

An Efficient Queue Length Based Estimated Rate for Congestion Control in Wireless Sensor Networks

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Abstract:

Over the last few years, Wireless Sensor Networks (WSNs) attained notice in analysts, producers and the clients of distantly tracking jobs as well as efficient information collecting within various environments. In recent years, energy efficiency and congestion control are the major concerns in applications of WSN. Clustering gives a valuable means of extending the duration of a wireless sensor network. Jamming within WSNs forms several concerns such as packet collision, buffer overflow, queuing delay which directs toward corrupt the Quality Of Service (QoS) factors as packet delivery ratio, end to end delay as well as Average energy consumption of the wireless node. The previous system designed an Adaptive Path Selection and Energy aware Cluster-based Routing (APSECR) protocol. However, the web outputs are yet further getting worse as a result of the web jamming. To solve this problem the proposed system designed a Congestion Control and Optimal Link Based Routing (CCOLR) mechanism for achieving traffic free packet transmission in WSN. In this proposed research work, the sensor nodes are grouped initially and an optimal cluster heads are selected by using Multi Objective Anarchic Society Optimization algorithm (MOASO). An optimal path selection is performed based on the link cost with includes Extended Medium Time Metric (EMTM) and throughput for inter cluster as well as intra cluster communications. After the selection of optimal paths, the congestion detection and control is performed based on the Queue Length based Estimated Rate (QLER) mechanism. In QLER approach, traffic rate for packet transmission has been modified depending upon the estimation rates. Based on the estimation rates packets are transmitted to the destination. The investigational outcomes shows so as the planned structure attains enhanced functioning than the actual structure with respect to throughput, end to end delay, packet delivery ratio as well as energy consumption.

Keywords: Wireless Sensor Network (WSN), Cluster Head (CH), Anarchic Society Optimization (ASO) and queue length

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I INTRODUCTION

Nowadays, Wireless Sensor Network (WSN) has achieved critical jobs within different application as well as has pulled in the consideration of specialists because of his/her mind boggling, versatile necessities that

frequently disclose inalienable transactions. WSNs are comprised of sensor hubs associated throughout a specially appointed and self-arranging availability [1]. A WSN comprises of a lot of spatially disseminated independent sensor hub for screening ecological factors, such as,

climate, pressure, also so on, and toward agreeably depart the data throughout the network toward a critical region named sink. The sink of a Wireless Sensor Network collects data as of the sensor hub as well as records the same toward customers by the Internet else by some Private Virtual Network (PVN) [2].

Sensor hubs are serious with restricted assets like battery control, transmission capacity, and memory and so forth. Duration of WSN relies on power unit intensity of hubs like each activity of hub expends vitality; henceforth hub leaves vitality. Along these lines, effective vitality utilization of hubs has been the primary plan concern of WSN. Clustering strategy diminishes the quantity of radio transmissions and builds sensor network lifetime. Thus, clustering procedure can, proficiently, increment lifetime of different sensor applications, for example, robot control, ecological control, workplaces, keen homes, fabricating situations, body zone networks, and submerged sensor network.

Cluster partitions the hubs within free groups as well as chooses a leader in every bunch [3]. Hubs transmit detected information toward individual group leader [4]; bunch head applies data combination/total to lessen the gathered information toward several important data as well as transmits collected information toward Base Station (BS). Correspondence among both hubs has been the primary vitality expending procedure which relies on the separation among the both. Clustering dodges extended separation correspondence of part hubs as well as just group leaders have been imparting toward BS. Toward stack stability the web, job of bunch leader has pivoted with the entire hubs.

Cluster leader determination has been unequivocal on behalf of the exhibition of cluster calculations [5-6]. Absolute intra-group correspondence separation and complete separation of bunch heads to put together station

depends with respect to number of group heads. In the event that quantities of group leaders have been fewer, complete separation of bunch leaders toward BS diminishes whereas all out intra-group correspondence separation increments. At the point when quantities of group heads are increasingly, absolute intra-bunch correspondence separation diminishes yet there is increment in all out separation of bunch heads to base station.

In wireless sensor network, congestion has gotten progressively critical. Congestion is a condition of being stuffed with network traffic and because of this there is a decrease of the Quality of Service (QoS) within the web. The QoS alludes to the capacity of the network to accomplish most extreme execution. Congestion in a WSN by and large happens when the traffic load on a specific sensor hub surpasses the accessible support limit. Congestion additionally happens in view of the uncalled for use of the network assets. It for the most part happens either at the sensor hub or at the wireless connection because of their accessible limitations and due to that there are two distinct degrees of congestion: hub level and connection level congestion. Customarily, in hub level congestion, either the cradle usage or the jumble between the traffic appearance rate and the traffic takeoff rate is analyzed. Then again, interface level congestion is observed through the channel use. In the two cases, the significant part is the convenient location of congestion and its relief from the network.

Current congestion control approaches have a few challenges, for example, the ahead traffic the executives don't contemplate the traffic estimations upon the alternate ways [7]. As for origin sensor hub toward sink, the needs are determined to jump tally rather than real parcel delay [7-9]. Additionally, the conveyance of traffic stacks by blocked as well as alternate ways is never dealt with appropriately.

The remaining of the article has

been prepared like below. In Module 2, clustering as well as congestion in WSN is explained along with the design challenges, clustering parameters, and taxonomy of hierarchical clustering. Module 3 is about optimal link based path selection and Queue Length based Estimated Rate (QLER) for congestion control is explained in detail. The experimental results are summarized within Module 4. The closure has been discussed in Module 5.

II LITERATURE SURVEY

Rao et al (2017) structured a Particle Swarm Optimization (PSO) depending on vitality effective bunch leader choice calculation for WSNs. The calculations are reproduced by a productive plan of particle encoding and wellness work. For the vitality proficiency of the proposed PSO approach, consider different parameters, for example, intra-bunch separation, sink separation and remaining vitality of sensor hubs. The framework likewise present group arrangement in which non-bunch head sensor hubs join their CHs based on determined weight work. The calculation is tried broadly on different situations of Wireless Sensor Networks, changing quantity of sensor hubs as well as the cluster head. The outcomes are contrasted with some current calculations with demonstrate the predominance of the proposed calculation [10].

