A Reliable Multicast Framework for Light-weight Sessions and Application Level Framing

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ABSTRACT

This paper[1] describes SRM (Scalable Reliable Multicast), a reliable multicast framework for application level framing and light-weight sessions. The algorithms of this framework are efficient, robust, and scale well to both very large networks and very large sessions. The framework has been prototyped in wb, a distributed whiteboard application, and has been extensively tested on a global scale with sessions ranging from a few to more than 1000 participants. The paper describes the principles that have guided our design, including the IP multicast group delivery model, an end-to-end, receiver-based model of reliability, and the application level framing protocol model. As with unicast communications, the performance of a reliable multicast delivery algorithm depends on the underlying topology and operational environment. We investigate that dependence via analysis and simulation, and demonstrate an adaptive algorithm that uses the results of previous loss recovery events to adapt the control parameters used for future loss recovery. With the adaptive algorithm, our reliable multicast delivery algorithm provides good performance over a wide range of underlying topologies.

1 Introduction

Several researchers have proposed generic reliable multicast protocols, much as TCP is a generic transport protocol for reliable unicast transmission. In this paper we take a different view: unlike the unicast case where requirements for reliable, sequenced data delivery are fairly general, different multicast applications have widely different requirements for reliability. For example, some applications require that delivery obey a total ordering while many others do not. Some applications have many or all the members sending data while others have only one data source. Some applications have replicated data, for example in an n-redundant file store, so several members are capable of transmitting a data item while for others all data originates at a single source. These differences all affect the design of a reliable multicast protocol. Although one could design a protocol for the worst-case requirements, e.g., guarantee totally ordered delivery of replicated data from a large number of sources, such an approach results in substantial overhead for applications with more modest requirements. One cannot make a single reliable multicast delivery scheme that simultaneously meets the functionality, scalability and efficiency requirements of all applications.

The weakness of "one size fits all" protocols has long been recognized. In 1990 Clark and Tennenhouse proposed a new protocol model called Application Level Framing (ALF) which explicitly includes an application's semantics in the design of that application's protocol [CT90]. ALF was later elaborated with a light-weight rendezvous mechanism based on the IP multicast distribution model, and with a notion of receiver-based adaptation for unreliable, real-time applications such as audio and video conferencing. The result, known as Light-Weight Sessions (LWS), has been very successful in the design of wide-area, large-scale, conferencing applications. This paper further evolves the principles of ALF and LWS to add a framework for scalable reliable multicast (SRM).

ALF says that the best way to meet diverse application requirements is to leave as much functionality and flexibility

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as possible to the application. Therefore our algorithms are designed to meet only the minimal definition of reliable multicast, i.e., eventual delivery of all the data to all the group members, without enforcing any particular delivery order. We believe that if the need arises, machinery to enforce a particular delivery order can be easily added on top of this reliable delivery service.

The design is also heavily based on the group delivery model that is the centerpiece of the IP multicast protocol [D91]. In IP multicast, data sources simply send to the group's multicast address (a normal IP address chosen from a reserved range of addresses) without needing any advance knowledge of the group membership. To receive any data sent to the group, receivers simply announce that they are interested (via a "join" message broadcast on the local subnet) — no knowledge of the group membership or active senders is required. Each receiver joins and leaves the group individually, without affecting the data transmission to any other member. Our multicast delivery framework further enhances the multicast group concept by maximizing information and data sharing among all the members, and strengthens the individuality of membership by making each member responsible for its own correct reception of all the data.

Finally, our design attempts to follow the core design principles of TCP/IP. First, we require only the basic IP delivery model — best-effort with possible duplication and reordering of packets — and build the reliability on an end-to-end basis. No change or special support is required from the underlying IP network. Second, in a fashion similar to TCP adaptively setting timers or congestion control windows, our algorithms dynamically adjust their control parameters based on the observed performance within a session. This allows applications using this model to adapt to a wide range of group sizes, topologies and link bandwidths while maintaining robust and high performance.

The paper proceeds as follows: Section 2 discusses general issues for reliable multicast delivery. Section 3 describes in detail the reliable multicast algorithm embedded in the wb implementation. Section 4 discusses the performance of the algorithm in simple topologies such as chains, stars, and bounded-degree trees, and Section 5 presents simulation results from more complex topologies. Section 6 discusses extensions to the basic scheme embedded in wb, such as adaptive algorithms for adjusting the timer parameters and algorithms for local recovery. Section 7 discusses both the application-specific aspects of wb's reliable multicast algorithms as well as the aspects of the underlying approach that have general applicability. Section 8 discusses related work on reliable multicast. Section 9 discusses future work on the congestion control algorithms. Finally, Section 10 presents conclusions.

2 The design of reliable multicast

2.1 Reliable data delivery: adding the word "multicast"

The problem of reliable (unicast) data delivery is well understood and a variety of well-tested solutions are available. However, adding the word 'multicast' to the problem statement significantly changes the solution set. For example, in any reliable protocol some party must take responsibility for loss detection and recovery. Because of the "fate-sharing" implicit in unicast communication, i.e., the data transmission fails if either of the two ends fails, either the sender or receiver can take on this role. In TCP, the sender times transmissions and keeps retransmitting until an acknowledgment is received. NETBLT [CLZ87] uses the opposite model and makes the receiver responsible for all loss detection and recovery. Both approaches have been shown to work well for unicast.

However, if a TCP-style, sender-based approach is applied to multicast distribution, a number of problems occur. First, because data packets trigger acknowledgments (positive or negative) from all the receivers, the sender is subject to the well-known ACK implosion effect. Also, if the sender is responsible for reliable delivery, it must continuously track the changing set of active receivers and the reception state of each. Since the IP multicast model deliberately imposes a level of indirection between senders and receivers (i.e., data is sent to the multicast group, not to the set of receivers), the receiver set may be expensive or impossible to obtain. Finally, the algorithms that are used to adapt to changing network conditions tend to lose their meaning in the case of multicast. E.g., how should the round-trip time estimate for a retransmit timer be computed when there may be several orders of magnitude difference in propagation time to different receivers? What is a congestion window if the delay-bandwidth product to different receivers varies by orders of magnitude? What self-clocking information exists in the ACK stream(s) if some receivers share one bottleneck link and some another?

These problems illustrate that single-point, sender-based control does not adapt or scale well for multicast delivery. Since members of a multicast group have different communication paths and may come and go at any time, the "fate-shared" coupling of sender and receiver doesn't generalize to multicast. None of the problems described above exist with NETBLT-style, receiver-based reliability (e.g., since each receiver keeps its own reception state, the per-host state burden is constant, independent of group size, and the fact that group membership can't be known is irrelevant). Thus it is clear that receiver-based reliability is a far better building block for reliable multicast [PTK94].

Another unicast convention that migrates poorly to multicast has to do with the vocabulary used by the sender and receiver(s) to describe the progress of their communication. A receiver can request a retransmission either in application data units ("sector 5 of file sigcomm-slides.ps") or in terms

of the shared communication state ("sequence numbers 2560 to 3071 of this conversation"). Both models have been used successfully (e.g., NFS uses the former and TCP the latter) but, because the use of communication state for naming data allows the protocol to be entirely independent of any application's namespace, it is by far the most popular approach for unicast applications. However, since the multicast case tends to have much weaker and more diverse state synchronization, using that state to name data works badly. E.g., if a receiver joins a conversation late and receives sequence numbers 2560 to 3071, it has no idea of what's been missed (since the sender's starting number is arbitrary) and so can neither do anything useful with the data nor make an intelligent request for retransmission. If receivers hear from a sender again after a lengthy network partition, they have no way of knowing whether "2560" is a retransmission of data they received before the partition or is completely new (due to sequence number wrapping during the partition). Thus the "naming in application data units (ADUs)" model works far better for multicast. Use of this model also has two beneficial side effects. As [CT90] points out, a separate protocol namespace can impose delays and inefficiencies on an application, e.g., TCP will only deliver data in sequence even though a file transfer application might be perfectly happy to receive sectors in any order. The ADU model eliminates this delay and puts the application back in control. Also, since ADU names can be made independent of the sending host, it is possible to use the anonymity of IP multicast to exploit the redundancy of multiple receivers. E.g., if some receiver asks for a retransmit of "sigcomm-slides.ps sector 5", any member who has a copy of the data, not just the original sender, can carry out the retransmission.

2.2 Reliable multicast requirements

While the ALF model says that applications should be actively involved in their communications and that communication should be done in terms of ADUs rather than some generic protocol namespace, we do not claim that every application's protocol must be completely different from every other's or that there can be no shared design or code. A great deal of design commonality is imposed simply because different applications are attempting to solve the same problem: scalable, reliable, multipoint communication over the Internet. As Section 2.1 pointed out, just going from unicast to multicast greatly limits the viable protocol design choices. In addition, experience with the Internet has shown that successful protocols must accommodate many orders of magnitude variation in every possible dimension. While several algorithms meet the constraints of Section 2.1, very few of them continue to work if the delay, bandwidth and user population are all varied by factors of 1000 or more.

In the end we believe the ALF model results in a skeleton or template which is then fleshed out with application specific details. Portions of that skeleton are completely determined by network dynamics and scaling considerations and apply to any application. So, for example, the scalable request and repair algorithms described in Sections 3 through 6 are completely generic and apply to a wide variety of reliable multicast applications. Each different application supplies this reliability framework with a namespace to talk about what data has been sent and received; a policy and machinery to determine how bandwidth should be apportioned between a participant in the group, the group as a whole, and other users of the net; and a local send policy that a participant uses to arbitrate the different demands on its bandwidth (e.g., locally originated data, repair requests and responses, etc.). It is the intent of this paper to describe the skeleton common to scalable, reliable multicast applications. However, to make the ideas concrete, we first describe a complete, widely used application — wb, the LBL network whiteboard — that has been implemented according to this model. After mentioning some details of its operation that are direct implications of the design considerations in Section 2.1, we then factor out the wb specifics to expose the generic, scalable, reliable multicast skeleton underneath. The remaining sections of this paper are an exploration of that skeleton.

2.3 The wb framework

Wb is a network conferencing tool designed and implemented by McCanne and Jacobson [J92, J94a, M92] that provides a distributed whiteboard. The whiteboard separates the drawing into pages, where a new page can correspond to a new viewgraph in a talk or the clearing of the screen by a member of a meeting. Any member can create a page and any member can draw on any page.[2] Each member is identified by a globally unique identifier, the Source-ID, and each page is identified by the Source-ID of the initiator of the page and a page number locally unique to that initiator. Each member drawing on the whiteboard produces a stream of drawing operations that are timestamped and assigned sequence numbers, relative to the sender. Most drawing operations are idempotent and are rendered immediately upon receipt. Each member's graphics stream is independent from that of other sites.

The following assumptions are made in wb's reliable multicast design:

- All data has a unique name.

[2] There are floor control mechanisms, largely external to wb, that can be used if necessary to control who can create or draw on pages. These can be combined with normal Internet privacy mechanisms (e.g., symmetric-key encryption of all the wb data) to limit participation to a particular group and/or with normal authentication mechanisms (e.g., participants signing their drawing operations via public-key encryption of a cryptographic hash of the drawop). The privacy, authentication and control mechanisms are completely orthogonal to the reliability machinery that is the subject of this paper and will not be described here. For further details see [MJ95, J94].

This global name consists of the end host's Source-ID and a locally unique sequence number.

- The name always refers to the same data.

  It is impossible to achieve consistency among different receivers in the face of late arrivals and network partitions if, say, drawop "floyd:5" initially means a blue line and later turns into a red circle. This does not mean that the drawing can't change, only that drawops must effect the change. E.g., to change a blue line to a red circle, a "delete" drawop for "floyd:5" is sent, then a drawop for the circle is sent.

- Source-ID's are persistent.

  A user will often quit a session and later re-join, obtaining the session's history from the network. By ensuring that Source-ID's are persistent across invocations of the application, the user maintains ownership of any data created before quitting.

- IP multicast datagram delivery is available.

- All participants join the same multicast group; there is no distinction between senders and receivers.

Wb has no requirement for ordered delivery because most operations are idempotent. Operations that are not strictly idempotent, such as a "delete" that references an earlier drawop, can be patched after the fact, when the missing data arrives. A receiver uses the timestamps on the drawing operations to determine the rendering order. This coarse synchronization mechanism captures the temporal causality of drawing operations at a level appropriate for the application, without the added complexity and delay of protocols that provide guaranteed causal ordering.

3 Wb's instantiation of the reliable multicast algorithm

Whenever new data is generated by wb, it is multicast to the group. Each member of the group is individually responsible for detecting loss and requesting retransmission. Loss is normally detected by finding a gap in the sequence space. However, since it is possible that the last drawop of a set is dropped, each member sends low-rate, periodic, session messages that announce the highest sequence number received from every member that has written on the page currently being displayed. In addition to the reception state, the session messages contain timestamps that are used to estimate the distance (in time) from each member to every other (described in Section 3.1).

