A Pragmatic Intervention Model For Network Consistency Evaluation In Wireless Networks

PALLAPATI SOLMON [#1], YEZARLA CHITTI BABU [#2]

[#1] II/II M.Tech (CSE), st.ann's college of engineering & technology, pallapatisolmon@gmail.com
[#2] Associate Professor, st.ann's college of engineering & technology,chitti510@gmail.com

ABSTRACT: Wireless sensor networks are originally designed as distributed event-based systems that differ from traditional communication networks in several ways. These networks typically have nodes with severe energy constraints, variable quality links, low data-rate and many-to-one event-to-sink flows. In distributed event-based systems, generally events occur infrequently. Some sensitive applications such as volcanic monitoring, fire detection data should be transmitted within a specified delay to the base station. In such systems, most of the energy is consumed when the radios are on, waiting for an arrival to occur. So, sleep-wake scheduling is implemented which is an effective mechanism to prolong the lifetime of these energy-constrained wireless sensor networks. However, sleepwake scheduling could result in substantial delays because a transmitting node needs to wait for its next-hop relay node to wake up. In this paper, to study the joint control problem of how to optimally control the sleep-wake schedule, the anycast candidate set of next-hop neighbors, and anycast priorities, to maximize the network lifetime subject to a constraint on the expected end-to-end delay. We provide an efficient solution to this joint control problem and analyze the end-to-end delay under anycast. We develop an optimal distributed anycast algorithm that minimizes the end-to-end delay of all nodes and solve the lifetime-maximization problem and it can be easily applied to energy-constrained event-driven wireless sensor networks.

Key Words: imaging & target tracking modules, frame measurement, Intraframe signal-to-interference-plus-noise ratio (SINR) distribution, link state, received signal strength indicator (RSSI), SINR, wireless Intervention Model.

I. INTRODUCTION

Sleep-wake scheduling is an effectivemechanism to prolong the lifetime of energy-constrained sensor networks. In this paper, we are interested in event-driven wireless sensor networks, where events occur occasionally. Therefore, by putting nodes to

Sleep when there are no events, the energy consumption of the sensor nodes can be significantly reduced.

In synchronized sleep-wake scheduling protocols, sensor nodes periodically or aperiodically exchange synchronization information with neighboring nodes. However, these synchronous protocols could incur additional communication overhead, and consume a considerable amount of energy. Here asynchronous sleep-wake scheduling protocols are preferred than others. In asynchronous sleep-wake scheduling protocols, the sleep-wake schedule at each node is independent of that of other nodes, and thus no synchronization is required. However, due to the lack of knowledge of the sleep-wake schedule of other nodes, it incurs additional delays for packet delivery when each node needs to wait for its next-hop node to wake up. But this delay could be unacceptable for some kind of applications like fire alarm etc.

Previous work has proposed the use of anycastto reduce event reporting delay. In anycast each sending node tries to wake up a group of neighboring nodes in a candidate set, and the sending node then picks the first node that wakes up to relay packets. Therefore, the delay to wake up the next-hop neighbors can be significantly reduced. On the other hand, the end-to-end delay not only depends on the

per-hop delay, but also the end-to-end path that packet traverses. Hence, the set of candidate nodes must be carefully chosen because it will also affect the possible routing paths.

In this paper, we directly optimize the system with respect to the end-to-end delay. In particular, we formulate the joint control problem of how to optimally control the sleep-wake schedule, the anycast candidate set of neighboring nodes, and anycast priorities among neighboring nodes, to maximize the network lifetime subject to a constraint on the end-to-end delay. We provide an efficient and appropriate solution to this joint control problem and show how to optimally choose the candidate set in order to minimize the end-to-end delay for all nodes.

II. BASIC MODEL

We consider a wireless sensor network with N nodes. Each sensor node is in charge of both detecting events and relaying packets. If a node detects an event, the node packs the event information into a packet, and delivers the packet to a sink s via multihop relaying.

With sleep-wake scheduling, nodes sleep for most of the time and occasionally wake up for a short period of time tactive. When a node ihas a packet for node j to relay, it will send a beacon signal followed by an ID signal (carrying sender information). Here, mainly four conditions will a raise. They are:

- When node j wakes up and senses a beacon signal, it keeps awake, waiting for the following ID signal to recognize the sender.
- When node j wakes up in the middle of an ID signal, it keeps awake, waiting for the next ID signal.
- If node j successfully recognizes the sender, and it is the next-hop node of node i, it then communicates with node i to receive the packet.
- If a node wakes up and does not sense any beacon signal or any ID signal, it will then go back to sleep.

Let C_i denote the set of nodes in the transmission range of node i. Suppose that node i has a packet, and it needs to pick up a node in C_i to relay the packet. Each node imaintains a list of nodes that node i intends to use as a forwarder. We call the set

of such nodes a forwarding set, which is denoted byFi.

A. Sleep-wake Schedule

The sleep-wake schedule is determined by the rate λj of the Poisson process with which each node j wakes up. If λj increases, the expected one-hop delay will decrease, and so will the end-to-end delay of any routing paths that pass though node j. However, it leads to higher energy consumption at node j so that the network lifetime may decrease.