Wang et al (2018) planned a Hybrid Multihop Partition-Based Clustering (HMPBC) steering convention that could fix a particular domain, gather attentions as well as wants, stability web loads, as well as drag out the web duration. In Hybrid Multihop Partition-Based Clustering, the single-chain structure inside bunch and the group leaders determination relying upon lingering vitality are operated without anyone else's input association, and the area least crossing hierarchy organization between groups has been built up through the BS. The planned convention has been reenacted within MATLAB through the

web replica of 200 m × 200 m 400 hubs. Reproduction outcomes illustrates such as Hybrid Multihop Partition-Based Clustering is 33.31% as well as 39.70% extended duration, little changes of normal vitality utilization of every hub for each iteration, also further exact estimation of zone factors compared to LEACH-MLOR as well as the vitality effective hybrid multihop clustering plan [11].

Brahma et al (2010) planned a circulated jamming command calculation on behalf of hierarchy supported interchanges within WSNs which looks toward flexibly allocate a reasonable as well as proficient transmission rates for every hub. In this calculation, every hub screens its total yield and info traffic rates. Based on the distinction of the two, a hub at that point chooses either to increment or reduction the transmission capacity allocable to a stream beginning from itself and to those being steered through it. Since the application necessities in sensor network pursue no basic quality, structure abstracts the idea of decency, taking into account the improvement of a nonexclusive utility controlling module. Such division of the utility and decency controlling modules empowers everyone to utilize a separate control law, in this manner depicting an increasingly adaptable structure. The working of planned congestion control is free of the fundamental directing calculation and is intended to adjust to changes in the hidden steering topology. The outcomes recommend that the proposed convention secures an essentially high great put of about 95% of the genuine communication rates, meets rapidly toward the ideal rates, and furthermore achieves the ideal reasonableness [12].

Jaiswal et al (2013) planned a Fuzzy Based Adaptive Congestion Control (FBACC) has been projected that utilizes cushion in habitation, members as well as traffic rates. The Fuzzy Based Adaptive Congestion Control gives a fuzzy rationale supported jamming assessment, shrewd

approach toward go down bundle if there should arise an occurrence of jamming toward worthy standard levels as well as the major significant is that direct traffic rates. This also develops execution of web above different conventions that is by and by accessible. Execution of FBACC is assessed and FBACC distinguishes the network congestion all the more vitally and consistently and it additionally adjusts to current traffic rate as for result of past members with past traffic rate to lessen the bundle misfortune. At long last, FBACC spares vitality of retransmission since parcel misfortune is negligible because of traffic adaption [13].

Ahmed and Paulus (2017) build up a congestion evasion and relief system. As a result, choose courses depending upon the separation among transmitter as well as recipient, Relative Success Rate (RSR) estimation of hub along with cushion inhabitation of a hub. Depending upon the above three factors, characterize an effectiveness capacity is functional toward all the bystander of a sender hub. Henceforth the sender hub picks the most noteworthy U-esteemed hub

since it subsequently jump hub between the bystanders within bundle sending. In this way the structured framework maintain a strategic distance from congestion by picking non-clogged hubs like it then jump hub as well as afterward personally alleviate jamming depending upon Relative Success Rate value [14].

III PROPOSED METHODOLOGY

In this research work, system designed an optimal link based path selection and Queue Length based Estimated Rate (QLER) for congestion control and lifetime improvement of WSN. Here, an optimal Cluster Head (CH) selection should be made with the help of Multi Objective Anarchic Society Optimization (MOASO). Then inter as well as intra cluster communication is performed mutually based on the Extended Medium Time Metric (EMTM) and throughput. Then the congestion detection and control is performed based on the Queue Length based Estimated Rate (QLER). The overall process of the system has been described within Figure 1 [21-23].

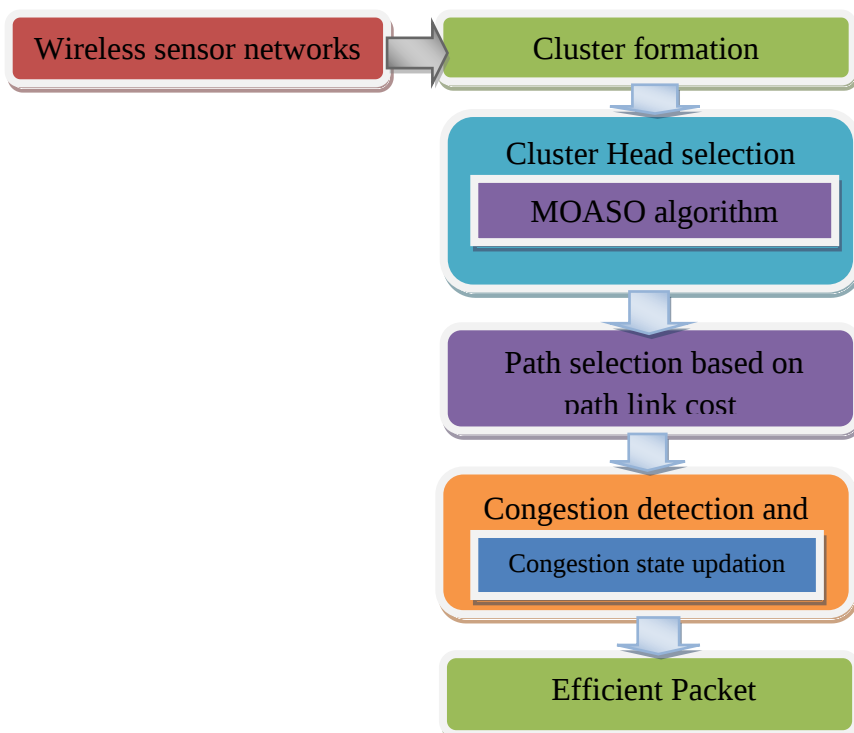


Fig. 1. Flow diagram of the proposed system

3.1 NETWORK MODEL

The projected web representations of WSNs have N sensor hubs conveyed within an irregular territory of M. The underlying vitality for every one of the hubs is same and the network is homogeneous. All hubs have been fixed once sent within the zone further more contain area data. It has only one BS situated exterior the zone. The hubs have been examined for kicking the bucket just when their vitality is depleted. Hubs can alter their transmission control based on beneficiary separation and sense data at a fixed interim.