When receiver(s) detect missing data, they wait for a random time determined by their distance from the original source of the data, then send a repair request (the timer calculations are described in detail in Section 3.2). As with the original data, repair requests and retransmissions are always multicast to the whole group. Thus, although a number of hosts may all miss the same packet, a host close to the point of failure is likely to timeout first and multicast the request. Other hosts that are also missing the data hear that request and suppress their own request. (This prevents a request implosion.) Any host that has a copy of the requested data can answer a request. It will set a repair timer to a random value depending on its distance from the sender of the request message and multicast the repair when the timer goes off. Other hosts that had the data and scheduled repairs will cancel their repair timers when they hear the multicast from the first host. (This prevents a response implosion). In a topology with diverse transmission delays, a lost packet is likely to trigger only a single request from a host just downstream of the point of failure and a single repair from a host just upstream of the point of failure.

3.1 Session messages

As mentioned above, each member sends periodic session messages that report the sequence number state for active sources. Receivers use these session messages to determine the current participants of the session and to detect losses. The average bandwidth consumed by session messages is limited to a small fraction (e.g., 5%) of the session data bandwidth using the algorithm developed for vat and described in [SCFJ94].

In a large, long-lived session, the state would become unmanageable if each receiver had to report the sequence numbers of everyone who had ever written to the whiteboard. The "pages" mentioned above are used to partition the state and prevent this explosion. Each member only reports the state of the page it is currently viewing. If a receiver joins late, it may issue *page requests* to learn the existence of pages and the sequence number state in each page. We omit the details of the page state recovery protocol as it is almost identical to the repair request / response protocol for data.

In addition to state exchange, receivers use the session messages to estimate the one-way distance between nodes. All whiteboard packets, including session packets, include a Source-ID and a timestamp. The session packet timestamps are used to estimate the host-to-host distances needed by the repair algorithm.

The timestamps are used in a highly simplified version of the NTP time synchronization algorithm [M84]. Assume that host A sends a session packet P_1 at time t_1 and host B receives P_1 at time t_2. At some later time, t_3, host B generates a session packet P_2, marked with (t_1, Δ) where $\Delta = t_3 - t_2$ (time t_1 is included in P_2 to make the algorithm robust to lost session packets). Upon receiving P_2 at time t_4, host A can estimate the latency from host B to host A as $(t_4 - t_1 - \Delta)/2$. Note that while this estimate does assume that the paths are symmetric, it does not assume synchronized clocks.

3.2 Loss recovery

The loss recovery algorithm provides the foundation for reliable delivery. In this section we describe the loss recovery algorithm originally designed for wb; Section 6.1 describes a modified version of this algorithm with an adaptive adjustment of the timer parameters.

When host A detects a loss, it schedules a repair request for a random time in the future. The request timer is chosen from the uniform distribution on $[C_1 d_{S,A}, (C_1+C_2) d_{S,A}]$ seconds, where $d_{S,A}$ is host A's estimate of the one-way delay to the original source S of the missing data. When the request timer expires, host A sends a request for the missing data, and doubles the request timer to wait for the repair.

If host A receives a request for the missing data before its own request timer for that data expires, then host A does a (random) exponential backoff, and resets its request timer. That is, if the current timer had been chosen from the uniform distribution on

$$2^i[C_1 d_{S,A}, (C_1 + C_2) d_{S,A}],$$

then the backed-off timer is randomly chosen from the uniform distribution on

$$2^{i+1}[C_1 d_{S,A}, (C_1 + C_2) d_{S,A}].$$

When host B receives a request from A that host B is capable of answering, host B sets a repair timer to a value from the uniform distribution on

$$[D_1 d_{A,B}, (D_1 + D_2) d_{A,B}]$$

seconds, where $d_{A,B}$ is host B's estimate of the one-way delay to host A. If host B receives a repair for the missing data before its repair timer expires, then host B cancels its repair timer. If host B's repair timer expires before it receives a repair, then host B multicasts the repair. Because host B is not responsible for host A's reliable data reception, it does not verify whether host A actually receives the repair.

Due to the probabilistic nature of these algorithms, it is not unusual for a dropped packet to be followed by more than one request. Thus, a host could receive a duplicate request immediately after sending a repair, or immediately after receiving a repair in response to its own earlier request. In order to prevent duplicate requests from triggering a responding set of duplicate repairs, host B ignores requests for data D for $3\,d_{S,B}$ seconds after sending or receiving a repair for that data, where host S is either the original source of data D or the source of the first request.

Because data represents idempotent operations, loss recovery can proceed independently from the transmission of new data. Similarly, recovery for losses from two different sources can also proceed independently. Since transmission bandwidth is often limited, a single transmission rate is allocated to control the throughput across all these different modes of operation, while the application determines the order of packet transmission according to their relative importance. In wb, the highest priority packets are repairs for the current page, middle priority are new data, and lowest priority are repairs for previous pages.

3.3 Bandwidth limitations

The congestion control mechanism for whiteboard sessions is based on a (fixed, in current implementations) maximum bandwidth allocation for each session. Each wb session has a sender bandwidth limit advertised as part of the sd announcement. A typical value is 64 Kbps; in this case a wb session costs no more (and typically considerably less) than the accompanying audio session. Individual members use a token bucket rate limiter to enforce this peak rate on transmissions. This peak rate is mostly relevant when a source distributes a postscript file for a new page of the whiteboard, or when a late arrival requests the past history of the whiteboard session.

3.4 Recovery from partitioning

The whiteboard does not require special mechanisms for the detection or recovery from network partitioning. Because wb relies on the underlying concept of an IP multicast group, where members can arrive and depart independently, wb does not distinguish a partitioning from a normal departure of members from the wb session.

During a partition of a session, users can simply continue using the whiteboard in the connected components of the partitions. Because pages are identified by the Source-ID of the initiator of the page, along with the page number for that initiator, members can continue creating new pages during the partition (e.g., "Floyd:3" in one half of the partition, and "Zhang:5" in the other). After recovery each page will still have a unique page ID and the repair mechanism will distribute any new state throughout the entire group.

Almost all of the design described in this section is present in version 1.59 of wb; some omissions are pending implementation. These omissions include the measurements of one-way delays and the rate-limiting mechanisms.

4 Request/repair algorithms for simple topologies

Building on our initial design experiences in wb, we turn to a more general investigation of the request/repair algorithms. The algorithms described in the remainder of the paper have been implemented only within our simulation framework.

Given that multiple hosts may detect the same losses, and multiple hosts may attempt to handle the same repair request, the goal of the request/repair timer algorithms is to de-synchronize host actions to keep the number of duplicates low. Among hosts that have diverse delays to other hosts in the same group, this difference in delay can be used to differentiate hosts; for hosts that have similar delays to reach

others, we can only rely on randomization to de-synchronize their actions.

This section discusses a few simple, yet representative, topologies, namely chain, star, and tree topologies, to provide a foundation for understanding the request/repair algorithms in more complex environments. For a chain the essential feature of a request/repair algorithm is that the timer value be a function of distance. For a star topology the essential feature of the request/repair algorithm is the randomization used to reduce implosion. Request/repair algorithms in a tree combine both the randomization and the setting of the timer as a function of distance. This section shows that the performance of the request/repair algorithms depends on the underlying network topology.

4.1 Chains

Figure 1 shows a chain topology where all nodes in the chain are members of the multicast session. Each node in the underlying multicast tree has degree at most two. The chain is an extreme topology where a simple deterministic request/repair algorithm suffices; in this section we assume that $C_1, D_1 = 1$, and that $C_2, D_2 = 0$.

For the chain, as in most of the other scenarios in this paper, link distance and delay are both normalized. We assume that packets take one unit of time to travel each link, i.e. all links have distance of 1.

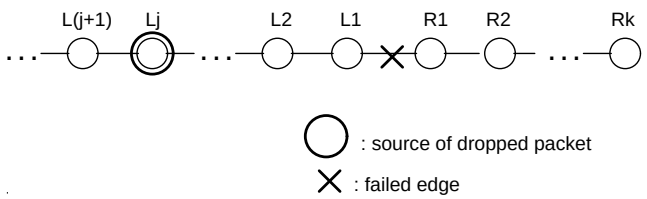

Figure 1: A chain topology.

In Figure 1 the nodes in the chain are labeled as either to the right or to the left of the congested link. Assume that source L_j multicasts a packet that is subsequently dropped on link (L_1, R_1), and that the second packet sent from source L_j is not dropped. We call the edge that dropped the packet, whether due to congestion or to other problems, the *congested* link. Assume that the right-hand nodes each detect the failure when they receive the second packet from L_j.

Assume that node R_1 first detects the loss at time t, and that each link has distance 1. Then node R_1 multicasts a request at time $t + j$. Node L_1 receives the request at time $t + j + 1$ and multicasts a repair at time $t + j + 2$. Node R_k receives the repair at time $t + k + j + 2$.

Note that all nodes to the right of node R_1 receive the request from R_1 before their own request timers expire. We call this *deterministic suppression*. We leave it as an exercise for the reader to verify that, due to deterministic suppression, there will be only one request and one repair.

Had the loss repair been done by unicast, i.e. node R_k sent a unicast request to the source L_j as soon as it detected the failure and L_j sent a unicast repair to R_k as soon as it received the request, node R_k would not receive the repair until time $t + 2j + 3k$. Thus, with a chain and with a simple deterministic request/repair algorithm, the furthest node receives the repair sooner than it would if it had to rely on its own unicast communication with the original source. While the original source and the intended recipient of the dropped packet could be arbitrarily far from the congested link, in the multicast repair algorithm both the request and the repair come from the node immediately adjacent to the congested link.

4.2 Stars

For the star topology in Figure 2 we assume that all links are identical and that the center node is not a member of the multicast group. For a star topology, setting the request timer as a function of the distance from the source is not an essential feature, as all nodes detect a loss at exactly the same time. Instead, the essential feature of the request/repair algorithm in a star is the randomization used to reduce implosion; we call this *probabilistic suppression*.

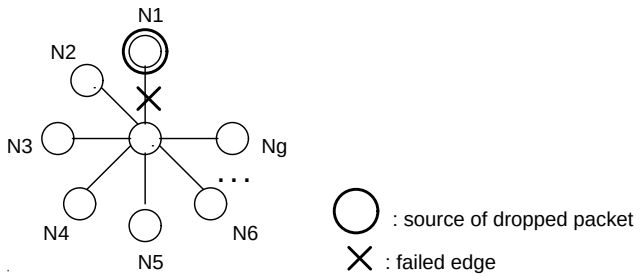

Figure 2: A star topology.

For the star topology in Figure 2 assume that the first packet from node N_1 is dropped on the adjacent link. There are G members of the multicast session, and the other members detect the loss at exactly the same time. For the discussion of this topology we assume that $C_1, D_1 = 0$; because all nodes detect losses and receive requests at the same time, C_1 and D_1 are not needed to amplify differences in delay. The only benefit in setting C_1 greater than 0 is to avoid unnecessary requests from out-of-order packets.

If C_2 is at most 1, then there will always be $G - 1$ requests. Increasing C_2 reduces the expected number of requests but increases the expected time until the first request is sent. For $C_2 > 1$, the expected number of requests is roughly $1 + (G - 2)/C_2$, and the expected delay until the first timer expires is $2C_2/G$ seconds (where one unit of time is one second).[3] For example, if C_2 is set to $\sqrt{G}$, then the expected

[3]The $G - 1$ nodes all detect the failure at the same time, and all set their timers to a uniform value in an interval of width $2C_2$. If the first timer

number of requests is roughly $\sqrt{G}$, and the expected delay until the first timer expires is $2/\sqrt{G}$ seconds. The same remarks apply to D_2 with respect to repairs.

4.3 Bounded-degree trees

The request/repair performance in a tree topology uses both the deterministic suppression described for chain topologies and the probabilistic suppression described for star topologies. Consider a network topology of a bounded-degree tree with N nodes where interior nodes have degree p. A tree topology combines aspects of both chains and stars. The timer value should be a function of distance, to enable requests and repairs to suppress request and repair timers at nodes further down in the tree. In addition, randomization is needed to reduce request/repair implosion from nodes that are at an equal distance from the source (of the dropped packet, or of the first request).

We assume that node S in the tree is the source of the dropped packet, and that link (B,A) drops a packet from source S. We call nodes on the source's side of the congested link (including node B) *good* nodes, and we call nodes on the other side of the congested link (including node A) *bad* nodes. Node A detects the dropped packet at time t, when it receives the next packet from node S. We designate node A as a *level-0* node, and we call a bad node a *level-i* node if it is at distance i from node A.

Assume that the source of the dropped packet is at distance j from node A. Node A's request timer expires at time

$$t + C_1 j + U_1[C_2]j,$$

where $U[C_2]$ denotes a uniform random variable between 0 and C_2. Assuming that node A's request is not suppressed, a level-i node receives node A's request at time

$$t + i + C_1 j + U_1[C_2]j.$$

Node B receives node A's repair request at time

$$t + 1 + C_1 j + U_1[C_2]j.$$

A bad level-i node detects the loss at time $t + i$, and such a node's request timer expires at some time

$$t + i + C_1(i + j) + U_2[C_2](i + j).$$

Note that regardless of the values of $U_1[C_2]$ and $U_2[C_2]$, a level-i node receives node A's request by time $t + i + C_1 j + C_2 j$, and a level-i node's request timer expires no sooner than $t + i + C_1(i + j)$. If

$$t + i + C_1 j + C_2 j \leq t + i + C_1(i + j),$$

that is, if

$$\frac{C_2}{C_1} j \leq i,$$

then the level-i node's request timer will always be suppressed by the request from the level-0 node. Thus, the smaller the ratio C_2/C_1, the fewer the number of levels that could be involved in duplicate requests. This relation also demonstrates why the number of duplicate requests or repairs is smaller when the source (of the dropped packet, or of the request) is close to the congested link.

