# **RESEARCH ARTICLE**

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# A Reliable Transport Protocol for Code Distribution in Large Sensor Networks

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ABSTRACT—Re-tasking and remote programming of sensor networks is an essential functionality to make these effective. networkspractical and As the availability more capable of sensornodesincreases and new functional implementations continue to be proposed, these large collections of wireless no sensor of the sensordeswillneed the ability to update and upgrade the software packagesthey are running. Standard flooding mechanisms are too energy-costlyandcomputationally expensive and they may interfere with the network's current tasks. A reliable method for distributingnewcodeorbinaryfilestoeverynodeinawirelesssensornetwork is needed. This paper proposes a more effective method, called PALER (Push Aggressively with Lazy Error Recovery), which builds upon the previously proposed **PSFO** protocol [1], are liable transport protocol which slowly paces the propagation of files egments, but uses an aggressive local recovery methods and the standard standardhod to avoid packet implosion due to loss propagation. PALER uses a more aggressive pushing mechanism and reduces the standard standardrecovery mechanism to a single inclusive NACK. Furthermore, PALER uses local neighbor information to reduce redundanttransmissions. Thispaperstudies this new protocol's energy efficiency and shows that it scales well to higher densities and fieldsizes.

### I. INTRODUCTION

Wirelesssensornetworksareacollectionofs mallsensordevices,typicallybatterypowered,thatma ybestaticormobile,andmayconfigureinanadhocfashi on.Theyhaveinspiredagreatdealofresearchintothede signandimplemen-

tationofsuchnetworks, such as routing mechanisms, ev enttriggering, and clustering methods, as well as metho dsof improving energy efficiency with radio power cont rolmecha-nisms and adaptive functionality. Re-

taskingsensornetworkswhichhavealreadybeendeplo yedisstillanopenareaofresearch. It requires the reliable distribution of a binary file orcode toallof thenodeswhichrequire thenewprogram, andacoordinatedmethodofloadingthenewprograma mongallofthe nodes. This can be an expensive task, requiring

significantbatterypowertofulfillallnecessaryradiotra nsmissions, and consuming the computational resources of the wireless nodes.Inthispaper,wewillfocusonthetaskoffiledistri butionin support of remote programming. There been have numberofproposalsforsuchafunctionality.Manyofth esehaveworkedundertheassumptionofasmallbinaryo rcodefile, requiring a limited number of segments to distributed. be

Thisconditionwastypicalofearlywirelesssensormote s,whichhavelimitedmemorycapacity,butasthecapabi lityofsensormotesincreases,sensornetworkprograms willinevitablyincrease with them. In this paper, we will explore a moreefficient method of binary or

distribution code for relativelylargedatafiles. Themostbasic method offiled istributionis by flooding the segments into the sensor network. Thismethod is very costly, though, everv node will receive as andbroadcasteachsegmentofafile.Manyofthesebroa dcastsareunnecessary since sensor networks may be dense, and havehighly overlapping broadcast zones. The redundant transmis-sions are an unnecessary expenditure of power and could leadto increased packet loss due to congestion. Another issue

with the naiveflooding method is that it provides no reliability. It is important to ensure that all no desinthenetwork receive the entire file in a timely manner, so reliability mechanisms will be needed.

Wan, et al. proposed Pump Slowly, Fetch Quickly (PSFQ)[1], a reliable transport protocol which reduces transmissions, making it energy efficient and scalable. The basic premisebehindtheprotocolistopropagatethesegment satarelativelyslow pace ("pump slowly"), and use an aggressive NACKmechanism to fetch missed segments ("fetch quickly"). Thisprotocolattemptstominimizethecostoflostpacke tsbyreducingNACKsandrebroadcastsinresponsetoN ACKstoa single hop. They also reduce the amount of transmissionsthrough forwarding by using a counter mechanism to limittransmissions. These methods are effective in reducing thetotal number of transmissions necessary to flood all segmentsof a file throughout a sensor network, however, the aggressivetiming of the fetch operation can lead to congestion in a

densenetworkwhennodesattempttorespondtoNACK s.Also,the retransmission delay that is added to allow for handlingNACKs places a lower bound on latency. We the proposePushAggressivelywithLazyErrorRecovery( PALER), which is based on PSFO with the enhancement that NACKs arereserveduntiltheendofthetransmissionofallsegme nts.ThisreducestheamountofNACKpackets,particul arlyinhigh-lossenvironments. A slower response period reduces collisions inNACK responses, and further reduces the number of totaltransmissions complete for dissemination. This modificationalsoeliminatestheneedforaretransmissi ondelay, allowing amore aggressive forwarding metho d, leading to reduced latency.