3.2 CLUSTER FORMATION AND CLUSTER HEAD SELECTION USING MULTIOBJECTIVE ANARCHIC SOCIETY OPTIMIZATION (MOASO)

Clusters proceed toward extending the web duration through the stack balanced web. In this proposed research work, clustering configuration has been done based on the distance. The cluster members are transmitting the packet Cluster Head (CH). For selecting the optimal cluster head, the key factors of wireless sensor network like Average intra-cluster space, Transmission power E_{tx} , Node degree, Remaining Energy (E) and number of neighbors have been used for performing the successful CH picking. Cluster head selection is performed by using Anarchic Society Optimization (MOASO).

The fitness function consists of following parameters

- ✓ Average intra-cluster distance
- ✓ Transmission power E_{tx}
- ✓ Node degree
- ✓ Remaining Energy (E)
- ✓ Number of neighbors

3.2.1 Multiobjective algorithm

1. Average intra cluster distance (Dis_{avg})

It is characterized as the normal of the whole of the separations of all the sensor nodes from their chose CH.

$$\frac{1}{l_j} \sum_{i=1}^{l_i} dis(s_i, CH_j) \quad (1)$$

In intra-cluster correspondence, all sensor nodes expend some energy to send data to their CH. So as to devour less energy, we have to decrease this normal intra-cluster correspondence separation. This implies we have to choose a sensor as a CH which is near all the sensor nodes.

2. Transmission power (T_x)

Then, the transmission power could be found as a result of the below formula:

$$P_{AB} = R^2 \quad (2)$$

Where,

P_{AB} - minimum required transmission power as of node A till gateway B,

R - Transmission range in (m).

3. Node degree (N_{deg})

The level of the node x meant by $d(x)$ is the quantity of neighbors of node x (i.e.) the quantity of connections. CH is chosen based on the node with greatest number of neighbors. A node of $d(x) = 0$ is detached (no neighbors). The level of chart G is meant in Equation (4).

$$d(G) = \{d(x)\} \quad (3)$$

4. Residual energy

It has been likewise a significant factor to consider which has also been characterized like the energy held within every nodes following the past accepting as well as broadcasting procedure. It is processed like:

$$R_e = I_e - (E_t + E_r) \quad (4)$$

Where,

I_e - Initial Energy

E_t, E_r - energy needed to send and receive correspondingly

5. Number of neighbor (N_{neigh})

The number of neighbor nodes is taken as an account

Weight of each node can be calculated using weight based clustering method.

$$W = w_1 \text{Dis}_{\text{avg}} + w_2 T_x + w_3 N_{\text{deg}} + w_4 R_e + w_5 N_{\text{neigh}} \quad (5)$$

$$w_1 + w_2 + w_3 + w_4 + w_5 = 1 \quad (6)$$

Where,

w_1, w_2, w_3, w_4, w_5 –weighting factor

The weighted clustering program chooses a CH just as the mass values of every node.

3.2.2 Multi Objective Anarchic Society Optimization algorithm (MOASO)

In this proposed research work, Anarchic Society Optimization (ASO) is utilized on behalf of optimal CH picking. Assume that S has been the solution space (wireless sensor network) and $f: S \rightarrow R$ (refers to real numbers) is a function which should be minimized in the space of S [15-17]. Also, assume that a community with N member (number of nodes) is inquiring in favor of the most excellent location of living (worldwide least of function f in the S) in a mysterious ground (solution space). $X_i(k)$ indicates the location of the i^{th} member within the k^{th} iterations. The most excellent location experienced through the entire member within the k^{th} iterations are $X^*(k)$. The finest location visited through the i^{th} members during the first k iterations are $P_i(k)$. The most excellent location visited in the entire member during the first k iterations are $G(k)$. The elements of the algorithm according to movement policies are detailed as follows.

Algorithm Procedure

Initial, various network individuals (number of sensor nodes) are chosen haphazardly inside the arrangement space. At that point, the wellness (weight estimations) of each part (node) is resolved. As per the determined weight worth and correlation with $X^*(k)$, and G_{best} , the development arrangement and another situation of the part will be resolved. After a sufficient number of emphases, at any rate one of the individuals will arrive at the ideal position.

Movement Policy Based on Current Positions

The principal development arrangement of the i^{th} part within the k^{th} emphasis [$FI_i(k)$] has been received depending upon the present location. The Fickleness Index $FI_i(k)$ of part I in cycle k are utilized toward choose these development strategies. This list quantifies the fulfillment of the present situation of the i^{th} part contrasted and other individuals' positions. On the off chance that the target work is certain within S , the Fickleness Index is communicated like anyone of the below accompanying structures

$$FI_i(k) = 1 - \alpha_i \frac{f(X^*(k))}{f(X_i(k))} - (1 - \alpha_i) \frac{f(P_i(k))}{f(X_i(k))} \quad (7)$$

$$FI_i(k) = 1 - \alpha_i \frac{f(G(k))}{f(X_i(k))} - (1 - \alpha_i) \frac{f(P_i(k))}{f(X_i(k))} \quad (8)$$

Where, α_i be a positive digit within $[0,1]$. In this manner, the Fickleness Index has been a digit in the scope of $[0,1]$. As per the estimations of Fickleness Index, the i^{th} part might choose their subsequent location. On the off chance that $FI_i(k)$ is littlest, the i^{th} part is the most excellent location with the entire individuals. Hence this has been smarter toward choose the development arrangement depending upon $FI_i(k)$. Something else, the i^{th} part is an absolutely eccentric development. Subsequently the development approach for the i^{th} part as per the estimation of $F(k)$ can be communicated as:

$$[MP_i^{\text{current}}(k)] = \begin{cases} \text{moving towards } X^*(k) & 0 \leq FI_i(k) \leq \alpha_i \\ \text{moving toward a random } X_i(k) & \alpha_i \leq FI_i(k) \leq 1 \end{cases} \quad (9)$$