Note that the parameter C_1 serves two different functions. A smaller value for C_1 gives a smaller delay for node B to receive the first request. At the same time, for nodes further away from the congested link, a larger value for C_1 contributes to suppressing additional levels of request timers. A similar tradeoff occurs with the parameter C_2. A smaller value for C_2 gives a smaller delay for node B to receive the first repair request. At the same time, as illustrated with star topologies, a larger value for C_2 helps to prevent duplicate requests from session members at the same distance from the congested link. Similar remarks apply to the functions of D_1 and D_2 in the repair timer algorithm.

5 Simulations of the request and repair algorithms

For a given underlying network, set of session members, session sources, and congested link, it should be feasible to analyze the behavior of the repair and request algorithms, given fixed timer parameters C_1, C_2, D_1, and D_2. However, we are interested in the repair and request algorithms across a wide range of topologies and scenarios. We use simulations to examine the performance of the request/repair algorithms for individual packet drops in random and bounded-degree trees. We do not claim to be presenting realistic topologies or typical patterns of packet loss.

The simulations in this section show that the request/repair algorithms with fixed timer parameters perform well in a random or bounded-degree tree when every node in the underlying tree is a member of the multicast session. The request/repair algorithms perform somewhat less well for a sparse session, where the session size is small relative to the size of the underlying network. This motivates the development on the adaptive request/repair algorithm in Section 6.1, where the timer parameters C_1, C_2, D_1, and D_2 are adjusted in response to past performance.

In these simulations the fixed timer parameters are set as follows: $C_1, C_2 = 2$, and $D_1, D_2 = \log_{10} G$, where G is the number of members in the same multicast session. The choice of $\log_{10} G$ for D_1 and D_2 is not critical, but gives slightly better performance than $D_1, D_2 = 1$ for large G.

Each simulation constructs either a random tree or a bounded degree tree with N nodes as the network topology. Next, G of the N nodes are randomly chosen to be session

expires at time t, then the other $G - 2$ receivers receive that first request at time $t + 2$. So the expected number of duplicate requests is equal to the expected number of timers that expire in the interval $[t, t + 2]$.

members, and a source S is randomly chosen from the G session members.

We assume that messages are multicast to members of the multicast group along a shortest-path tree from the source of the message. In each simulation we randomly choose a link L on the shortest-path tree from source S to the G members of the multicast group. We assume that the first packet from source S is dropped by link L, and that receivers detect this loss when they receive the subsequent packet from source S.

5.1 Simulations on random trees

We first consider simulations on random labeled trees of N nodes, constructed according to the labeling algorithm in [Pa85, p.99]. These trees have unbounded degree, but for large N, the probability that a particular vertex in a random labeled tree has degree at most four approaches 0.98 [Pa85, p.114]. Figure 3 shows simulations of the request/repair algorithm for this case, where all N nodes in the tree are members of the multicast session (that is, $G = N$). For each graph the x-axis shows the session size G; twenty simulations were run for each value of G. Each simulation is represented by an jittered dot, and the median from the twenty simulations is shown by a solid line. The two dotted lines mark the upper and lower quartiles; thus, the results from half of the simulations lie between the two dotted lines.

The top two graphs in Figure 3 show the number of requests and repairs to recover from a single loss. The bottom graph shows the delay of the last node in the multicast session to receive the repair. For each member affected by the loss, we define the delay as the time from when the member first detected the loss until the member first received a repair. The graph shows this delay as a multiple of RTT, the roundtrip time from the receiver to the original source of the dropped packet.

Note that with unicast communications the ratio of delay to RTT is at least one. For a unicast receiver that detects a packet loss by waiting for a retransmit timer to time out, the typical ratio of delay to RTT is closer to 2. As the earlier discussion of chain topologies shows, with multicast request/repair algorithms the ratio of delay to RTT can sometimes be less than one, because the request and/or repair could each come from a node close to the point of failure.

Figure 3 shows that the repair/request algorithm works well for a tree topology where all nodes of the tree are members of the multicast session. There is usually only one request and one repair. (Some lack of symmetry results from the fact that the original source of the dropped packet might be far from the point of failure, while the first request comes from a node close to the point of failure.) The average recovery delay for the farthest node is roughly 2 RTT, competitive with the average delay available from a unicast algorithm such as TCP. The results are similar in simulations where the congested link is chosen adjacent to the source of the dropped packet, and for simulations on a bounded-degree tree of size $N = G$ where interior nodes have degree 4.

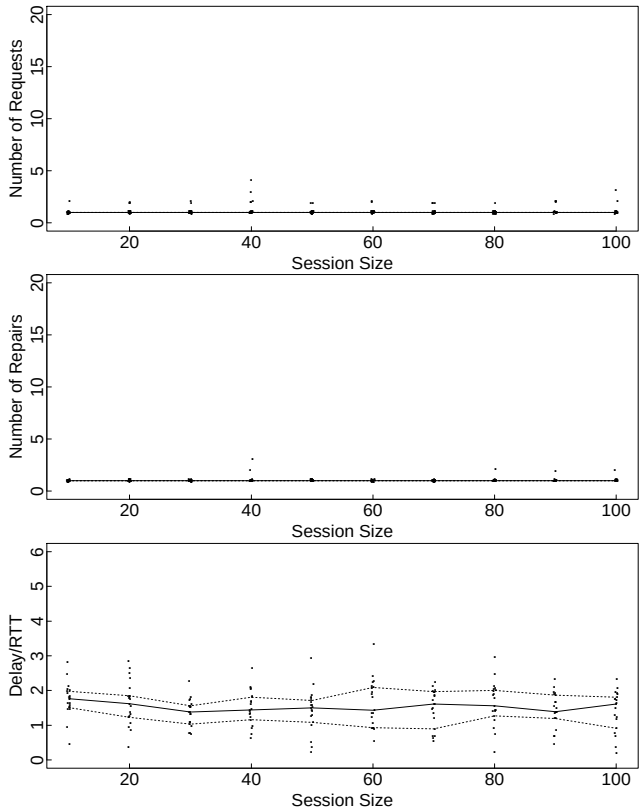

Figure 3: Random trees with a random congested link, where all nodes are members of the multicast session.

5.2 Simulations on large bounded-degree trees

The performance of the request/repair algorithms with fixed timer parameters is less optimal when the underlying network is a large bounded-degree tree. The underlying topology for the simulations in this section is a balanced bounded-degree tree of $N = 1000$ nodes, with interior nodes of degree four. In these simulations the session size G is significantly less than N. For a session that is sparse relative to the underlying network, the nodes close to the congested link might not be members of the session.

As Figure 4 shows, the average number of repairs for each loss is somewhat high. In simulations where the congested link is always adjacent to the source, the number of repairs is low but the average number of requests is high.

[FJLMZ95] shows the performance of the request/repair algorithm on a range of topologies. These include topologies where each of the N nodes in the underlying network is a router with an adjacent ethernet with 5 workstations, point-to-point topologies where the edges have a range of propagation delays, and topologies where the underlying network is more dense that a tree. None of these variations that we

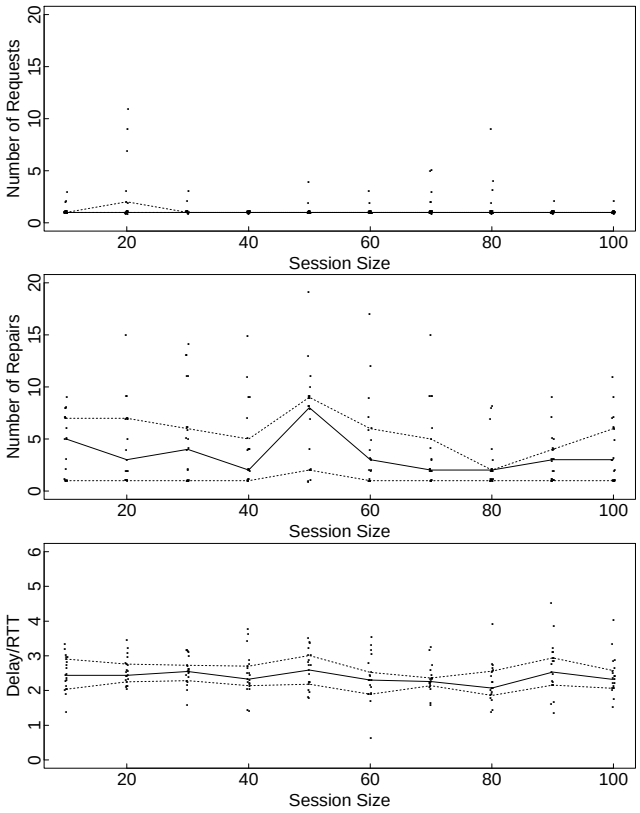

Figure 4: Bounded-degree tree, degree 4, 1000 nodes, with a random congested link.

have explored have significantly affected the performance of the request/repair algorithms with fixed timer parameters.

6 Extending the basic approach

6.1 Adaptive adjustment of random timer algorithms

The close connection of the loss recovery performance with the underlying topology of the network suggests that the timer parameters C_1, C_2, D_1, and D_2 be adjusted in response to the past behavior of the request/repair algorithms. In this section we describe an adaptive algorithm that adjusts the timer parameters as a function of both the delay and of the number of duplicate requests and repairs in recent loss recovery exchanges.

Figure 5 gives the outline of the dynamic adjustment algorithm for adjusting the timer parameter C_2, which controls the width of the request timer interval. If the average number of duplicate requests is too high, then the adaptive algorithm increases the request timer interval. Alternately, if the average number of duplicates is okay but the average delay in sending a request is too high, then the adaptive algorithm decreases the request timer interval. In this fashion the adaptive algorithm can adapt the timer parameters not only to fit the fixed underlying topology, but also to fit a changing session

```
Before each new request timer is set:
   if ave. dup.  requests high
      increase request timer interval
   else if ave. dup. requests low
   and ave. req. delay too high
      decrease request timer interval
```

Figure 5: Dynamic adjustment algorithm for request timer interval.

membership and pattern of congestion.

First we describe how a session member measures the average delay and number of duplicate requests in previous loss recovery rounds in which that member has been a participant. A *request period* begins when a member first detects a loss and sets a request timer, and ends only when that member begins a new request period. The variable *dup_req* keeps count of the number of duplicate requests received during one request period; these could be duplicates of the most recent request or of some previous request, but do not include requests for data for which that member never set a request timer. At the end of each request period, the member updates *ave_dup_req*, the average number of duplicate requests per request period, before resetting *dup_req* to zero. The average is computed as an exponential-weighted moving average,

$$ave_dup_req \leftarrow (1-\alpha)\, ave_dup_req + \alpha\, dup_req,$$

with $\alpha = 1/4$ in our simulations. Thus, *ave_dup_req* gives the average number of duplicate requests for those request events for which that member has actually set a request timer.

When a request timer either expires or is reset for the first time, indicating that either this member or some other member has sent a request for that data, the member computes req_delay, the delay from the time the request timer was first set (following the detection of a loss) until a request was sent (as indicated by the time that the request timer either expired or was reset). The variable req_delay expresses this delay as a multiple of the roundtrip time to the source of the missing data. The member computes the average request delay, ave_req_delay.

In a similar fashion, a *repair period* begins when a member receives a request and sets a repair timer, and ends when a new repair period begins. In computing *dup_rep*, the number of duplicate repairs, the member considers only those repairs for which that member at some point set a repair timer. At the end of a repair period the member updates *ave_dup_rep*, the average number of duplicate repairs.

When a repair timer either expires or is cleared, indicating that this member or some other member sent a repair for that data, the member computes rep_delay, the delay from the time the repair timer was set (following the receipt of a request) until a repair was sent (as indicated by the time that the repair timer either expired or was cleared). As above, the

variable rep_delay expresses this delay as a multiple of the roundtrip time to the source of the missing data. The member computes the average repair delay, ave_rep_delay.

```
After a request timer expires or is first
          reset:
   update ave_req_delay
Before each new request timer is set:
   update ave_dup_req
   if (ave_dup_req ≥ AveDups))
      C1+ = 0.1
      C2+ = 0.5
   else if (ave_dup_req < AveDups−ε)
      if (ave_req_delay > AveDelay)
         C2− = 0.1
      if (ave_dup_req < 1/4)
         C1− = 0.05
   else C1+ = 0.05
```

Figure 6: Dynamic adjustment algorithm for request timer parameters. In our simulations $\epsilon = 0.1$

Figure 6 gives the adaptive adjustment algorithm used in our simulator to adjust the request timer parameters C_1 and C_2. The adaptive algorithm is based on comparing the measurements *ave_dup_req* and *ave_req_delay* with AveDups and AveDelay, the target bounds for the average number of duplicates and the average delay. An identical adjustment algorithm is used to adapt the repair timer parameters D_1 and D_2, based on the measurements *ave_dup_rep* and *ave_rep_delay*. Figure 7 gives the initial values used in our simulations for the timer parameters. All four timer parameters are constrained to stay within the minimum and maximum values in Figure 7.

```
Initial values:
   C1 = 2
   D1 = log10G
   C2 = 2
   D2 = log10G
Fixed parameters:
   MinC1 = 0.5;  MaxC1 = 2
   MinC2 = 1;  MaxC2 = G
   MinD1 = 0.5;  MaxD1 = log10G
   MinD2 = 1;  MaxD2 = G
   AveDups = 1
   AveDelay = 1
```

Figure 7: Parameters for adaptive algorithms

We are not trying to devise an optimal adaptive algorithm for reducing some function of both delay and of the number of duplicates; such an optimal algorithm could involve rather complex decisions about whether to adjust mainly C_1 or C_2, possibly depending on such factors as that member's relative distance to the source of the lost packet. Recall that increasing C_2 is guaranteed to reduce the number of duplicate requests; in contrast, increasing C_1 reduces the number of duplicate requests only when the members of the multicast group have diverse delays to reach each other. Our adaptive algorithm relies largely on adjustments of C_2 to reduce duplicates. Our adaptive algorithm only decreases C_1 when the average number of duplicates is already quite small (e.g., in scenarios where there are only one or two nodes capable of sending a request).