PSFQandPALERhavebeenimplementedandevaluat ed through simulation to exhibit the effects of node density andhop count on total transmissions and latency. These resultsshowthatamorerelaxedrecoverymechanismav oidsmuchof the contention and collision found in a

more aggressivemechanism.Italsoprovidesanopportunityf orimproved collaboration among nodes to avoid redundant responses. Inaddition, withholding negative acknowledgementsu ntilallsegments of a dataset have been broadcast allows a moreaggressive propagation of segments. This allows PALER to improve energy efficiency by reducing transmissions whilereducing latency. The results show that PALER is capable offunctioning efficiently in very dense networks, and latencyscales very well across an increasing field size. We comparetwo variations of PALER to evaluate the advantages of differingpruningtechniquesinthefloodingphase.VersionIu tilizesa counter method to reduce redundant

transmissions, whileversion II utilizes local neighbor data to prune unnecessarytransmissions.

## II. RELATED WORK

Code distribution has many similarities with reliable mul-ticast, reliable broadcasting and energy-efficent broadcastingresearch. Many traditional techniques for making multicastreliable in wired networks, such as [2], are too computation-ally expensive for the limited resources of a wireless node.Multicast protocols which have been developed for wirelessnetworks, such as [3] and [4], tend to favor robustness toprovidereliability.Thisbuilt-

inredundancycomesattheexpense of energy efficiency. Wireless broadcasting protocolsfocused on reliability typically require significant overhead incontrolpackets[5,6]. Energy-efficient wireless broadcast protocols typically focusintwoareas:minimizingforwardingnodes[7,8,9,1 0],

or minimizing transmission power [11, 12, 13]. [7] offersseveral methods of reducing forwarding nodes, such as prob-abilistic, counter, and cluster methods. each with differinglevelsofreliability.[10]developedanalgorith mtoidentifya dominating set in a network, which would make up theintermediatenodesin allbroadcasts.[9] and[8] builtuponthismethodbyreducingthesizeofthedomina tingset.Selection of a dominating set may reduce total

transmissionsinabroadcast, however, it does not balanc etheload.asthenodesinthedominatingsetwillincurall ofthecost.Aninterestingextensionof[9]isaneighborel iminationmethodwhichisverysimilartothemethodus edinPALERto reduce forwarding nodes. [14] also used an approximation of minimum dominating set to achieve reliability and reducethe cost of scheme recovery. Their is capable of providingreliable broadcast of a single packet, and limits the cost ofdropped packets with local recovery, but the focus of thisworkis onreliability ratherthantheenergyefficiencyofthe downstream Broadcast propagation. protocols aimed atminimizing transmission power are typically based upon а setcover[12]orminimumspanning[11,13]problem.T hesetendtoprovideveryefficientdistributiontreesand balanceloads evenly; however, they require a knowledge of the completenetwork topology, and are best used in а static environmentwhereoptimalroutescanbepredetermine d.

Others methods of achieving reliability or efficiency includeFECandnetworkcodingtechniques.FECisusedi n conjunction with probabilistic forwarding by [15] to addreliability to an efficiency scheme. The FEC technique

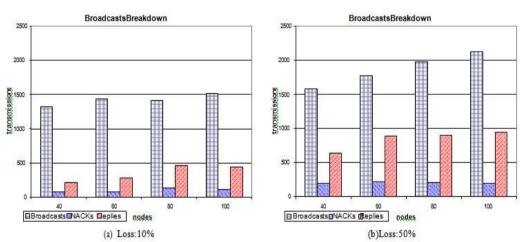
offersflexibilityinthepropagationandrecoverymetho ds.Withalargedatadistribution,thenumberofoverhea dpacketsgeneratedbytheFECisexpectedtobeminimal relativetothe total packets sent, but this assumes a lossless environmentand ideal MAC laver. A realistic scenario could require anundesirablenumberofoverheadpacketsproducedb ytheencoding method to provide reliability. The forwarding prob-ability used for their simulations is optimized for the networktopology, which provided efficient propagation results. Thiswould make implementations sensitive to alterations in thenetwork size and density, however, which could lead to a lossof efficiency or reliability. Network coding is used in [16] toimprove efficiency by reducing the number of transmissions. They provide an alternative algorithm in their approach formulticastinwirelessnetworkstoexploitthewireless multicastadvantage. Whilethenumberoftransmission smaybereducedin this method, the potential overhead cost of the control datawhich must be appended to each packet is not evaluated. Forlarge data transmissions, as may be used for re-tasking, thiscost could be significant. Extensions would also need to

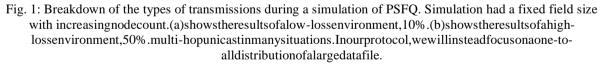
bemadetothisschemetoprovidereliabilityandlossrec overy.