Movement Policy Based on Positions of Other Members

The 2nd development approach of the i^{th} part within the k^{th} cycle is received based on the places of different individuals. Albeit every part should move toward G_{best} legitimately, the development of the part isn't unsurprising because of the rebel idea of the part and may push toward another network part. In this way, the outside inconsistency list $E(k)$ of the i^{th} part within the k^{th}

cycle is determined through

$$EI_i(k) = 1 - e^{-\theta_i [f(X_i(k)) - f(G(k))]} \quad (10)$$

$$EI_i(k) = 1 - e^{-\delta_i D(k)} \quad (11)$$

in which θ_i and δ_i are sure numbers and $D(k)$ is a fitting scattering measure like coefficient of variety $CV(k)$. Above Equation characterizes the separation of network part I from G_{best} . In the event that the network part is near G_{best} , it will have a more rationale conduct. Else, it shows an anarchic conduct based on political agitation. Above Equation characterizes an assorted variety list in network which has an immediate association with the decent variety of the network individuals. For the situation that this record is chosen, the network individuals should carry on more consistently and they are less differentiated. In this manner, with thought of a limit for $E(k)$, it is conceivable to characterize the development approach based on the places of different individuals as pursues:

$$[MP_i^{society}(k)] = \begin{cases} \text{moving towards } G^{best} & 0 \leq EI_i(k) \leq \text{threshold} \\ \text{moving toward a random } X_i(k) & \text{threshold} \leq EI_i(k) \leq 1 \end{cases} \quad (12)$$

The nearer the limit has been 0, the further outlandish the part developments likely. Since the limit meets comes to 1, the individuals might act all the further coherently.

Movement Policy Based on Previous Positions

The 3rd development approach of the i^{th} part within the k^{th} cycle is received based on the past places of the individual part. So as to choose this development arrangement, the situation of the i^{th} part in the k^{th} cycle is contrasted with P_{bestI} . In the event that the situation of the part is near, the part acts all the more sensibly. Something else, the part shows nonsensical conduct. To decide the development arrangement based on past positions, the inward inconsistency record $I(k)$ of the i^{th} part within the k^{th} cycle has been characterized like pursues:

$$II_i(k) = 1 - e^{-\beta_i [f(X_i(k)) - f(P_i(k))]} \quad (13)$$

Where, β_i is a positive number. Like the past approach, with choice of an edge for $II_i(k)$, the development strategy can be characterized based on past positions as pursues:

$$[MP_i^{past}(k)] = \begin{cases} \text{moving towards } P_i^{best} & 0 \leq II_i(k) \leq \text{threshold} \\ \text{moving toward a random } X_i(k) & \text{threshold} \leq II_i(k) \leq 1 \end{cases} \quad (14)$$

The nearer the edge equals 0, the further sillier the part developments likely. Since the edge combines to 1, the individuals might act all the further coherently.

Combination of Movement Policies

For choosing the last development arrangement, the three strategies talked about above are joined with one another. After the development strategies are determined, every part should consolidate these approaches with a strategy and push toward another position. Perhaps the least complex technique is to choose the strategy with the best answer. The following option is to consolidate the development arrangements with one another consecutively that has been alluded towards since the successive mix rules. The hybrid strategy could further utilized on behalf of persistent issues programmed like chromosome, otherwise been utilized within a successive method for consolidating the development approaches.

Algorithm 1: ASO Algorithm

Input: N members (Number of nodes)

Output: Cluster Head (CH)

1. Initialize the N number of nodes, node position and set of iterations counter $I=0$
2. Generate M initial solutions and evaluate their fitness values (weight values)
3. While (termination criteria are not satisfied)
4. Determining the values of $X^*(k)$, P_i^{best} and G^{best}
5. For $i=1:M$
6. Computing $FI_i(k)$

7. If $FI_i(k)$ is less than threshold $X_i(k)$, then $X_i(k)$ moves towards $X^*(k)$
8. Else $X_i(k)$ moves towards a random member
9. End-If
10. End-For
11. For $i=1:M$
12. Computing $EL_i(k)$
13. If $EL_i(k)$ is less than threshold $X_i(k)$, then $X_i(k)$ moves towards G^{best}
14. Else $X_i(k)$ moves towards a random member
15. End-If
16. End-For
17. For $i=1:M$
18. Computing $IL_i(k)$
19. If $IL_i(k)$ is less than threshold $X_i(k)$, then $X_i(k)$ moves towards P_i^{best}
20. Else $X_i(k)$ moves towards a random member
21. End-If
22. End-For
23. For $i=1:M$
24. Updating the position by combining the movement policies
25. Calculating the fitness values (weight values) for the members (nodes) of society (network)
26. End-for
27. End while
28. Report the best solution
29. .End

3.3 PATH SELECTION BASED ON LINK COST

Link-quality benchmark appearstoward the standard of every link within the pathwayas well as the benchmarkhave beendepending upon factors of link standard. The previous ETX methods, the probe packet sizes have beensetalsodoesn't represent the actual traffic packet size mergeas well as transmitted by the webvital rates. as a result, in the ETX representation, the packet sizes and the diversebroadcast rate isneverenclosed in the mode. To solve this problem, the proposed system designed an Extended Medium Time

Metric (EMTM)-dependingprocessthat adopts the real time packet size. The weight of a path p is simply based on the reliability of the links on the path. The Extended Medium Time Metric (EMTM)-based method can be expressed as

$$EMTM = LQ = \sum_{k=1}^{\infty} (\text{overhead} + S/B)/\text{rel} (1-p) \quad (15)$$

Here, p means error rate of the transmission. Overhead is the per packet overhead of the link as well as S is the size of the packet furthermore B is the bandwidth of the link. at last, rel is the reliability index and has been similarlike the delivery ratio of the packet.