Because of the probabilistic nature of the repair and request algorithms, the behavior might vary over a fairly wide range even with a fixed set of timer parameters. Thus, the adaptive algorithm does not assume that the average number of duplicates is controlled until *ave_dup_req* is less than AveDups$-\epsilon$.

The numerical parameters in Figure 6 of 0.05, 0.1, and 0.5 were chosen somewhat arbitrarily. The adjustments of ± 0.05 and $+0.1$ for C_1 are intended to be small, as the adjustment of C_1 is not the primary mechanism for controlling the number of duplicates. The adjustments of -0.1 and $+0.5$ for C_2 are intended to be sufficiently small to minimize oscillations in the setting of the timer parameters. Sample trajectories of the request/repair algorithms confirm that the variations from the random component of the timer algorithms dominate the behavior of the algorithms, minimizing the effect of oscillations.

In our simulations we use a multiplicative factor of 3 rather than 2 for the request timer backoff described in Section 3.2. With a multiplicative factor of 2, and with an adaptive algorithm with small minimum values for C_1 and C_2, a single node that experiences a packet loss could have its request timer expire before receiving the repair packet, resulting in an unnecessary duplicate request.

Figures 8 and 9 show simulations comparing adaptive and non-adaptive algorithms. The simulation set in Figure 8 uses fixed values for the timer parameters, and the one in Figure 9 uses the adaptive algorithm. From the simulation set in Figure 4, we chose a network topology, session membership, and drop scenario that resulted in a large number of duplicate requests with the non-adaptive algorithm. The network topology is a bounded-degree tree of 1000 nodes with degree 4 for interior nodes, and the multicast session consists of 50 members.

Each of the two figures shows ten runs of the simulation, with 100 loss recovery rounds in each run. For each round of a simulation, the same topology and loss scenario is used, but a new seed is used for the pseudo-random number generator to control the timer choices for the requests and repairs. In each round a packet from the source is dropped on the congested link, a second packet from the source is not dropped, and the request/repair algorithms are run until all members

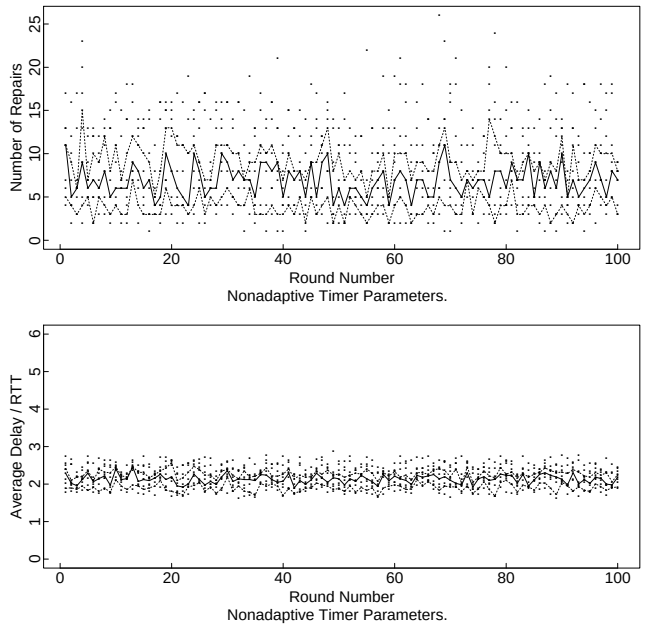

Figure 8: The non-adaptive algorithm.

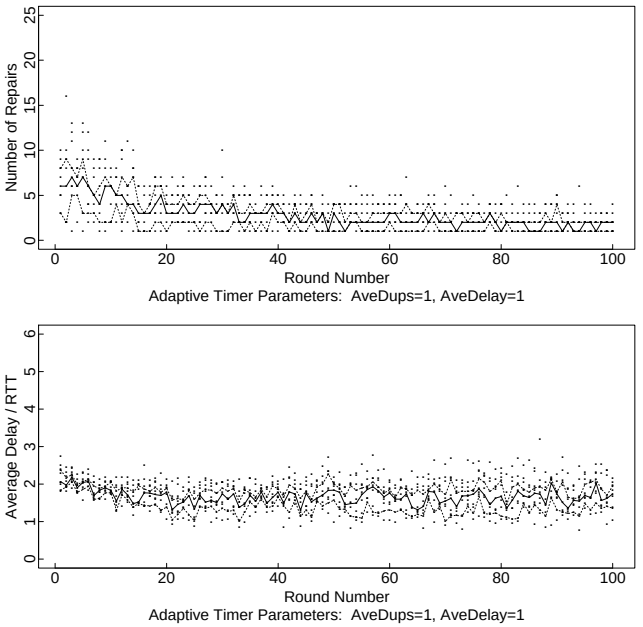

Figure 9: The adaptive algorithm.

have received the dropped packet. Each round of each simulation is marked with a dot, and a solid line shows the median from the ten simulations. The dotted lines show the upper and lower quartiles.

For the simulations in Figure 8 with fixed timer parameters, one round differs from another only in that each round uses a different set of random numbers for choosing the timers.

For the simulations with the adaptive algorithm in Figure 9, after each round of the simulation each session member uses the adaptive algorithms to adjust the timer parameters, based on the results from previous rounds. Figure 9 shows that for this scenario, the adaptive algorithms quickly reduce the average number of repairs with little penalty in additional delay. The average delay is roughly the same for the adaptive and the non-adaptive algorithms, but with the adaptive algorithm the delay has a somewhat higher variance.

Figure 10 shows the results of the adaptive algorithm on the same set of scenarios as that in Figure 4. For each scenario (i.e., network topology, session membership, source member, and congested link) in Figure 10, the adaptive algorithm is run repeatedly for 40 loss recovery rounds, and Figure 10 shows the results from the 40th loss recovery round. Comparing Figures 4 and 10 shows that the adaptive algorithm is effective in controlling the number of duplicates over a range of scenarios.

Simulations in [FJLMZ95] show that the adaptive algorithm works well in a wide range of conditions. These include scenarios where only one session member experiences the packet loss; where the congested link is chosen adjacent to the source of the packet to be dropped; and for a range of underlying topologies, including 5000-node trees, trees with

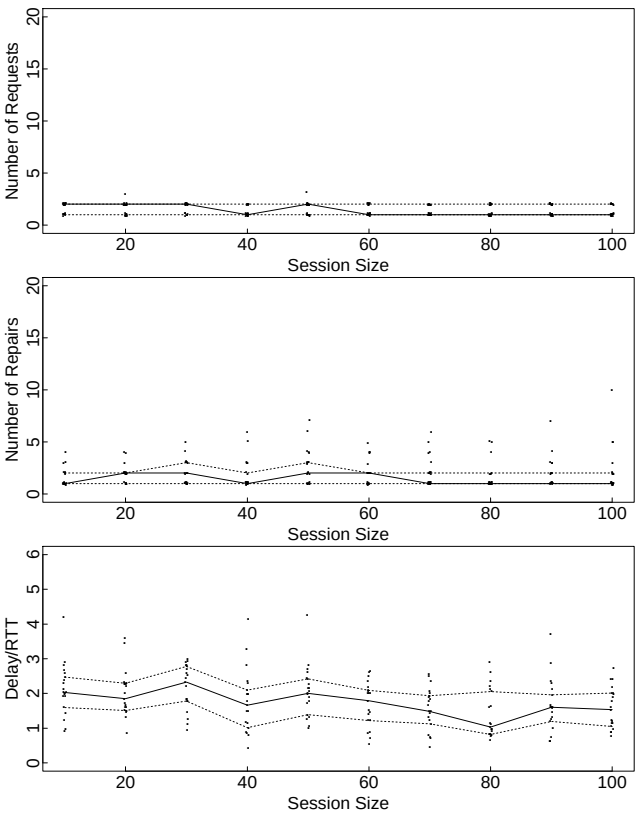

Figure 10: Adaptive algorithm on round 40, for a bounded-degree tree of 1000 nodes with degree 4 and a randomly picked congested link.

interior nodes of degree 10; and connected graphs that are

more dense that trees, with 1000 nodes and 1500 edges.

In actual multicast sessions, successive packet losses are not necessarily from the same source or on the same network link. Simulations in [FJLMZ95] show that in this case, the adaptive timer algorithms tune themselves to give good average performance for the range of packet drops encountered. [FJLMZ95] explores the benefits of adding additional conditions to the adaptive algorithm to monitor the worst-case as well as the average delay and number of duplicates. Simulations in [FJLMZ95] show that, by choosing different values for AveDelay and AveDups, tradeoffs can be made between the relative importance of low delay and a low number of duplicates.

In the simulations in this section, none of the requests or repairs are themselves dropped. In more realistic scenarios where not only data messages but requests and repairs can be dropped at congested links as well, members have to rely on retransmit timer algorithms to retransmit requests and repairs as needed. Obviously, this will increase not only the delay, but also the number of duplicate requests and repairs in different parts of the network.

6.2 Local recovery

With the request/repair algorithm described above, even if a packet is dropped on an edge to a single member, both the request and the repair are multicast to the entire group. In cases where the neighborhood affected by the loss is small, the bandwidth costs of the request/repair algorithm can be reduced if requests and repairs are multicast with limited scope. This use of limited scope can be implemented by setting an appropriate "hop count" in the *time-to-live* (TTL) field of the IP header.

Local recovery requires that the member sending the request have some information about the neighborhood of members sharing the same losses. However, end nodes should not know about network topology. We define a *loss neighborhood* is a set of members who are all experiencing the same set of losses. End nodes can learn about "loss neighborhoods" from information in session messages, without learning about the network topology. For each member, we call a loss a *local loss* if the number of members experiencing the loss is much smaller than the total number of members in the session. To help identify loss neighborhoods, session messages could report the names of the last few local losses. In addition, session messages could report the fraction of received repairs that are *redundant*, that is, those repairs received for known data, for which that member never set a request timer.

Assume for the moment that after a number of local losses with a stable loss neighborhood a member M can use session messages to estimate the size of the local neighborhood, that is, the minimum TTL h_1 needed to reach all members sharing the same losses. Further assume that from previous loss recoveries M can estimate h_2, the minimum TTL needed to reach some member not in the loss neighborhood. To use local recovery for the next request, M sends the request with TTL $h_3 = \text{Max}(h_1, h_2)$. If this loss follows the same history as the previous local losses, then h_3 is sufficient to suppress requests from other members in the loss neighborhood and to reach some member capable of answering the request. A member receiving a request from M that was sent with TTL h_3 answers with a repair of TTL $h_3 + h(M)$, where $h(M)$ is the number of hops to reach M. For a network with symmetric paths and thresholds, this repair reaches M with a remaining TTL of h_3, and therefore reaches all members reached by the original request.

Scenarios that could particularly benefit from local recovery include sessions with persistent losses to a small neighborhood of members, and isolated late arrivals to a multicast session asking for back history. [FJLMZ95] explores local recovery for a range of environments, including environments like the current Mbone where regions of the network are separated from each other by paths with high thresholds. We are also investigating the use of separate multicast groups for local recovery.

7 Application-specific and general aspects of reliable multicast

Section 2 discussed some of the underlying assumptions in the design of reliable multicast for wb. In this section we explore some of the ways that the reliable multicast framework described in this paper could be used and modified to meet the needs of other applications for reliable multicast.

A fundamental concept in our reliable multicast algorithm is a *multicast group*, i.e. a set of hosts that (1) can be reached by a group address without being identified individually first, and (2) share the same application data and thus can help each other with loss recovery. This group concept is also appropriate for applications such as routing protocol updates and DNS updates, as well as for the group distribution of stock quotes, Usenet news, or WWW-based mass media.

Let's take the Border Gateway Protocol (BGP) as an example. The Internet is viewed as a set of arbitrarily connected autonomous systems (AS) that are connected through border gateways that speak BGP to exchange routing information. One AS may have multiple BGP speakers, and all BGP speakers representing the same AS must give a consistent image of the AS to the outside, i.e. they must maintain consistent routing information. In the current implementation, this consistency is achieved by each BGP router opening a TCP connection to each other BGP router to deliver routing updates reliably. There are several problems with this approach. First, achieving multicast delivery by multiple one-to-one connections bears a high cost. Second, for an AS with N BGP routers, one has to manually configure the $(N-1)$ TCP connections for each of the N routers, and repeat again

whenever a change occurs. Both of these problems could be solved by applying our reliable multicast algorithm, perhaps with some minor adjustments to the data persistence model.

