.

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