The output $T_i(P)$ of a flow on a path link have beenpresumed like the below:

$$T_i(P) = \frac{P}{d_i(P)} \quad (16)$$

The transmission delay of a packet over link i with size P have beenshapedroughlylike

$$d_i(P) = (\theta_i + P/b_i) \quad (17)$$

Where, θ_i is the overhead time, along with b_i is the effective transmission rate for link i .

$$Cost_{Link} = \sum_{l=1}^N \alpha LQ + \beta T_i(P) \quad (18)$$

Where, α and β are the weights assigned to each metric.

Based on the link cost the optimal paths are selected between the Cluster Head (CH). Once the optimal paths are selected the congestion detection is performed to reduce the packet transmission delay.

3.4 CONGESTION DETECTION AND CONTROL

The CHnodes advance the example information parcel toward the goal through the middle of the road cluster head node. Upon gathering of the information bundle by the moderate cluster head nodes, queue length is anticipatedas well as nodes are confirmed on behalf ofjammingposition. These procedures are rehashed next to middle of the each road CH node; lastly the bundle arrives at the goal nodes. Subsequent to the gathering of the information parcel, the goal nodesverify for the

rates data within the bundles Internet Protocol header fields. Alongside further fundamental zones, anticipated rates are replicated toward an affirmation parcel as well as transmitted as an input to the transmitter. The transmitter executes rate control as indicated by the anticipated rates got as of the goal.

3.4.1 Rate Estimation

The origin nodes advance the information bundle toward the goal by the middle of the road CH node. Depending upon the cradle position of the middle CH node, the parcel showing up otherwise parting by the halfway CH nodes are moreover disposed as of the nodes otherwise refreshed by the position of the nodes.

The rates of the approaching as well as active bundle are anticipated like pursues.

The rates of the approaching packets (τ_i) are anticipated since the proportional of the appearance time interim of the approaching parcel.

$$\tau_i = \frac{1}{T_i} \quad (19)$$

T_i is characterized since the time interim of two sequential parcels got at the nodes.

The rates of the active packets (τ_o) are estimated since the proportional of the administration time of the parcel.

$$\tau_o = \frac{1}{T_o} \quad (20)$$

T_o is characterized since the time interim among the time where a bundle lands next to nodes as well as time which has been broadcasted effectively. This has been likewise characterized like cumulativeness of the ideal opportunity of queues, impact, shrink as well as broadcast.

The binomial distribution for a random packet with parameters n and p represents the sum of n independent packets which may assume the values 0 or 1.

$$T_{ic} = (1 - (np(1 - p)))T_{ip} + (np(1 - p))(T_l - Tp) \quad (21)$$

$$T_{oc} = (1 - \varepsilon)T_{op} + \varepsilon T_s \quad (19)$$

Where T_{ic} = time interval of the recently received packets

T_{oc} = service time of recent packets

T_{ip} is time interval of finally received packets

T_{op} = service time of the final packet

T_l = arrival time of the final packet

T_p = arrival time of the penultimate packet

Hence $T_l - T_p$ gives the packet arriving interval

T_s = service time of the final departing packet.

ε – Exponential smoothing Function for weighing T_l and T_o with value between 0 and 1.

3.4.2 Estimation of Queue Length

The queue length explains the entire traffic load within intermediate CH nodes. Generally, once overload traffic runs throughout the neighbor nodes, subsequently it would have additional amount of packet within the crossing point queues. Therefore average queue size (LQ) has been explained like the node's traffic within a lengthy tenure LQ can be found out by the below formula

$$L_Q = \psi * L_{Qold} + (1 - \psi) * L_{Qc} \quad (22)$$

Where,

L_{Qc} = present values of the queue length based on the T_{ic} and T_{oc}

ψ = stable within the array [0, 1]

Algorithm 2: Congestion detection

1. The origin transmits the sample information packet toward the target by the intermediary CH node.

2. Set L_{Qth}

Where,

L_{Qth} - predetermined threshold values of queue length.

3. On receiving the sample information packet, intermediary CH nodes validate the queue length

as well as channel utilization mutually furthermore allocate values for Cb like below scenarios.

4. IF $LQ > LQ_{th}$
5. Set congestion ($Cb = 1$)
6. End IF
7. IF $LQ = LQ_{th}$
8. Set no congestion ($Cb = 0$)
9. End IF

Subsequent to the location of jamming, the moderate CH nodes evaluate the latest rate data utilizing conditions (21) and (22). The source refreshes its parcel transmitting rates among its estimation rates. It could be achieved in the wake of getting affirmation bundle from recipient as criticism with respect to the nodes congestion status.

3.4.3. Estimated rate based congestion control

The proposed research work accepts the accompanying measurements of the rate control method

Let

Cb = congestion bit located within the parcels internet protocol header

τ_s = present rate of the transmitter (or) transmission rate

τ_c = present rate of the nodes

τ_e = rate anticipated within the intermediary CH nodes

τ_p = anticipated rates of the preceding nodes

δ = parameter that regulates the rate values.

ACK = acknowledgement packets

The calculation for rate depending jamming controls are portrayed beneath.

1. The present rates of the nodes are considered as inverse of their queue length.

$$\text{Current rate, } \tau_e = \frac{1}{q} \quad (23)$$

Therefore the rate anticipated on the intermediary CH nodes (τ_e) has been given below

$$\tau_e = \delta \cdot \tau_p + (1 - \delta) \cdot \tau_c \quad (24)$$

Where p = anticipated rate of the preceding nodes. The parameter δ has been utilized for getting the uniform/regular values of the rate.