Our reliable multicast framework could easily be adapted for the distribution of such delay-insensitive material as Usenet news. Different applications have different tradeoffs between minimizing delay and minimizing the number of duplicate requests or repairs. For an interactive application such as wb, close attention must be paid to minimizing delay. For reliably distributing Usenet news, on the other hand, minimizing bandwidth would be more important than minimizing delay. Again some minor tuning to our request and repair timer algorithms may make our work readily applicable to the news distribution.

As a third example, we could consider applying the basic approach in this work to data caching and replication for Web pages. Like wb, all objects in the Web have a globally unique identifier. With HTTP, all requests for a specific object are handled by the original source, even though in many cases, especially for "hot" objects, a copy may be found within the neighborhood of a requester. When distributed Web caches are implemented, our reliable multicast framework could be used to reliably distribute updates to the caches. In addition, when a user makes a request to a remote object, the request could be multicast to the cache group. By using our timer algorithms, the cache closest to the requester would be most likely to send a reply.

We believe that the approach to reliable multicast described in this paper could be useful to a wide range of applications based on multicast groups. Even for applications that may require partial or total data ordering, the reliable multicast framework described in this paper could be used to reliably deliver the data to all group members, and a partial or total ordering protocol could be built on top that is specifically tailored to the ordering needs of that application.

8 Related work on reliable multicast

The literature is rich with architectures for reliable multicast. Due to space limitations, we will not describe the details of each solution. Instead, we focus on the different goals and definitions of reliability in the various architectures, and the implications of these differences for the scalability, robustness, handling of dynamic group membership, and overhead of the algorithms.

The Chang and Maxemchuk protocol [CM84] is one of the pioneer works in reliable multicast protocols. It is basically a centralized scheme that provides totally ordered delivery of data to all group members. All the members are ordered in a logical ring, with one of them being the master, called the token site. The token site is moved around the ring after each data transmission. Sources multicast new data to the group, and the token site is responsible for acknowledging (by multicast) the new data with a timestamp, as well as retransmitting (through unicast) all missing packets upon requests from individual receivers. The order of data reception at all the sites is determined by the timestamp in the ACK. Each ACK also serves to pass the token to the next member in the ring. By shifting the token site among all the members, with a requirement that a site can become the token site only if it has received all the acknowledged data, it is assured that after shifting the token site through all the N members in the group, everyone will have received all the data that is at least N smaller than the current timestamp value.

Because the token site is responsible for all the acknowledgments and retransmissions, it becomes the bottleneck point whenever losses occur. The scheme also requires reformation of the ring whenever a membership change occurs. Therefore it does not scale well with the size of the group.

RMP (Reliable Multicast Protocol) [WKM95] is an enhanced implementation of the Chang and Maxemchuk algorithm with added QoS parameters in each data transfer and better handling of membership changes.

The reliable multicast protocol for ordered delivery described in [KTHB89] is similar to, but simpler than, the Chang and Maxemchuk protocol. Basically, all data is first unicast to a master site, called a sequencer, which then multicasts the data to the group. Therefore the sequencer provides a global ordering of all the data in time; it is also responsible for retransmitting, by unicast, all the missing data upon requests. The sequencer site does not move unless it fails, in which case a new sequencer is elected. To avoid keeping all the data forever, the sequencer keeps track of the receiving state of all the members to determine the highest sequence number that has been correctly received by all the members.

MTP (Multicasting Transport Protocol) [AFM92] is again a centralized scheme for totally ordered multicast delivery. A master site is responsible for granting membership and tokens for data transmission; each host must obtain a token from the master first before multicasting data to the group, thus the total order of data packets is maintained. A window size defines the number of packets that can be multicast into the group in a single heartbeat and a retention size defines the period (in heartbeats) to maintain all client data for retransmission. NACKs are unicast to the data source which then multicasts the retransmission to whole group.

Compared to the above cited works, the Trans and Total protocols described in [MMA90] are the closest in spirit to our work. These protocols assume that all the members in a multicast group are attached to one broadcast LAN. Each host keeps an acknowledgment list which contains identifiers of both positive and negative ACKs. Whenever a host sends a data packet, it attaches its acknowledgment list to the packet, as a way to synchronize the state with all other members in the group. Because the single LAN limits data transmissions from all hosts to one packet at a time, partial and total ordering of data delivery can be readily derived from data and acknowledgment sequences.

Perhaps the most well-known work on reliable multicast is the ISIS distributed programming system developed at Cornell University [BSS91]. It provides causal ordering and, if desired, total ordering of messages *on top of* a reliable multicast delivery protocol. Therefore the ISIS work is to some extent orthogonal to the work described in this paper, and further confirms our notion that a partial or total ordering, when desired, can always be added on top of a reliable multicast delivery system. The reliable multicast delivery in existing ISIS implementations is achieved by multiple unicast connections using a windowed acknowledgment protocol similar to TCP [B93]. A new implementation has been announced recently that can optionally run on top of IP multicast.

9 Future work on congestion control

SRM assumes that the multicast session has a maximum bandwidth allocation for the session. We are continuing research on a number of congestion control issues related to this bandwidth allocation.

Given this bandwidth allocation, in an application tuned to the worst-case receiver members could give priority to the transmission of repairs, refraining from sending new data in the absence of available bandwidth. In an application like wb not tuned to the worst-case receiver, the application gives the transmission of new data priority over the repairs for previous pages. In such a reliable multicast session limited by a fixed or adaptive target bandwidth, a session member that is falling behind could either wait for the congestion to clear or unsubscribe from the multicast session.

The congestion control mechanisms required from an application using reliable multicast depend in part on the resource management services available from the network. For realtime traffic (i.e., traffic such as audio and video that is constrained by a fixed or adaptive playback time), some researchers have proposed that the network provide realtime services with an explicit reservation setup protocol, admission control procedures, and appropriate scheduling algorithms, to provide for guaranteed and predictive service [BCS94]. If members of a reliable multicast application were to take advantage of such services, and make reservations for a fixed target bandwidth, then each member simply requires a procedure for determining whether the session is over or under its bandwidth allocation.

On the other hand, if the application uses an adaptive rather than a fixed target bandwidth, adapting the target bandwidth for the session in response to congestion in the network, then the additional question remains of how this adaptive target bandwidth would be determined. One possibility that requires additional research would be to use multiple multicast groups, with a low-bandwidth multicast group targeted to the needs of the worst-case receivers, and limited to low-bandwidth data and repairs for the current page.

10 Conclusions and future work

This paper described in detail SRM, a scalable reliable multicast algorithm that was first developed to support wb. We have discussed the basic design principles as well as extensions of the basic algorithm that make it more robust for a wide range of network topologies.

Many applications need or desire support for reliable multicast. Experience with the wb design shows, however, that individual applications may have widely different requirements of multicast reliability. Instead of designing a generic reliable multicast protocol to meet the most stringent requirements, this work has resulted in a simple, robust, and scalable reliable multicast algorithm that meets a minimal reliability definition of delivering all data to all group members, leaving more advanced functionalities, whenever needed, to be handled by individual applications.

The work described in this paper is based on the fundamental principles of application level framing (ALF), multicast grouping, and the adaptivity and robustness in the TCP/IP architecture design. Although the work started with the goal of supporting wb, the end results should be generally applicable to a wide variety of other applications.

11 Acknowledgments

This work benefited from discussions with Dave Clark and with the End-to-End Task Force about general issues of sender-based vs. receiver-based protocols. We would also like to thank Peter Danzig for discussions about reliable multicasting and web-caching.

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A Details of simulator

In this section we show one of the tools that we use to verify that our simulator is implementing the request/repair algorithms correctly. This figure also serves as a concrete illustration of the request/repair algorithms in operation.

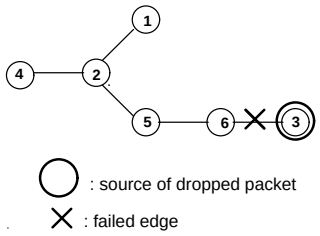

Figure 11: A simulation network for the figure above.

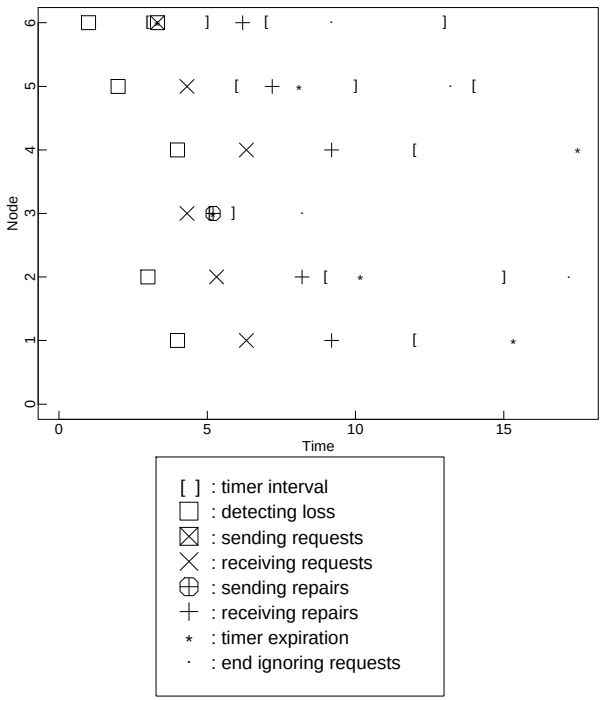

Figure 12: A request/repair exchange from a single dropped packet.

Figure 12 shows a single request/repair exchange for the network in Figure 11. This is one of a series of automated tests that we run after each change we make to our simulator.

The underlying network shown in Figure 11 consists of a randomly-created tree of six nodes. A packet takes one unit of time to traverse each link.

In Figure 12 the x-axis shows time and the y-axis shows a row for each session member, indicating when timers are set and repair and request packets sent and received by that member. This simulation uses the fixed timer parameters $C_1, C_2 = 2$ and $D_1, D_2 = 1$. The simulator's summary statistics correctly report that there was one request and one repair, and that the delay/RTT for the last node to receive the repair is 0.65. This is for node 1, which detects the loss at time 4, receives the repair at time 9.2, and has a RTT of 8 to the source of the dropped packet.

Figure 12 can be read in two ways to verify the correctness of the algorithms implemented in the simulator. First, a single row shows the history of a single member. We leave the verification of each row as an exercise for the reader. Second, for each multicast request or repair, the figure shows when that message was received by each of the other nodes.

In our simulator, we don't backoff the request timer for every duplicate request that is received. For example, if a member receives several duplicate requests immediately after receiving the initial request, then that member only backs off its request timer once, not several times. After the initial timer backoff, we only backoff the timer again if a request is received close to the time when the timer is set to expire. More precisely, when we backoff the request timer, then we set an *ignore-backoff* variable to a time halfway between the current time and the expiration time, and we ignore additional duplicate requests until *ignore-backoff* time.

B Simulations with fixed timer parameters

In this section we show additional simulations with fixed request/repair timer parameters (in addition to those in Section 5). These simulations show the performance of the non-adaptive algorithm for a wide range of topologies, including networks with a range of link distances, networks with local ethernets, large networks of high degree, and networks that are more dense than a simple tree. None of these variations on the underlying network topology significantly change the performance of the non-adaptive algorithm. That is, for the simulations with random congested links, the number of duplicate repairs can be somewhat high.

We also show a simulation where the congested link is chosen to be adjacent to the source of the dropped packet. In this case, the number of duplicate repairs is low, but the number of duplicate requests is high instead.

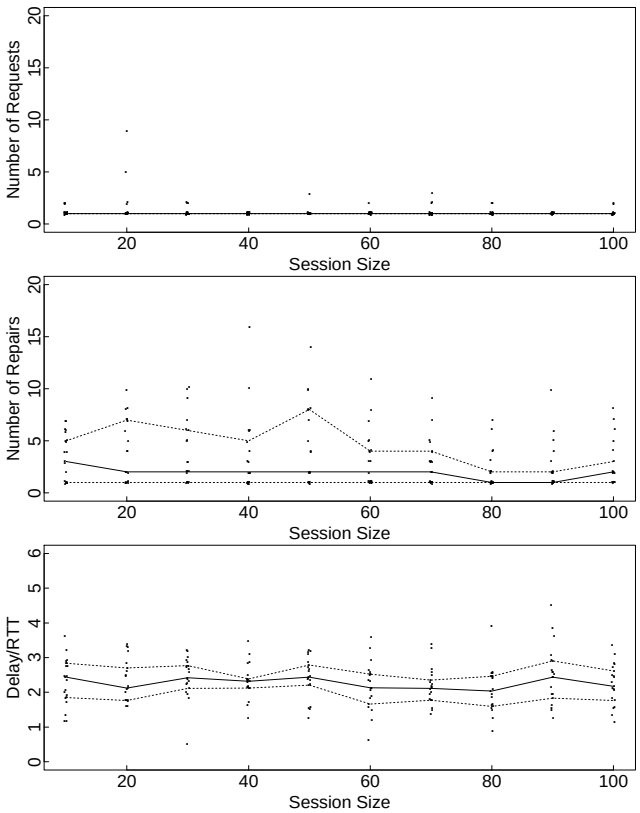

Figure 13: Non-adaptive algorithm for a bounded-degree tree, degree 4, 1000 nodes, with a random congested link and link distances uniform in [1,10].