Trickle [17] is a popular recent method of code propagationand maintenance. It uses a gossiping protocol with periodicmetadata broadcasts to identify nodes which require an updatetoanewversionofcode.Itusesacountermethodt olimitthe number of gossip messages broadcast during an interval, which makes Trickle a very of energy efficient method maintainingasensornetwork.Trickleisnotgreatlyconcerne dwithlatency, and is based upon an expectation of a very smallcodesegmentorbinaryfile, onewhich will fit in as

mall number of packets. Our protocol will be more focusedonanefficientmethodforasinglereprogrammi ngeventof a relatively large binary file. Another gossip based codepropagation protocol is GCP [18]. It uses periodic beacons todetect outdated code versions, similar to Trickle. However, italso includes a forwarding control mechanism to balance theload of distribution. Each node has a limited number of tokensthatitmayusefordistributingeachnewversiono fcode.

Melete [19] builds upon Trickle to support dynamic group-ing and concurrent applications in sensor networks. It uses aperiodic metadata broadcast to maintain the network. It alsosupports group-based code propagation. Nodes may dynami-cally enter and exit a group, and must broadcast a request forthe new group code. Melete avoids broadcast implosion bypacing requests through a probabilistic and progressive flood-ing mechanism. Because of the dynamic grouping nature oftheirsystem,codepropagationwillbeaccomplishedt hrough





# **III. PROTOCOL DESIGN**

# A. PSFQDesign

PSFQ [1] distributes data from a single source to a networkof sinks by slowly pacing the propagation of packets. Nodesthat detect a lost packet due to out-of-order packet receptionwillaggressivelyfetchthemissingpacketbys endingaNACKto its immediate neighbors. Nodes will refrain from forward-ing packets received outof-order until all packets leading upthat one have been recovered.In other words, regardless of the order of reception, nodes will only rebroadcast packetsin-

order. This prevents the propagation of a loss event to the

key to PSFQ is the relation between  $T_r$  and  $T_{max}$ . The ratiobetween  $T_r$  and  $T_{max}$  determines how many opportunities anodehastofetchamissingsegmentbeforethenextse gmentisexpectedtoarrive. The higher the ratio  $T_{max}/T_r$ the greater the probability that the segment will be successfully received over multiple hops. In their implementation, they selected  $T_{max} = 5$   $T_r$ , and  $T_{min} = {}^{1}T_{rkax}$ .  $T_r$  must also be chosen to provide an adequate window for recovery. To provide aminimum cessary window,  $T_r$  should be at least fo urtimes the latency for a single packet, but they proposed that areasonable value would be  $T_r = 6 * T_p$ .

WhenanodereceivesaNACK, itcheckstherequestagai

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nstits own cache to determine if it has any of the

missing segments. If it does, it will schedule a reply with the missing segment at a random time with in the interval  ${}^{1}T_{r}$  to  ${}^{1}T_{r}$ . To

downstreamnodes.Requiringin-

orderbroadcastsalsoensures 4

that lost packets can be retrieved from at least one immediateneighbor, since the neighbor that broadcast the higher thanexpected sequence number must also contain the expected sequence number. By localizing recovery, it reduces lost recovery cost by suppressing the propagation of loss events

andnegativeacknowledgements,andreducingrecover ytoasinglehop transmission. The PSFQ protocol is built upon a tightlycoupled timing between the pushing mechanisms and fetchingmechanisms. The pumping mechanism relies on two timers, $T_{min}$  and  $T_{max}$ . After reception of each in-order packet, thepacket will<sup>×</sup> be scheduled for re-broadcast following a randomdelaybetween $T_{min}$  and  $T_{max}$ . A counteris

maintainedforthenumberoftimeseachpacketisheard. Ifapacketisreceived four times prior to rebroadcasting, the

rebroadcasteventiscanceledtosuppressredundanttran smissions.