2. If $\tau_e > \tau_p$
3. Fix the packet by τ_e
4. End if
5. The intermediary node pop in the data of congestion bit (Cb) as well as the anticipated rate (τ_e) with the alternative fields of regular internet protocol header.
6. The procedure of rate estimations are reshaped within each middle CH nodes and packet is refreshed by estimation rates as an incentive according to the above cases lastly the bundle arrives at the goal node.
7. past the gathering of the information parcel, the goal nodes verify the estimation of Cb as well as rate data within the bundles internet protocol header field.
8. In addition to Cb , τ_e has likewise duplicated toward an ACK packet as well as criticism toward the transmitter through the goal nodes. The ACK parcel has the origin IP, goal IP, as well as data time parameters.
9. After getting an ACK parcel, transmitter verifies the estimation of Cb . On the off chance that $Cb = 1$, at that point it refreshes the present transmitting rates toward the estimation rates e . (i.e)

$$\tau_s = \tau_e \quad (25)$$

10. End

As the transmitting speed is balanced depending upon the estimation speed as of the middle of the road nodes, this method is superior to anything the conventional congestion control strategy.

IV EXPERIMENTAL RESULTS

In this proposed research work, the simulation was conducted using NS2 simulator. The proposed Congestion Control and Optimal Link Based Routing (CCOLR) and existing Adaptive Energy aware Cluster-based Routing (AECR) and

Adaptive Path Selection and Energy aware Cluster-based Routing (APSECR) protocols are compared in terms of end to end delay,

throughput, packet delivery ratio and energy consumption. The simulation settings are given in Table 1.

TABLE 1
SIMULATION PARAMETERS

Parameter	values
No. of Nodes	100
Protocol	Congestion Control and Optimal Link Based Routing protocol
Area Size	1100 X 1100 m
Mac	802.11
Radio Range	250m
Simulation Time	60 sec
Packet Size	80 bytes

V PERFORMANCE EVALUATION

1. End-to-end delay

The average time taken by a packet to transmit

from source to destination across the network is well-known as End to End delay.

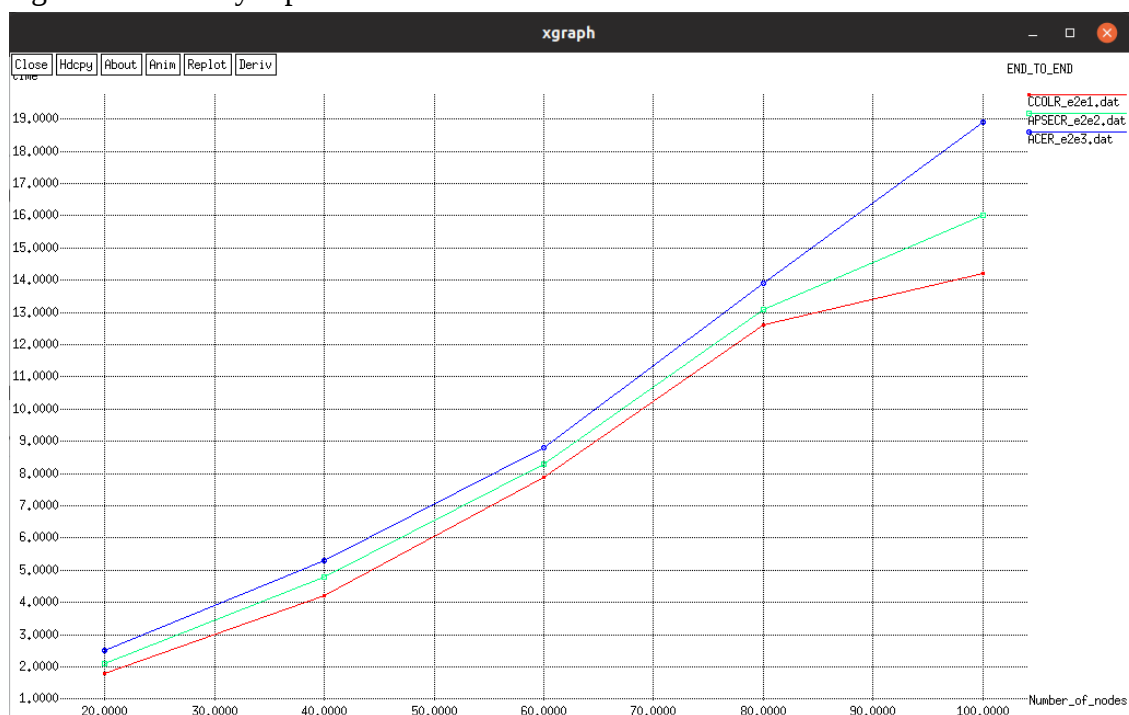


Fig. 2. End-to-end delay comparison

The end to end delay performance of the proposed Congestion Control and Optimal Link Based Routing (CCOLR) approach is compared with the existing Adaptive Energy Aware Cluster-based Routing (AEER) and Adaptive Path Selection and Energy aware Cluster-based Routing (APSECR) protocols. In x-axis number of nodes is taken and end to end delay is plotted in

y-axis. From the

graph it is clear that the proposed CCOLR approach achieves less end to end delay compared with the existing protocols.

2. Energy consumption

The amount of energy consumed for sending the data from source node to destination.