Figure 13 shows simulations where the simulation scenario is exactly the same as that in Figure 4, except instead of each edge having the same unit length, each edge in the tree is randomly assigned an integer length in [1, 10]. The performance of the request/repair algorithms is somewhat better in the graph with random edge lengths. This is to be expected; the more diverse the distances between pairs of nodes, the easier it is for deterministic suppression to reduce duplicate requests and repairs.

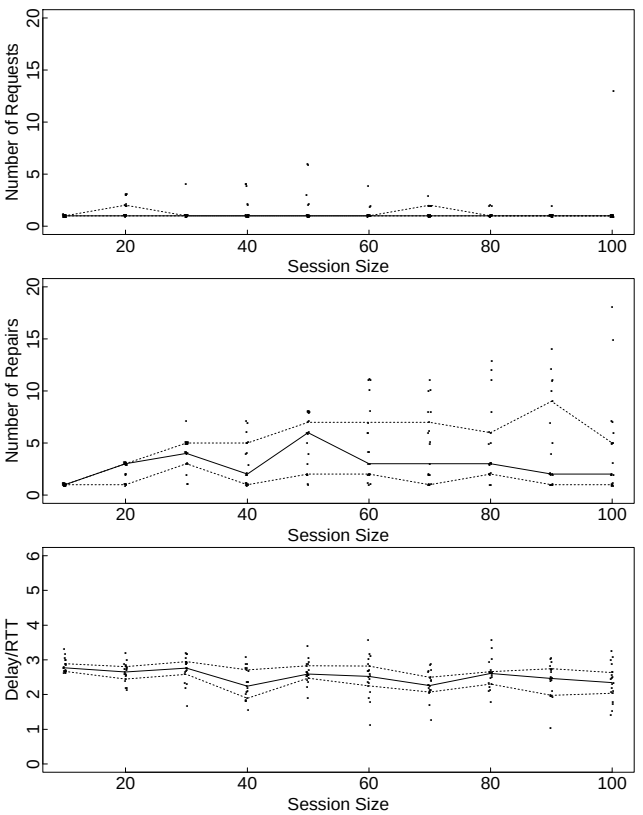

Figure 14: Non-adaptive algorithm for a bounded-degree tree, degree 4, 1000 nodes, with a 5-member ethernet adjacent to each node, with a random congested link.

Figure 14 shows simulations where each of the N nodes in the underlying network is a router with an adjacent ethernet with 5 workstations. For a session of size G, $G/5$ routers are randomly chosen. For each router chosen, all five members of the adjacent ethernet are session members. The congested link is randomly chosen from the point-to-point links between routers.

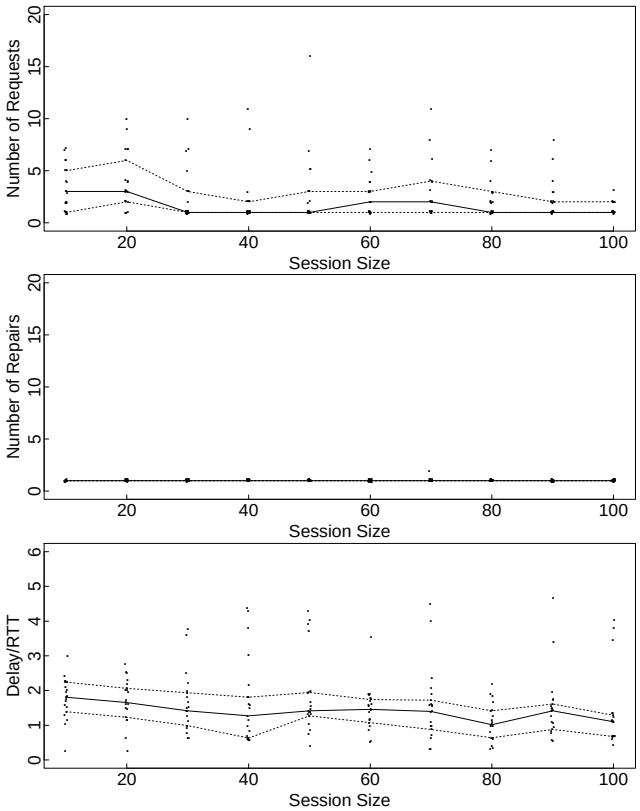

Figure 15: Non-adaptive algorithm for a bounded-degree tree, degree 4, 1000 nodes, with a congested link adjacent to the source.

Figure 15 shows simulations that differ from Figure 4 only in that for Figure 15, the congested link is chosen adjacent to the source of the dropped packet. In Figure 15 the number of repairs is low, but the number of requests is high instead. Note also that in some cases the ratio Delay/RTT is much less than one. This can occur when the distance from the source to the furthest node affected by the dropped packet is large relative to the distance from the source to the sender of the first request. In this case, the furthest node can receive both the request and the answering repair soon after it detects that it is missing the packet.

Figures 21 and 34 show this scenario for simulations using adaptive timer parameter algorithms.

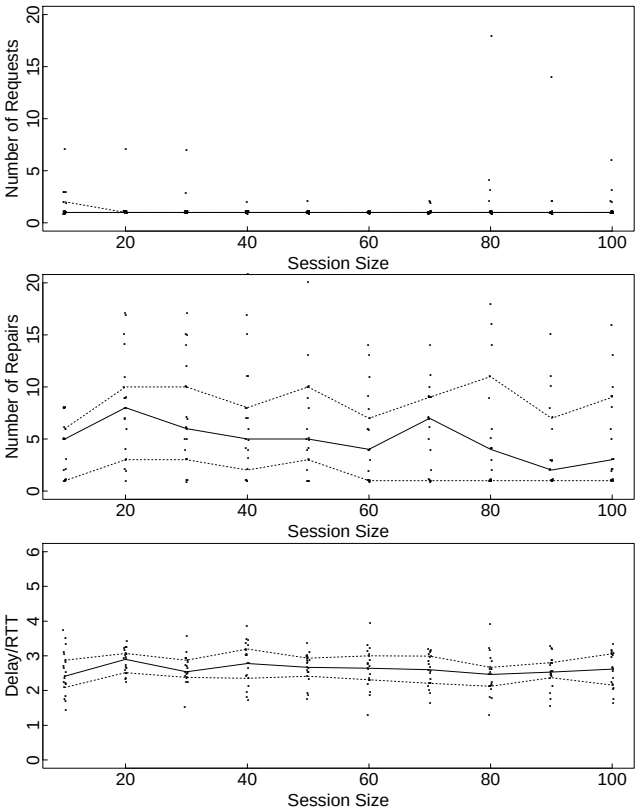

Figure 16: Non-adaptive algorithm for a bounded-degree tree, degree 10, 2000 nodes.

Figure 16 shows the non-adaptive algorithm when the underlying network is a bounded-degree tree of degree 10 with 2000 nodes. Note the high number of duplicate repairs.

Figures 22 and 35 show this scenario for simulations using adaptive timer parameter algorithms.

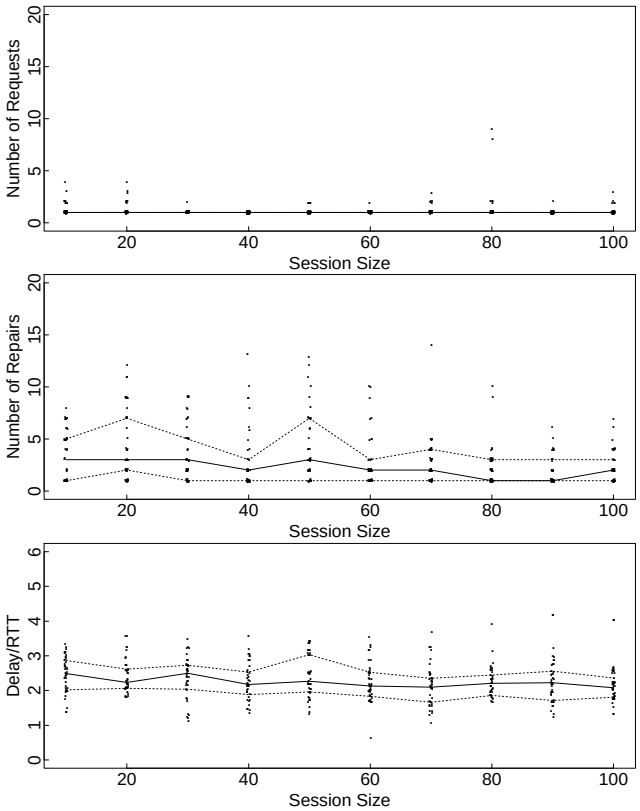

Figure 17: Non-adaptive algorithm for a bounded-degree graph, degree 4, 1000 nodes, 1500 edges, with a random congested link.

Figure 17 shows the request/repair algorithms where the underlying network is more dense than a tree. For this network, with 1000 nodes and 1500 edges, first a 1000-node bounded-degree tree is constructed with degree four for interior nodes. Next, extra edges are added as follows: the first node A is randomly chosen, and then node B is randomly chosen from among the nodes three hops from node A in the graph as constructed so far. The edge (A,B) is added to the graph. This procedure is used so that the additional edges are not insensitive to the geographical constraints of the underlying network.

Because we use shortest-path routing in our simulations, when the network topology is not a tree the data message from the source, the request messages, and the repair messages will not necessarily all follow the same tree. Comparing Figure 17 with Figure 4, the addition of the extra edges to the underlying network does not significantly affect the results of the simulations.

Figure 23 shows this scenario for simulations using adaptive timer parameter algorithms.

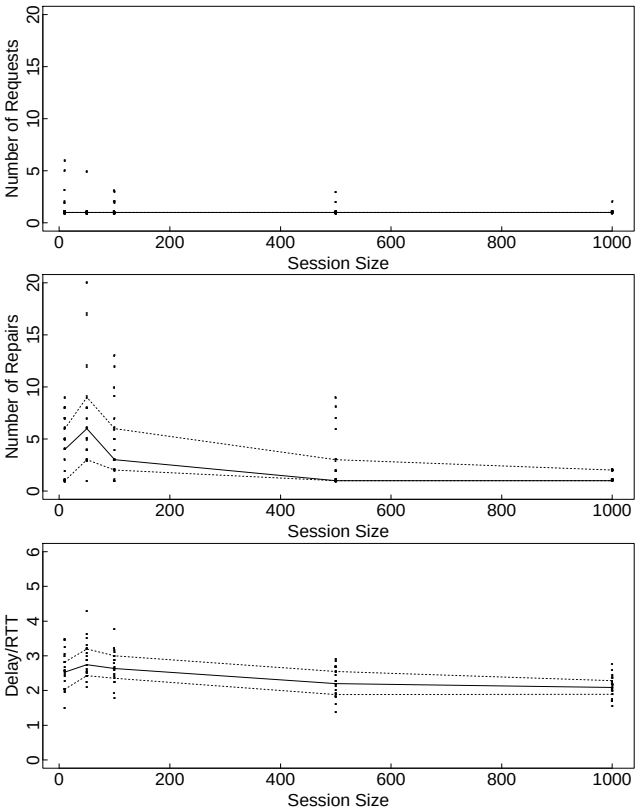

Figure 18: Non-adaptive algorithm for a bounded-degree tree, degree 4, 5000 nodes.

Figure 18 shows the non-adaptive algorithm for a bounded-degree tree of degree 4 on 5000 nodes. Note that when the session size is large relative to the size of the underlying network, the performance of the non-adaptive algorithm is quite good.

Figures 27 and 36 show this scenario for simulations using adaptive timer parameter algorithms.

C Simulations of the duplicates-based adaptive algorithm

This section gives additional simulations on a range of topologies for the *duplicates-based adaptive algorithm* described in Section 6.1. These scenarios include large trees of high degree, networks that are more dense that a simple tree, and networks with a rotating set of failure modes (rather than repeated rounds of the request/repair algorithms with a dropped packet from the same source and on the same congested link). In all of these scenarios the duplicates-based adaptive algorithm out-performs the non-adaptive algorithm.

Appendix D describes an alternate adaptive algorithm, the *delay-based adaptive algorithm*, where the request/repair parameters are adjusted with the primary goal of reducing delay rather than reducing the number of duplicates.

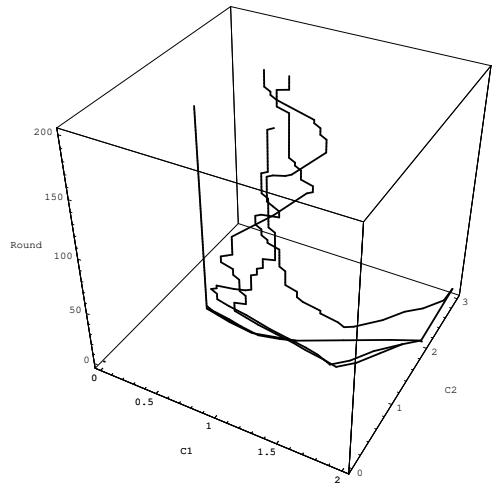

Figure 19: Request timer parameters for three executions of a simulation.