Ifapacketisreceivedout-of-

order, then ode will schedule a NACK with a request for the missing segments after a shortrandom delay. The node will continue to send a NACK everyT<sub>r</sub>untilallofthemissingsegmentshavebeenrec eived.The reduce contention and redundant transmissions, it will cancelthe reply event if it overhears any neighbors responding to thesame missing segment prior to its own reply. To improve thelikelihood that only one neighbor will respond to a NACKrequest, each node maintains a table with the average signalstrengthsofitsparentnodes.WhenitsendsaNAC K, it includes its preferred neighbor (node with the highest averagesignal strength) in the header. The neighboring nodes willchecktodetermineifitisthepreferredneighbor; ifn

ot,it will double it's delay interval before sending the reply,

givingitmoretimetooverhearareplyfromthepreferred neighbor.

The slow pumping mechanism of PSFQ along with thecountermethod for pruningof forwarding nodesis effectiveinreducingcontentionandcollision,andeffici

entlyprop-\_\_\_\_agatingafilewithminimalredundanttransmissions.Th efetching mechanism avoids the NACK implosion problem bylocalizing recovery. However, the aggressive nature of thefetch mechanism and the relatively short time frame of therecoveryintervalsleadstoagreatdealofcontention. While the neighboring nodes listen for duplicate replies from

otherneighbors, the response time frame does not allow a large window to randomly disperse the responses and to overhear responses. Each node is generating a random delay period between  ${}^{1}T_{r}$  to  ${}^{1}T_{r}$ . This interval equates to just  ${}^{3}T_{p}$ , which

iscanceled.BecauseNACKsarewithhelduntilthefinal segment, intermediate nodes do not have to have a minimumdelayperiodpriortorebroadcasting, sotheint ervalcanbeanyrandomperiodwithin100ms.Sincethec ounterlimitstherebroadcastwithinaregiontothefirstth reeattempts, the

provideslittleopportunityforoverhearing. The inclusi onofa preferred neighbor based on received signal strength

helpsincreasethewindowofopportunityforoverheari ngotherresponses, but nodes that computed a small random delay willstill be transmitting close to the same time as the preferrednode.Inadensenetwork,thismayresultinahi ghnumber of collisions, possibly resulting in more redundant responses.Figure 1 shows a breakdown of the types of transmissionsobservedinasimulationofPSFO.Inthiss imulation.thefieldissetto1km

1km, and the density ranges from 40 to 100 nodes perkm<sup>2</sup> . The first graph shows the results for an average 10% packet loss, the second graph depicts the same results for a 50% loss environment. These figures show that the number of response messages are a significant percentage of the total transmissions. Many of these reply transmissions are redundant, and could be avoided with a less aggressive method. Additionally, in a high loss environment, the in-order requirement can result in a high number of NACKt ransmissions. Another impact of the timing constraints is

that it places allower bound on the latency that is equivale  $ntto T_{min}$  hops, where hops is the maximum number of hops needed to reach all nodes from the source.

## B. PALERDesign

To avoid the contention and collisions resulting from theaggressive recovery mechanism PSFO. PALER eliminatestheinof orderreceptionrequirementandmaintainsalistofmissi ng segments. After all packets have been broadcast, eachnode will broadcast a NACK to its neighbors. An important spect of PALER is that, even though it does not require in-order reception, it still maintains local а recoverv mechanism. There as on this is possible is because if an ei ghborreceivesa NACK, it can first check its own cache for the missingsegments. Any segments that present the local are in cachemaybesenttotherequesterinasinglehoptransmis sion.As for segments that are not present in the local cache, these segments must also be among the list of missing segments on he local node, which means that they will be included in the local nodes NACK. Therefore, it does not need to propagate the NACK from its neighbor, because the missing segments will be redundant. This maintains local recovery with singlehopNACKtransmissions, while still ensuring th atrequestfor missing segments will be propagated until the missingsegmentsarefound.

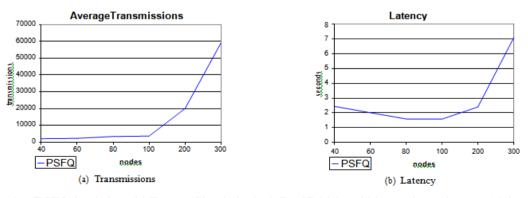
Whenanodereceivesasegment, it checks if th esegmentisalreadycontainedinitscache; ifitis, it incre mentsacounter for that segment. If it is the first reception of thissegment, it schedules a forwarding event at a random delaybetween 0 and 100 ms.If the counter for the segment reachesthreebeforethesegmentisrebroadcast, the forw ardingeventaverageintervalpriortoarebroadcasttends totrendcloserto0 than 100 ms, particularly in dense networks. This eliminatesthelowerboundonlatency, and provides am uchimprovedlatencythatscaleswellacrosslargemultihopenvironments.Whenanodereceivesthelastsegme ntofadistribution, itschedules abroadcast of a NACK to