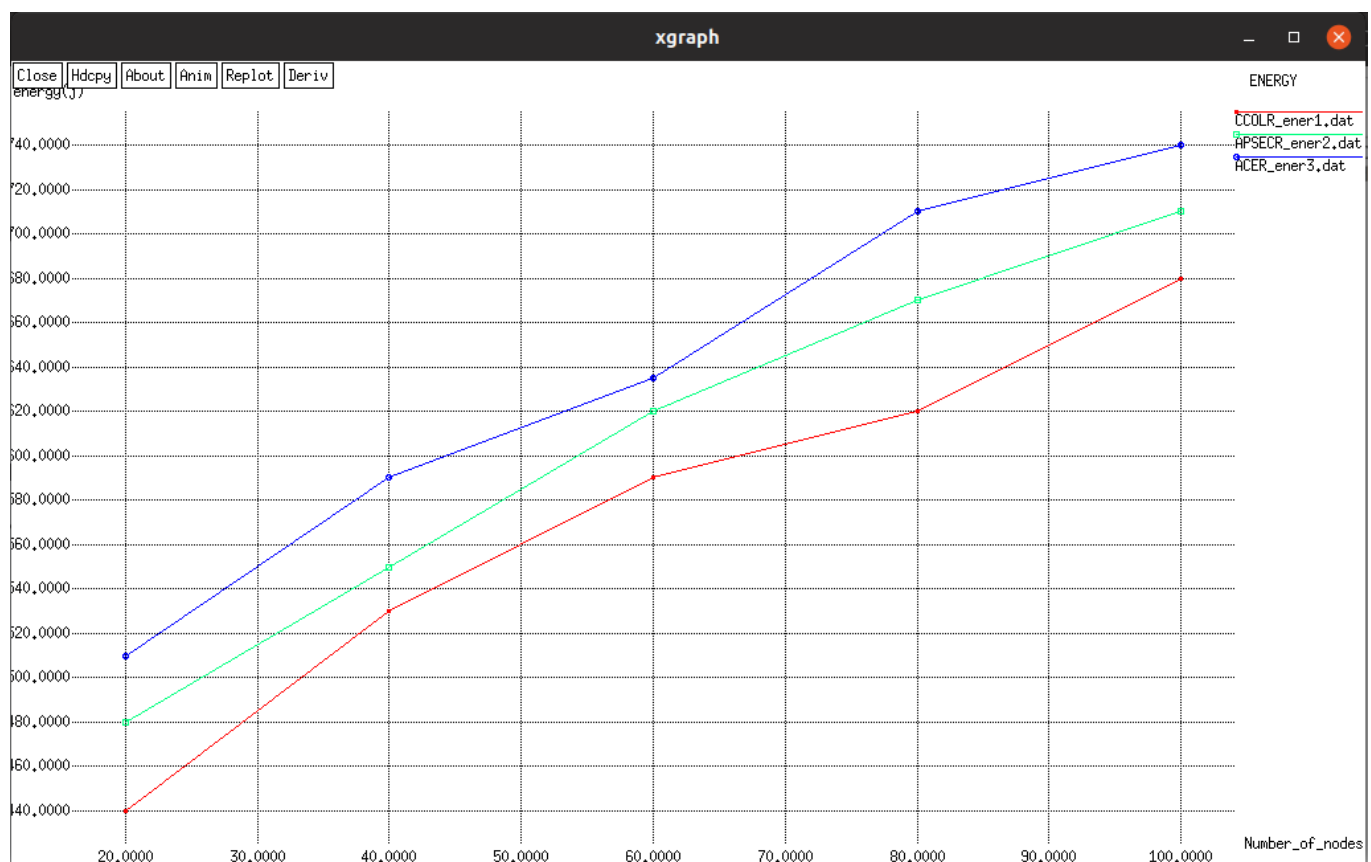


Fig. 3. Energy consumption comparison

The performance of the proposed Congestion Control and Optimal Link Based Routing (CCOLR) approach is compared with the existing Adaptive Energy Aware Cluster-based Routing (AEER) and Adaptive Path Selection and Energy aware Cluster-based Routing (APSECR) protocols in terms of energy consumption. In x-axis number of nodes is taken and energy consumption is taken as y axis. In this proposed research work, the nodes are grouped initially and an optimal cluster

heads are selected by using Multi Objective Anarchic Society Optimization algorithm (MOASO). And also an optimal path selection is performed based on the link cost with includes Extended Medium Time Metric (EMTM) and throughput for inter-cluster and intra-cluster communication. It reduces the energy consumption. From the experimental results, it concluded that the proposed CCOLR achieves better energy consumption compared with the

existing system.

3. Packet Delivery Ratio (PDR)

The ratio between the number of data packets

successfully delivered to the destination and the number of packets transmitted by the source.

$$PDR = \frac{\text{Number of packets received}}{\text{Number of packets transmitted}} \quad (25)$$