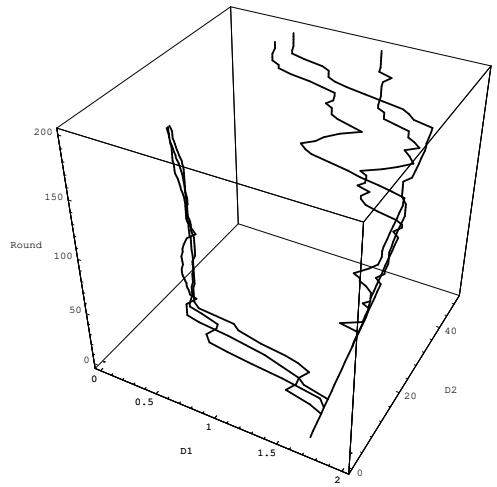

Figure 20: Repair timer parameters for three executions of a simulation.

Figures 19 and 20 show the request and repair timer parameters for three 200-round executions of the simulations in Figure 9, with the duplicates-based adaptive algorithm. For this scenario, the loss neighborhood consists of only two members. For each execution of the simulation, Figure 19 shows the request timer parameters C_1 and C_2 for one of the two session members in the loss neighborhood. The number of duplicate requests can be at most one. The parameters C_1 and C_2 are both slowly decreased from their initial values of 2 to fluctuate near their minimum values.

Figure 20 shows the repair timer parameters D_1 and D_2 for one of the session members not in the loss neighborhood. After the 100th round, the parameter D_2 has almost reached its maximum value of 50, and D_2 remains close to 50 for the remaining rounds. The initial rapid increase of D_2 results in a decrease in the number of duplicate repairs.

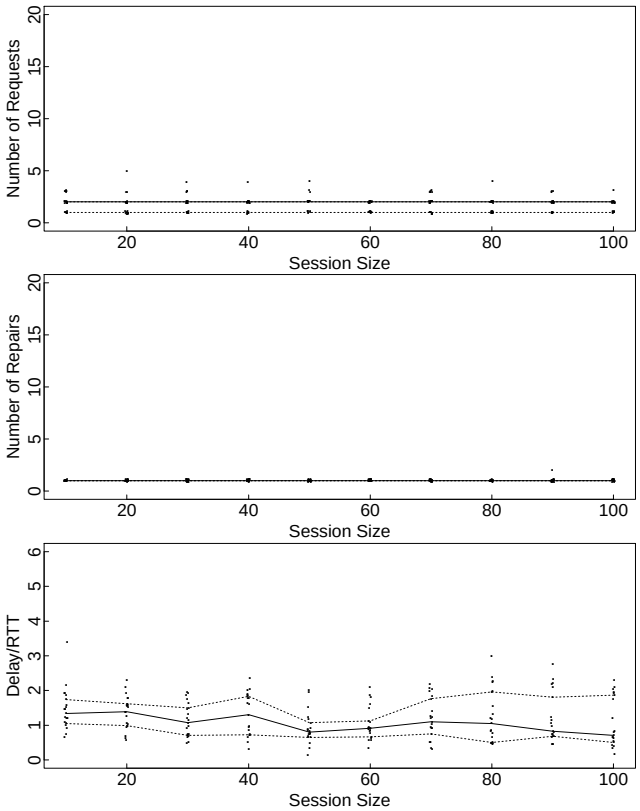

Figure 21: Adaptive algorithm on round 40, for a bounded-degree tree, degree 4, 1000 nodes, with a congested link adjacent to the source.

Figure 21 shows the adaptive algorithm when the congested link is chosen adjacent to the source of the dropped packet. Comparing Figure 21 to the non-adaptive algorithm on the same scenarios in Figure 15 shows the effectiveness of the adaptive algorithm in controlling the number of requests as well as the number of repairs.

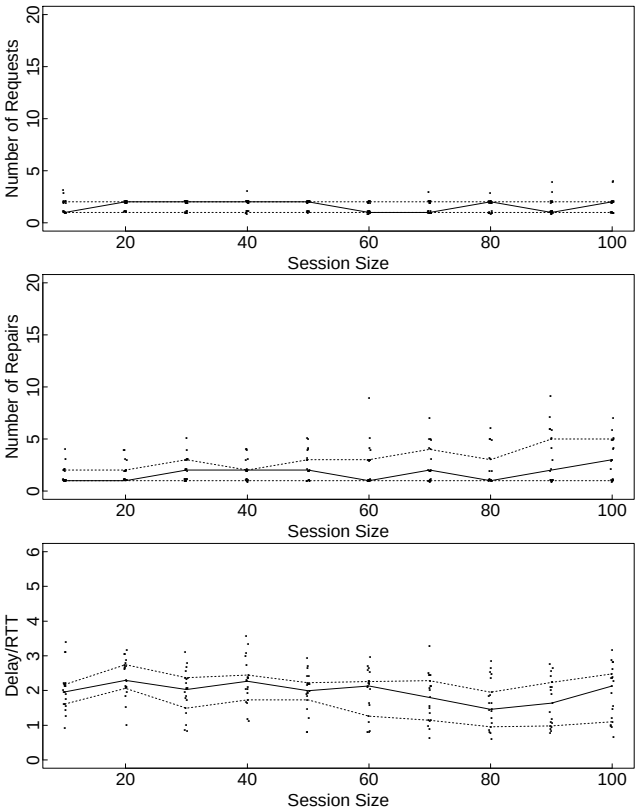

Figure 22: Adaptive algorithm on round 40, for a bounded-degree tree, degree 10, 2000 nodes.

Figure 22 shows the adaptive algorithm for the same scenarios used in Figure 16. Comparing the two figures shows the success of the adaptive algorithm.

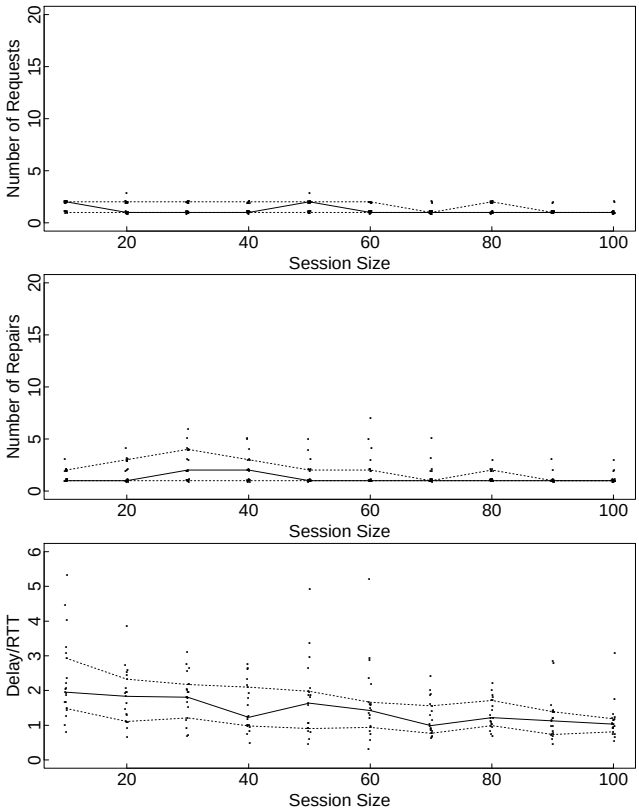

Figure 23: Adaptive algorithm for a graph, 1000 nodes, 1500 edges, on an underlying tree of degree 4.

Figure 23 shows the the adaptive algorithm when the underlying network is not a tree, on the same scenarios used in Figure 17. Comparing Figure 23 with Figure 10 shows that the additional edges in the underlying graph have little effect on the performance of the loss recovery algorithms.

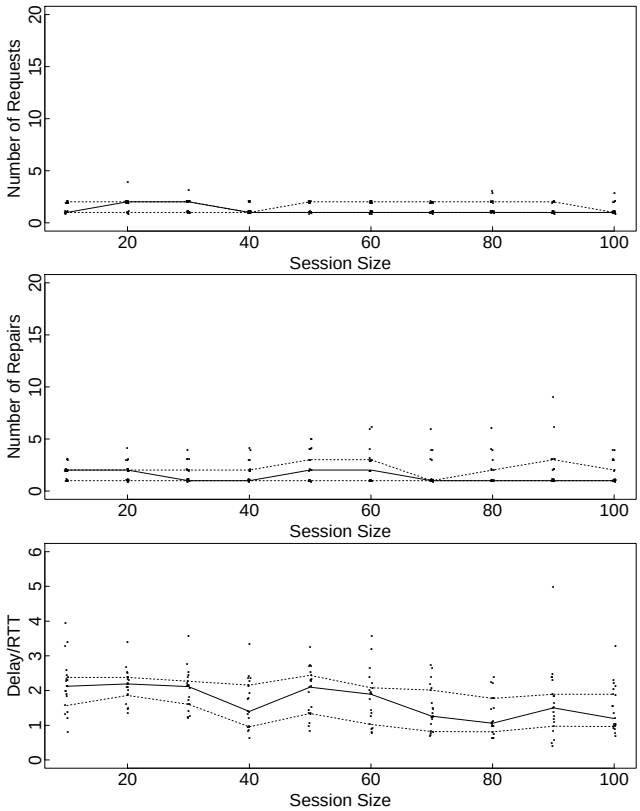

Figure 24: Adaptive algorithm for a bounded-degree tree, degree 4, 1000 nodes, with two alternating failure modes, round 41.

Figure 24 shows the adaptive algorithm with two alternating failure modes. For each failure mode, there is a source and a congested link on which that source's first packet is dropped. On round 1 of the simulation, and on succeeding odd-numbered rounds, a packet is assumed to be dropped according to the first of the two failure modes. On round 2 of the simulation, and on succeeding even-numbered rounds, a packet is assumed to be dropped according to the second failure mode. During each round, a subsequent packet from the source is not dropped, the request/repair algorithms are run, and each session member follows the adaptive algorithm in adjusting its timer parameters.

For each of the simulations, the first failure mode is identical to that in Figure 10. Figure 24 shows the adaptive algorithm on round 41, which uses the first of the two failure modes, and Figure 25 shows the adaptive algorithm on round 40, which uses the second failure mode. (Note that the graphs of Delay/RTT uses different ranges for the y-axis. This is also true of other graphs later in this section.) Comparing Figure 24 with Figure 10 shows that, while the average performance is similar in the two simulation sets, the introduction of alternating failure modes can lead to a higher worst-case delay and number of duplicates.

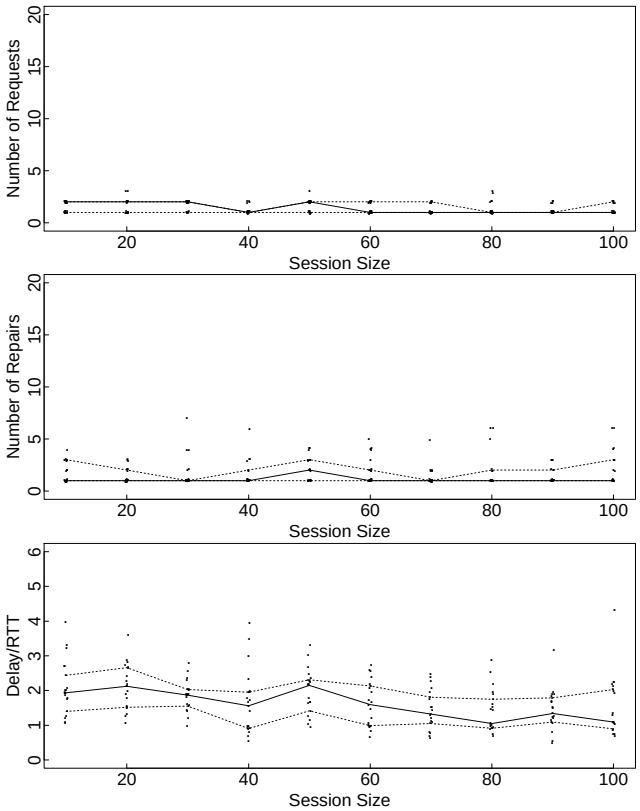

Figure 25: Adaptive algorithm for a bounded-degree tree, degree 4, 1000 nodes, with two alternating failure modes, round 40.

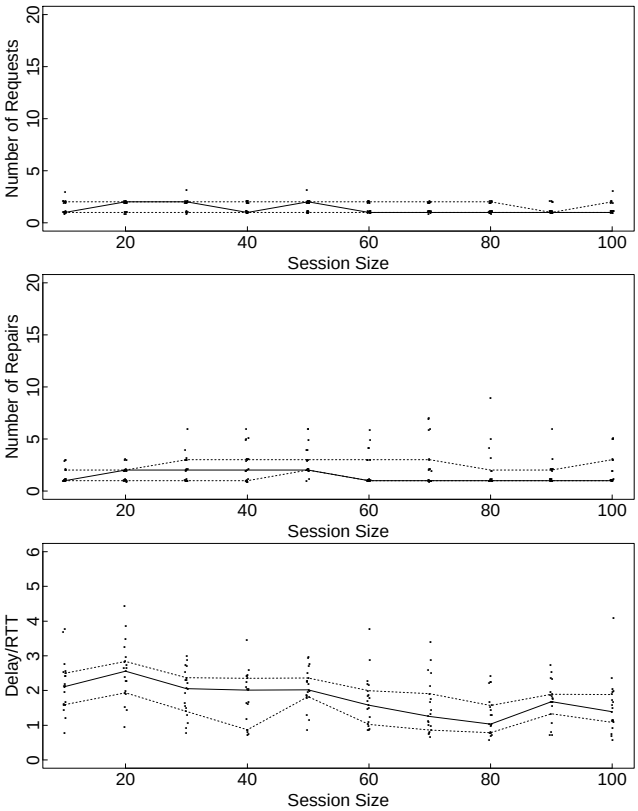

Figure 26: Adaptive algorithm for a bounded-degree tree, degree 4, 1000 nodes, with five alternating congested links, round 41.