B. Forwarding Set

The forwarding set Fi is the set of candidate nodes chosen to forward a packet at node i. In principle, the forwarding set should contain nodes that can quickly deliver the packet to the sink. However, since the end-to-end delay depends on the forwarding set of all nodes, choosing the correct forwarding set is not easy.

C. Priority

When multiple nodes send an acknowledgement after the same ID signal, the source node i needs to pick one of them as a forwarder. We assume that node i assigns priorities to all nodes in C_i, and will pick the node with the highest priority among these nodes that wake up. Clearly, the priority assignment will also affect the expected delay.

III. METRICS

The performance metrics that we are interested in:

End-to-End Delay: We assume that the end-to-end delay for event delivery is dominated by the cumulative sum of the delay for each hop to wake up and to relay a packet to its next-hop neighbor. We define the end-to-end delay as the delay incurred by the first packet, which is the sum of the delay for each hop to wake up and to relay the packet to its next-hop neighbor.

Network Lifetime: We assume that the network lifetime is determined by the shortest lifetime of all nodes. In other words, the network lifetime for a given awake probability vector p.

The objective of this paper is to choose awake probability vector p, forwarding matrix A, and priority matrix B to maximize the network lifetime, subject to the constraint that the expected delay from each node to sink s.

Examining Interference

In order to conduct a simulation studywith a general scenario,we built up a network with 16 randomly located nodes. Each node broadcasts frames at constant bit rate randomly chosen in [1, 4]Mb/s and continuously transmits frames on a random basis during the period of 150 s.

In this simulation, we used RCPI rather than RSSI in calculating SINR for each frame. We compared the following cases: 1) the relationships between the FDR and the SINR values of delivered frames (as [11] did); and 2) those relationships for detected frames (not to put weight on successfully received frames too much) every 1-s interval at each link. Fig. 1 presents those results for 200 or more transmitted frames: Fig. 1(a) is for the case 1) and Fig. 1(b) is for the case 2).ven though the average SINR value over detected frames seemed to be more correlated to FDR, any SINR value cannot correctly represent the link state (FDR). Specifically, we examined all received frames at one tagged node (node 13), which is presented in Fig. 2. We cannot see any difference between Figs. 1 and 2, and thus become more sure about it that the average INR is not appropriate for determining link state. Next, we investigated the relationship between two time intervals, time interval 1 and time interval 2, at the same link. Fig. 3
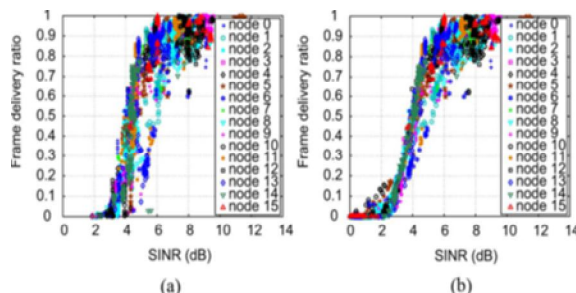
Fig. 1. Relationship between the FDR and the average SINR values of delivered and (b) detected frames at all links.
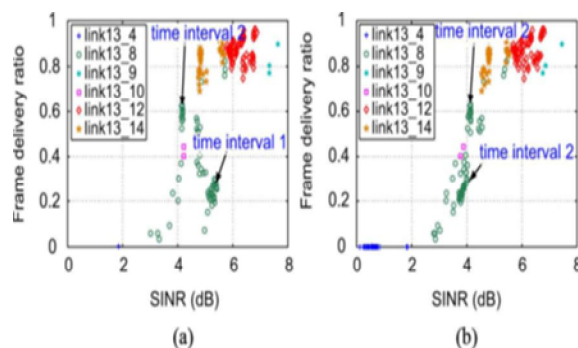
Fig. 2. Relationship between the FDR and the average SINR values of (a) delivered and (b) detected frames at receiver node 13.
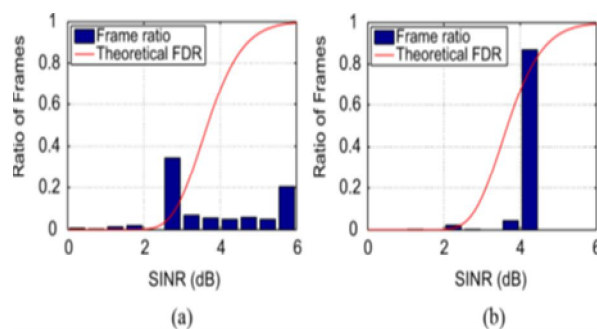
Fig. 3. SINR distributions of received frames during (a) time interval 1 and b) time interval 2 at link13_8 in Fig. 2.

shows the ratio of received frames to transmitted frames at each SINR value, and also the theoretical FDR (that comes from Fig. 4). In time interval 1, fewer frames are likely to be delivered, but the average SINR of the frames is higher than that of frames that are likely to be delivered in time interval 2. That is why the higher average SINR can produce the lower FDR, as we often found in Roofnet links, which has been also discussed in Section II-B. From those simulations,we conclude that the link state cannot be simply determined with its average SINR value.