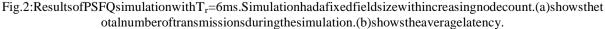
hopneighborswhichincludesalistofsegmentsitismiss ing.TheNACKisscheduledfollowingarandomdelayp eriod,whichisusedtoreducecollisions,andtoallowthe nodeanopportunitytooverhearrebroadcastedsegment sinresponsetootherNACKs.Ifitoverhearsasegmentth atitwasmissingpriortobroadcastingitsNACK,itwillre movethatsegmentnumberfromitslistofmissingsegme nts.WhenaneighboringnodereceivesaNACK,itcheck sthelistofmissingsegmentsagainstitsowncache.Ifito wnsanyofthemissingsegments,itwillscheduleareply withthemissingsegmentfollowingarandomdelaybet ween0to50ms.Ifitownsmorethanonesegment,itwills cheduleeachadditionalsegmentforareplyat10msinter vals.Thedelaypriortorepliesreducescontentionandall owsthenodestooverhearotherreplies.Ifanodeoverhea rsareplyforasegmentthatithasscheduled areplyfor,itwillcancelitsreplyevent.

Since this recovery mechanism is dependent upon receptionofthelastsegment.atimeoutperiodisusedin caseoflossofthe last segment. The timeout period is continuously updatedto be T<sub>begin</sub>+ a  $(T_{avg}[Seg_{tot}Seg_{rec}])$ . Here,  $T_{begin}$  is the time the first segment was received, T<sub>avg</sub>is the averageinterval between reception of each segment, Seg<sub>tot</sub> is thetotal number of segments, Seg<sub>rec</sub> is the number of segments received, and  $\alpha 1$ is a small multiplier that determines how aggressive the timeout value should be, typically this wouldbe less than 1.5. Following the transmission of each NACK, anew timeout value is set to specify maximum time а expectedbeforeallmissingpacketsarerecovered.Ifan ysegmentsarestillmissingattheendofthistimeoutper iod,anotherNACKwillbetransmitted,andanewtime outvaluewillbeset.Thefirst timeout following a NACK will be calculated using thesame formula above. If additional NACK's as are necessary, each additional timeout period will be doubled to avoid aNACK implosion resulting from downstream nodes waitingforamultihoprecoverytopropagate.

#### C. PALERFloodingMechanism

With the lazy error recovery of PALER, NACK and re-covery transmissions are greatly reduced, leading to a veryefficient broadcast with minimal wasted energy. However, thepushing operation still requires a relatively large number oftransmissions, even in fairly dense networks, and many ofthese transmissions are redundant. PSFQ increased efficiencyoverstandardfloodingbyusingacounterwit hacutoffof4to





itsone-

locally limit the number of rebroadcasts within a region.

[7]showedthatifanodeoverhearsamessage4times,the averageadditional coverage that can be achieved by performing itsown rebroadcast of the message is just 5%. The additionalcoveragethatcanbeexpectedafteroverheari ngamessage3 times was shownto beabout9%. In PALER,a

counterwithacutoffof3isusedtolimitredundanttrans missions.Theselectionof3asthecutoffprovedtoreduc eforward-ing transmissions without significantly affecting NACKs orrecoverytransmissions.

Tofurtherimprovetheefficiencyofthepropagationmet hod, a new mechanism was devised that used twohopneighbor information that is dynamically acquired. To gener-ate a representation of each node's two-hop neighborhood, each node will include a list of its immedi ateneighbors with the first n segments of a broadcast. The value n is animplementation set value that will determine the strength of the neighborhood representation; in our implementat ionnis

10. When a node receives a segment with this list in

theheader, it will add pairing softhesending node with ea chofitsneighbors to a table that list all of the immediate neighbors ofnodes within its two-hop neighborhood. To reduce the amount f contention and collision during the propagation of the firstnsegments, particularly in dense networks, a probabilisticmethod can be used to limit the number of rebroadcasts. Aprobabilistic method is used instead of a counter, because acounter could result in the same nodes rebroadcasting eachtime, resulting in an incomplete neighborhood representation. The probability selected should be high enough to ensure that each node broad cast its neighborhoodinformationatleast once, which is implementationdependentonn. It canbe determined by using cumulative probability а distributionfunctiontodeterminewhatvalueofpisneed edforthedesiredconfidencelevel.