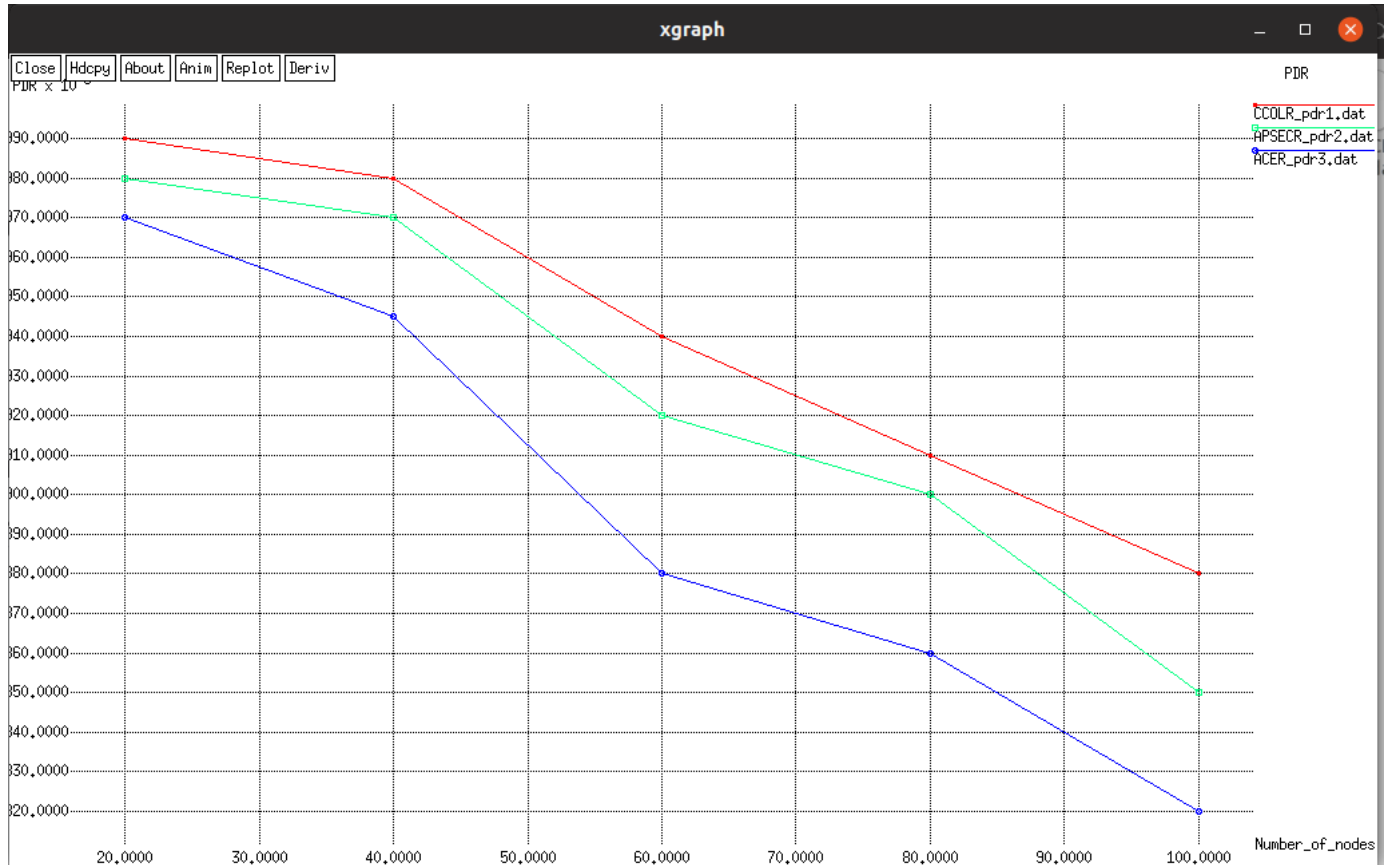


Fig. 4. Packet delivery ratio comparison

The Packet delivery ratio of the proposed Congestion Control and Optimal Link Based Routing (CCOLR) approach is compared with the existing Adaptive Energy Aware Cluster-based Routing (AEER) and Adaptive Path Selection and Energy aware Cluster-based Routing (APSECR) protocols which are shown in figure 4. In x-axis number of nodes is taken and packet delivery ratio is taken as y axis. The experimental results show that the proposed CCOLR approach achieves

higher packet delivery ratio compared with the existing system

4. Throughput

The rate in which the data packets are successfully transmitted over the network or communication links is defined as throughput. It is measured in bits per second (bit/s or bps). It is also specified by units of information processed over a given time slot.

$$\text{Throughput} = \frac{\text{Number of delivered packet} * \text{packet size}}{\text{Total duration of simulation}} \quad (26)$$

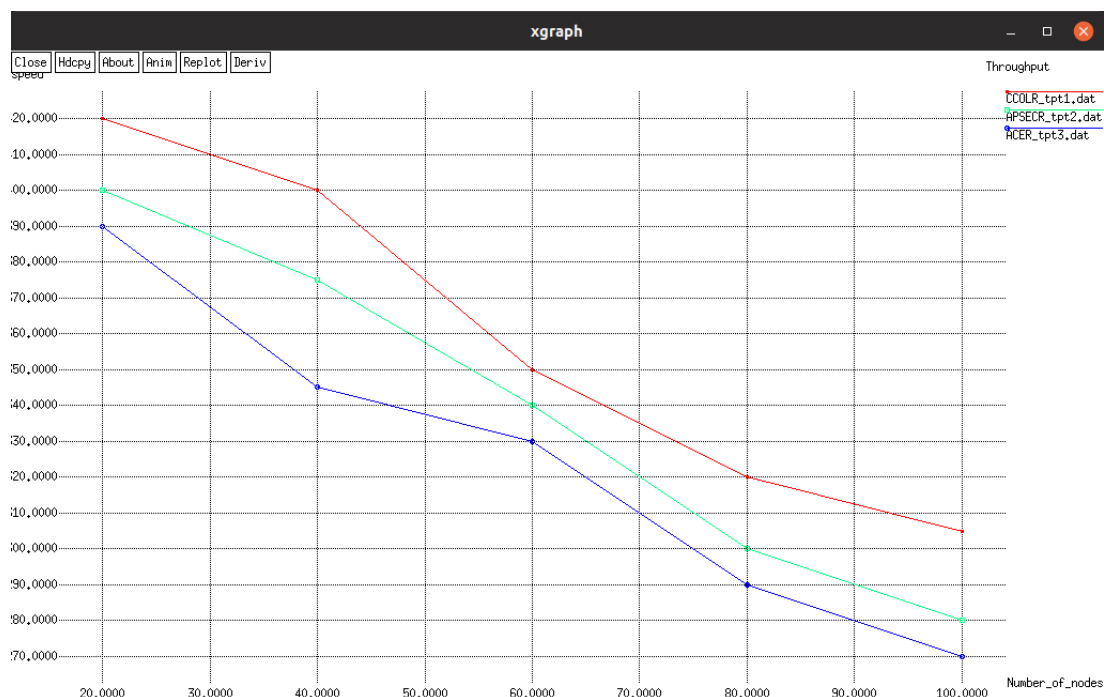


Fig. 5. Throughput comparison

Figure 4 shows the comparison of throughput performance for proposed Congestion Control and Optimal Link Based Routing (CCOLR) approach is compared with the existing Adaptive Energy Aware Cluster-based Routing (AEER) and Adaptive Path Selection and Energy aware Cluster-based Routing (APSECR) protocols. In X axis number of nodes are taken and in y axis throughput is taken. From the graph it is clear that the CCOLR approach provides higher throughput than existing method.

VI CONCLUSION

The proposed system designed a Congestion Control and Optimal Link Based Routing (CCOLR) for congestion control and maximizing the lifetime of WSN. In this work, optimal Cluster Head (CH) is selected by using Multi Objective Anarchic Society Optimization (MOASO) and the path selection is performed based on the link quality. Then, depending upon the channel consumption proportion as well as queue length as of the intermediary CH node, Jamming position along with estimation rates have been considered through the target nodes. The

origin node once receiving the response values as of the target execution speed control. The experimental outcome shows that the projected method attains improved execution than the actual structure with respect to end to end delay, packet delivery ratio, and output as well as energy utilization.

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