IV. MINIMIZATION OF END-TO-END DELAYS

In this section, we consider how each node should choose its forwarding set and assign priorities to neighboring nodes to minimize the delay Di(p,A,B). We first derive a recursive relationship for the delay, Di(p,A,B) where awake probability vector

p, forwarding matrix A, and priority matrix B. At each iteration, each node uses the delay estimates from the previous iteration to update the forwarding set and the priority assignment. We will show that the algorithm converges in N iterations, and the resulting A and B minimize the expected delay Di(p,A,B). Steps for OPT-DELAY algorithm are:

Step (1) At iteration 0, each node isets: if i= s then Di(0) = 0, otherwise Di(0) = ∞, and Fi(0) = θ. Each node arbitrarily assigns priorities to neighboring nodes.

Step (2) At iteration h (≥ 1), each node isets b i(h)= b i *(π i

(h−1)   (h−1)   (h−1)

), where π i   = (Dj   , j ∈ Ci).

Step (3) Each node iupdates Fi(h) by finding the optimal forwarding set for π i(h−1) and also updates Di(h) as follows Di

(h)   (h−1)   (h)

= f′(π i   , Fi   )

Step (4) If Di(h)= Di(h-1)for all nodes i ∈ N, this algorithm terminates. Otherwise, each node increases h by one and goes back to Step (2).

Where, We call the function f() is the local delay function, b i

* Isoptimal priority assignment and πi is neighboring delay vector. Here, in order to minimize f(), the optimal priority assignment b i*can be completely determined by the neighboring delay vector πi.

V.   LIFETIME-MAXIMIZATION

In the previous section, we solved the delay-minimization problem. In this section, we use the result to develop a solution to the lifetime-maximization problem. We develop an efficient binary search algorithm for computing the optimal value.

Step (1) Initially, sink s sets p(1) = 0.5 and k = 1.

Step (2) Sink s sets q(k) = ln(1 − p(k))−maxϵNei

Step (3) Nodes run the OPT-DELAY algorithm for given

$$p^{(k)} = (p_i^{(k)} = 1 - e^{-q^{(k)}/\epsilon_i}, i \in \mathcal{N})$$

Step (4) After N iterations, the optimal forwarding set and the optimal priority assignment under p (k) are found. Nodes j that are not in the other node's forwarding set, i.e., j   F* i(Aϵ(p(k))) for all nodes i, send feedback of their delays Dj(p(k),A*(p(k)),B*(p(k))) to sink s.

Step (5) Let Dmax be the maximum feedback delay arrived at sink s.

•   If Dmax> ξ*+ϵ, then sink s sets p(k+1) = p(k)+0.5k+1, increases k by one, and goes back to Step (2).

•   If Dmax< ξ*−ϵ, then sink s sets p(k+1) = p(k)−0.5k+1, increases k by one, and goes back to Step (2).

•   If Dmaxϵ [ξ* −ϵ, ξ* +ϵ], then the algorithm terminates, and returns q(k) as the optimal solution.

The reason that we take q(k) with respect to the maximum eiin Step (2) is because this makes all p(k) iless than or equal to p(k). (Note that we only search p(k) over (0, 1].) InStep (4), only such a node j that does not belong to any other forwarding set needs to send the feedback delay to the sink s because the node with the maximum delay does not belong to any other forwarding set. Since sink s only needs to know the maximum delay, there is no need for the other nodes to feedback their delays.

CONCLUSION

In this paper, we study how to use anycast to reduce the end-to-end delay and to prolong the lifetime of wireless sensor networks employing asynchronous sleep-wake scheduling. In particular, we study the joint control problem of how to optimally control the sleep-wake schedule, the anycast candidate set of next-hop neighbors, and the anycast priorities, in order to maximize the network lifetime subject to a upper limit on the expected end-

to-end delay. We provide an efficient solution to this joint control problem, and as a part of the solution, we also show how to optimally choose the anycast candidate set to minimize the end-to-end delay from all sensor nodes. The algorithms that we have developed can be easily applied to energy-constrained eventdriven wireless sensor networks.

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PALLAPATI SOLMON **is** a student of Computer Science & Engineering from ST.ANN'S COLLEGE OF ENGINEERING & TECHNOLOGY, CHIRALA. Presently pursuing M.Tech (CSE) from this college. He received B.Tech from JNTUA in the year Of 2011.

YEZARLA CHITTI BABU is a Associate Professor of ST.ANN'S COLLEGE OF ENGINEERING & TECHNOLOGY, CHIRALA. .He received M.Tech in Computer Science & Engineering from JNTU,Kakinada University and Presently pursuing PhD in Computer Science at Acharya NagarjunaUniversity. He has presented nearly 8 International conferences .He gained 10years Experience on Teaching . He is a good Researcher in Computer Networks and Network Security.