Following the initial n segments, when a node receives asegment for the first time, it generates a list of its neighborsandschedulesarebroadcasteventatarandom timeasbefore

Thelistofneighborsrepresentsalloftheimme

diateneighborsthat may need this current segment. each transmission For ofthatsegmentitreceivespriortorebroadcasting,itrem ovesthe sending node from the list of neighbors associated withthat packet. It then acquires the list of the sending node'simmediate neighbors from its neighbor table. and removeseachofthesenodes(ifpresent)fromitslistofne ighborsfor this segment. These are removed because it is ahle toassumethateachofthesenodesreceivedthesegmentf romthesendingnode, and therefored on otneed this seg ment.Ifthelistofneighborsforasegmentbecomesempt ypriorto forwarding the segment, the rebroadcast event is canceled, sinceit is assumed that all neighbors received t hesegment.

### **IV. SIMULATION METHOD**

To evaluate PALER, it was implemented in the Jist/Swanssimulation environment [20, 21], a scalable wireless ad hocnetwork simulator based in Java. PSFQ was also implemented for comparison. provides a full representation Swans of thecomplete network layer model, with accurate representations of a wireless environment, including path loss. environmentalnoise and collision interference. Each node in the simulationwasimplementedwithan802.11radio,with arangeofapproximately250meters.Twosetsofsimula tionswereperformed, one with a fixed field size of 1 km1 km, withan increasing density, and another with a fixed density and increasing field size. For each simulation, a file of size e50KB was distributed. The file was segmented into 50

chunks,each1KB.Theprimarymetricsusedtoevaluate theprotocolsare total transmissions and latency, where latency is

measuredasthetimeneededforallnodestoreceiveever ysegmentof the data file. The goal of PALER is to improve on

theefficiencyofPSFQbyreducingtransmissions,whil estillmeetingorexceedingthelatencyperformanceofP SFQ.Inou simulation results, we show that this was accomplished.Our evaluation of PALER will show that energy

efficiencycanbefurtherimprovedusingdynamic,local neighborhood

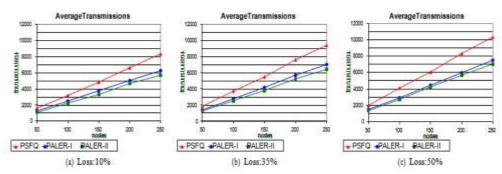


Fig. 3: Comparison of total transmissions of PSFQ, PALER-I and PALER-II, measured across increasing field size. Density isfixedat50nodesperkm<sup>2</sup>.A50KBfileistransmittedin1KBsegments.(a)showstheresultsforalow-lossenvironment,10%.(b)showstheresultsfor35%loss,and(c)fora50%loss.PALER-IandPALER-IIexhibitasignificantdecreaseinthe number of total transmissions required to complete a code distribution, with PALER-II demonstrating the best energyefficiency.

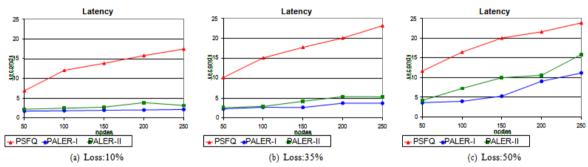


Fig. 4: Comparison of average latency of PSFQ, PALER-I and PALER-II with a fixed density and increasing field size.

TheselatenciescorrespondtothesimulationresultsinF igure3.PALERshowsanabilitytoscaleextremelywell acrossmultiplehops,scalingbelowalinearrateofincrea se.

information, though this may slightly impact latency. F or comparison, we have implemented both flooding mechanisms in PALER. The results for the PALER implementation using the counter method are presented as PALER-I, and the results for the implementation utilizing the neighbor pruning method are presented as PALER-II.

To determine the timing parameters used for the **PSFO** implementation, at ests cenario was implemented in Jist/S wansto determine the average time for a packet transmission with the range of densities used in our simulations. The≈calibrationresults showed that T<sub>p</sub>1ms. Based on the recommendation in [1], T<sub>r</sub>was set to 6T<sub>p</sub>. Therefore, in our initial imple-mentation of PSFQ, the timing parameters were  $T_r$ = 15ms, and  $T_{max} =$ The  $6ms, T_{min} =$ 30ms. simulation results forthis implementation exhibited a very poor energy efficiency, and did not scale well with high noded ensitie s.Theresultsof these simulations for a low loss

environment are shown inFigure 2. The simulations were implemented with a field sizeof 1km1km, with an average packet loss of 10% and 40 to300randomlyplacednodes,eachwitharangeof250m .Even

at40nodes,thenumberoftransmissionsisabove2000, which is what would result from a basic flooding mechanism. It canbeseenthatasthedensityincreasesabove100perkm <sup>2</sup>,PSFQdoes not scale well, and the number of transmissions quicklyescalatestomultipletimesthevalueofflooding. Therefore,to provide a stronger basis for comparison, the timing metricswerechangedtothoseusedinWanetal.'simple mentation[1].For their implementation, the timing metrics were  $T_r =$ 20ms, T<sub>min</sub>=50ms, T<sub>max</sub>=100ms. This provided agrea tlyimproved energy efficiency with transmissions an order of magnitude smaller, though the latency did increase as a resultofthelowerboundimposed by  $T_{min}$ .

### V. RESULTS

The first simulation performed a

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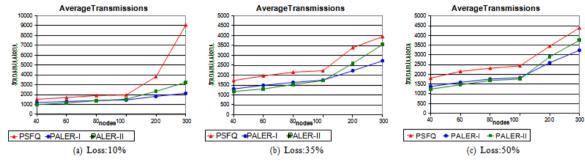
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comparison

for varyingfieldsizeofanetwork. The density for these sim ulationsis fixed at 50 nodes per km<sup>2</sup>. The number of nodes in

thesimulationrangesfrom50to250, which correspond

stoafield size ranging from 1km 1kmto 2.236km 2.236km(orawidthof4hopstoawidthof9hops).Theres ultsin



#### Fig.5:ComparisonoftotaltransmissionsofPSFQ,PALER-IandPALER-

 $II measured again stincreasing density. The field is fixed at 1\,km, each node has a range of 250 m. A 50 KB file is transmitted at 1\,km, and 1\,k$ edin1KBsegments.(a)shows the results for a low-loss environment, 10%. (b) shows the results for 35% loss, and (c) for a 50% loss. The benefits of PALER-II aregreatestinnetworkswithanaveragenodeconnectivitylessthan20.

Figure 3 show the average number of transmissions, measuredby the network size, for an average loss environment of 10%,35% and 50%. The graphs show that PALER-I exhibits a 20-30% reduction intransmissions from PSFO, while PAL ER-Ilexhibitsa30-

35% reduction. The increase in efficiency is fairly consi stentacrossdifferinglossenvironmentsanddifferingfi eldsizes.

The second set of graphs in Figure 4 show the correspond-ing latencies for the same set of simulations. Due to its lowerbound on latency, PSFO scales fairly linearly with field size(number of hops). However, it does perform rather consistently among increasing loss environments. As seen, PALERscales very well across increasing particularly inmoderate field sizes. loss environments. In high environments, loss suchasthatofFigure4c,latencyincreasesinstrongerrel ationtofieldsize, butstillperformsstrongly incomparis ontoPSFO.PALER-

Iexhibitedthebestlatency, revealing the tradeoff of ener gyefficiencyandlatencybetweenthefloodingmechani sms. This slight increase in latency in PALER-II isdue in part to the random timing mechanism of PALER. InPALER-I, the nodes which do not choose the lowest randomdelay periods are the ones that suppress forwarding, whereasin PALER-II, the pruning selection is partially independent ofthedelayperiodchosen.

To examine how PALER performs in differing densities, another set of simulations was performed wit hdensityvaryingfrom 40 nodes per km<sup>2</sup>to 300 per km<sup>2</sup>. The results are shownin Figure 5, for loss environments of 10%. 35% and 50%. Theefficiency improvement of PALER-II is

### approximately

30%, which is consistent across all densities when a vera gedoverthedifferent loss environments. PALER-I increases in efficiencyimprovement as density increases. It results in a reduction oftransmissions of 23% for a density of 40 nodes per km<sup>2</sup>, butnearly 45% for densities of 300 nodes per  $\text{km}^2$ . In extremelydenseenvironments,PSFQactuallyperform edworseinalow-

lossenvironment, due to high contention during recover y.Asaresult,Figure5aisshownonadifferentscaletoacc ommodatetheplotofPSFO.TheseresultsshowthatPA LERscaleswell to extremely dense networks. The overhead of generatingneighborhood information in PALER-II can be seen in verydense networks. Since PALER-II requires nodes to broadcastat a certain probability for an initial period, collisions canresult in very dense networks. For densities nodesperkm<sup>2</sup>,PALERbelow 100 IIperformsmoreefficiently, butatdensities above 100n odesperkm<sup>2</sup>PALER-Iperformsmoreefficiently.This point of intersection corresponds to an average nodeconnectivity of approximately 20. In many cases, this may bean acceptable upper bound, but if an extremely dense networkisexpected, PALER-Iwillperformefficiently.

The greatest factor in the reduction of transmissions byPALER is the reduction of contention and collisions duringrecovery. The lazy recovery method allows nodes to carefullypace recovery operations and avoid redundant transmissions.Figure6showsthebreakdownofPALE Rtransmissionsbetween broadcasts, NACKs, and reply messages. The graphsfrom Figure 1 are included for comparison, it can be seen thatthe number of broadcasts is fairly similar, but the

### number

of NACKs and replies make up a significantly smaller per reentage of total transmissions in PALER.

### VI. CONCLUSIONS AND FUTURE WORK

The results from the previous section show that PALER isable to consistently improve energy efficiency above PSFQacrossabroadrangeofnetworksizesanddensitie s.Inaddition, latency proved to scale very well in PALER, par-ticularly in moderate loss

The environments. main reason fortheimprovedefficiencyinthenumberoftotalbroadc astisdue to the more relaxed nature of the recovery mechanism, which avoided much of the contention that affected PSFO.ThecomparisoninFigure6showsthatPALERp erformsfairlyconsistentlywithPSFOinthetotalnumbe rofforwardedbroadcasts.However.thepercentageofN ACKandreplymessages is greatly reduced in PALER. PALER implementedwithaneighborpruningmechanismsho wedanadditional

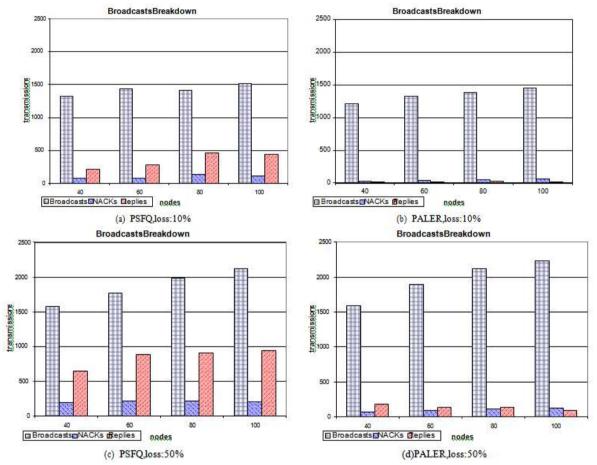


Fig. 6: Breakdown of the types of transmissions during a simulation of PALER. Simulation had a fixed field size withincreasing node count. PSFQ graphs as shown in Figure 1 are included for comparison (a)&(b) shows a comparison for alow-loss environment, 10%. (c)&(d) shows a comparison for a high-loss environment 50%.

PALER shows a much smallerpercentageofNACK and replymessages, relativetobroadcasts, incomparison to PSFQ.

improvement in energy efficiency in moderate densities. Thisincreaseinenergyefficiencycameatthecostofslig htlyhigherlatenciesthanthecountermethodimplemen tation.However,theaveragelatenciesofbothimplemen tationsscaledconsiderably better than PSFQ, and in many sensor networkenvironmentsthesavingsinpowerconsumpti onwilloutweighthesmallincreaseinlatency.

Weplantofurtherdevelopacomprehensivesensornetw orkreprogramming framework. Additional methods to

improveenergyefficiencywillbeexplored,suchasdyn amicallyadjust-ing radio transmission power. Neighbor locality informationmay be estimated from received signal strength. This infor-mation may be used to determine the minimum transmissionpower needed to reach all neighbors. The neighbor pruningmechanism leads nicely in this direction, since a list of neighborsrequiringthesegmentismaintained. If the estimate d

transmission power to reach a neighbor is added to each nodein the list, nodes may dynamically adjust their

transmissionpowertothemaximumofthesevaluesinth eremaininglistof neighbors. This has the potential to reduce the total powerexpendedfor distributing newcode,andreducecontentionand collisions since the transmission range will be limited tothedesiredcoveragearea.

Otherareasforfutureworkincludeaddingversionmeta datatofacilitateautomaticupdatesandmaintenance.Af ullyimplemented system should be capable of ensuring that allnodes maintain the most current version of code. This willrequire the ability to detect new or outdated versions andupdateanynecessarynodesinamannerthatallowst henetworktocontinuetoperformreliablyduringtheup date.Wewouldalsoliketosupportgroup-centricretasking.Sensornetworknodesmaybeorganizedintogr oupswhichperform

specialized functions. These group assignments may adjustdynamically to respond to the environment. We will exploremethodstoefficientlysupportsuchfunctionalit y.

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