For session members that are in the loss neighborhood for only one of the two failure modes, the two failure modes do not interact; the member only adjusts its request timer parameters in one failure mode, and only adjusts its repair timer parameters in the second failure mode. But for session members that are in the loss neighborhood for both failure nodes (or not in the loss neighborhood for both failure modes), the adaptive algorithm is trying to adapt to a changing set of conditions.

Figure 26 shows the adaptive algorithm with five alternating failure modes. This is a more stringent test of the adaptive algorithm; the adaptive algorithm performs best with a stable packet drop scenario. Comparing Figure 26 with Figure 4, however, shows that even with five alternating failure modes, the adaptive algorithm gives better performance that the nonadaptive algorithm.

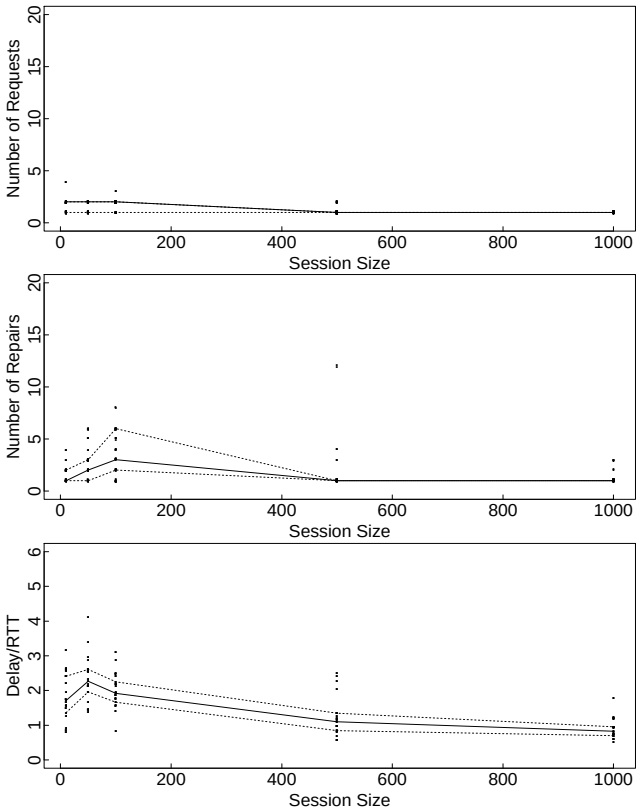

Figure 27: Adaptive algorithm for a bounded-degree tree, degree 4, 5000 nodes.

Figure 27 shows the adaptive algorithm for a bounded-degree tree of degree 4 on 5000 nodes. Comparing this with the non-adaptive algorithm in Figure 18 shows that for small sessions the adaptive algorithm helps to reduce the number of repairs. For large sessions, the performance is good either with or without the adaptive algorithm (although the delay is lower with the adaptive algorithm).

D A delay-based adaptive algorithm

This section describes a *delay-based adaptive algorithm*, where the request/repair timer parameters are adjusted with the primary goal of reducing delay rather than reducing the number of duplicates. This differs from the duplicates-based adaptive algorithm described in Section 6.1 in several respects.

The first major change is that each member that sends a request decreases C_1 slightly. The result is that members that have frequently sent requests in the past will send requests somewhat sooner in the future. Given a stable loss scenario, this helps to reduce both the delay and the number of duplicate requests.

The second major change is that the delay-based adaptive algorithm looks at delay first and numbers of duplicates second. That is, if *ave_req_delay* and *ave_dup_req* are both unacceptably large, the delay-based adaptive algorithm decreases C_1 to reduce delay, but does not increase C_2. While a larger C_2 generally reduces the number of duplicates, it also increases delay.

As the simulations in this section show, the delay-based adaptive algorithm generally outperforms the duplicates-based adaptive algorithm in both the delay and the number of duplicates.

```
after sending a request
   decrease start of req. timer interval
if ave. req. delay too high
   decrease start of req. timer interval
   if ave. dup. requests okay
      also decrease req. timer interval
else if ave. dup. requests too high
   increase req. timer interval
   if ave. req. delay okay
      decrease start of req. timer interval
else if both ave. req. delay
      and ave. dup. requests okay
   decrease req. timer interval
```

Figure 28: Overview of delay-based adjustment algorithm for request timer parameters.

Figures 28 and 29 give the delay-based adjustment algorithm for setting the request timer parameters. A similar delay-based adjustment algorithm is used for setting the repair timer parameters.

The delay-based adaptive algorithm makes additive changes of -0.05 and $+0.05$ to C_1, and additive changes of -0.1 and $+0.5$ to C_2. Because C_1 starts at 2, and can only vary between 0.5 and 2, small changes to C_1 are appropriate. On the other hand, C_2 starts at 2 and can vary between 1 and the session size G. Thus, the incremental changes are larger, and the increase of $+0.5$ allows C_2 to increase fairly rapidly.

```
after sending a request
    C1− = 0.05
if (ave_req_delay ≥ AveReqDelay)
    C1− = 0.05
    if (ave_dup_req < AveDups − ε))
        C2− = 0.1
else if (ave_dup_req ≥ AveDups))
    C2+ = 0.5
    if (ave_req_delay < AveReqDelay − ε)
        C1+ = 0.05
else if ((ave_dup_req < AveDups − ε) &&
    (ave_req_delay < AveReqDelay − ε))
    C2− = 0.1
```

Figure 29: Delay-based adjustment algorithm for request timer parameters. In our simulator $\epsilon = 0.1$.

The simulations show that for the delay-based adjustment algorithm compares favorably with the duplicates-based adjustment algorithm described earlier, particular for a fixed failure mode. We emphasize that, while these simulations illustrate the effectiveness of performance-based adaptive algorithms in adjusting the request and repair timer parameters, there are still many open questions in designing optimal adaptive algorithms.

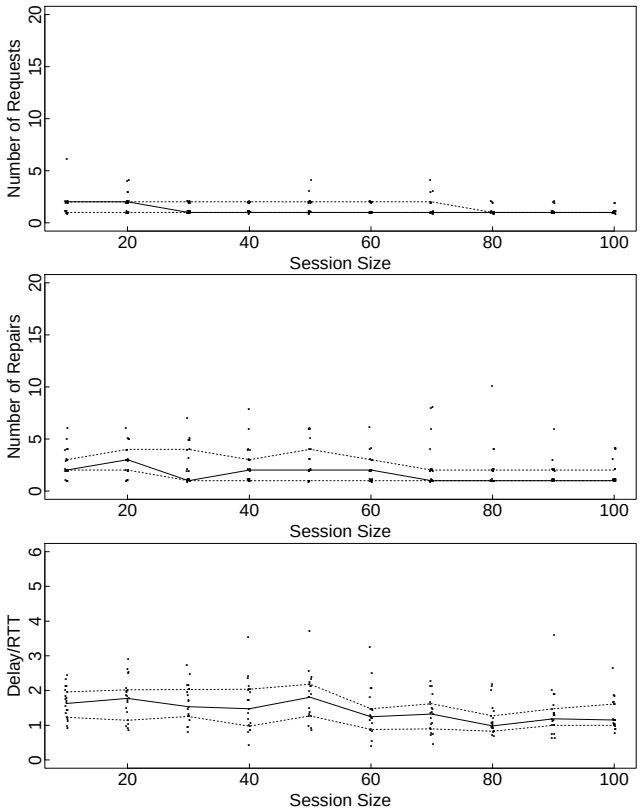

Figure 30: Delay-based adaptive algorithm for a bounded-degree tree, degree 4, 1000 nodes. $AveReqDelay = 1$, $AveRepDelay = 2$, $AveDups = 3$.

Figure 30 shows the results of the delay-based adaptive algorithm, with the following parameters: $AveDups = 3$, $AveReqDelay = 1$, and $AveRepDelay = 2$. Comparing Figure 30 with either the non-adaptive algorithm in Figure 4 or the duplicates-based adaptive algorithm in Figure 10 shows that when there is a single repeated failure mode, the delay-based adaptive algorithm is successful in giving low delay with an acceptable number of duplicate packets.

One reason to use a higher value for $AveRepDelay$ than for $AveReqDelay$ is that repairs usually come from a member near the source of the request, so the delay for waiting for a member to send a repair after receiving a request is usually a small component of the total delay, compared to the delay of waiting for some member to first send a request.

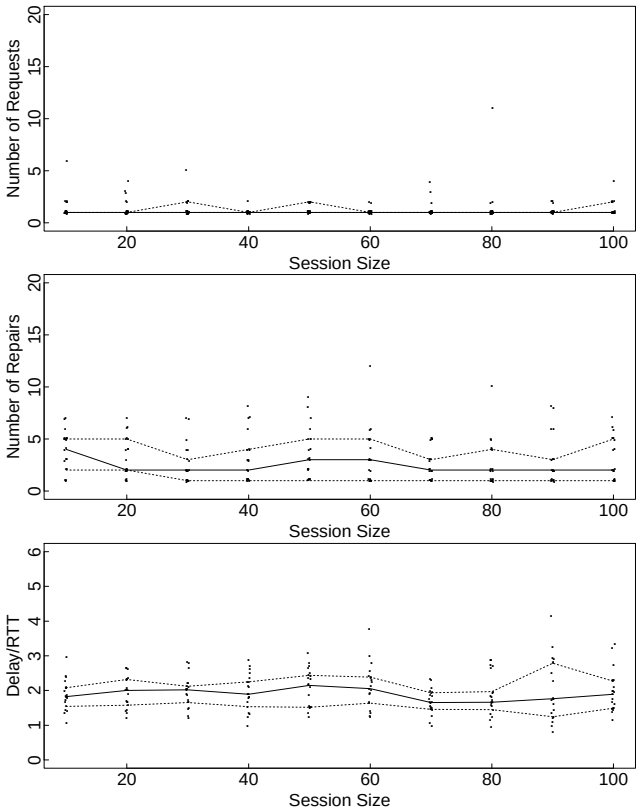

Figure 31: Delay-based adaptive algorithm for a bounded-degree tree, degree 4, 1000 nodes, with five alternating errors. $AveReqDelay = 1$, $AveRepDelay = 2$, $AveDups = 3$.

Figure 31 shows the results of the delay-based adaptive algorithm when there are five alternating failure modes (as opposed to the single failure mode in Figure 30). Comparing Figure 31 and Figure 30 shows that the presence of alternating failure modes limits the effectiveness of the delay-based adaptive algorithm. Comparing Figure 31 with the non-adaptive algorithm in Figure 4 shows that even with five alternating failure modes, the delay-based adaptive algorithm outperforms the non-adaptive algorithm. Comparing Figure 31 with the delay-based adaptive algorithm in Figure 26, however, shows that for five alternating failure modes, while the delay-based adaptive algorithm outperforms the duplicates-based algorithm in terms of delay, the delay-based adaptive algorithm can have a higher worst-case number of duplicates.

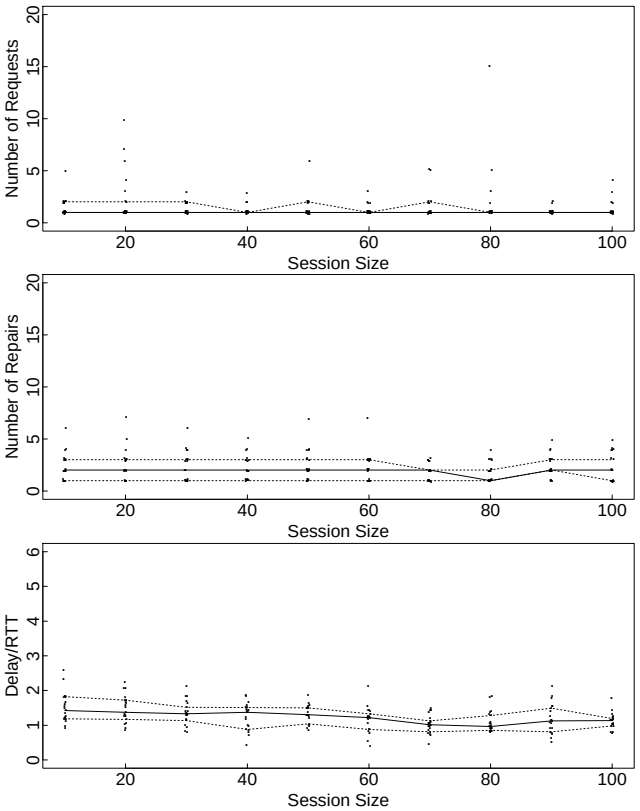

Figure 32: Delay-based adaptive algorithm for a bounded-degree tree, degree 4, 1000 nodes. $AveReqDelay = 0.5$, $AveRepDelay = 1$, $AveDups = 3$.

Figures 32 and 33 show the delay-based algorithm with a single failure mode, with different values for the parameters $AveReqDelay$, $AveRepDelay$, and $AveDups$. In Figure 32 the delay-based adaptive algorithm uses the parameters $AveReqDelay = 0.5$, $AveRepDelay = 1$, and $AveDups = 3$. Comparing Figure 32 with Figure 30, the lower goals for average delay do result in a lower average delay, with little effect on the average number of duplicates.

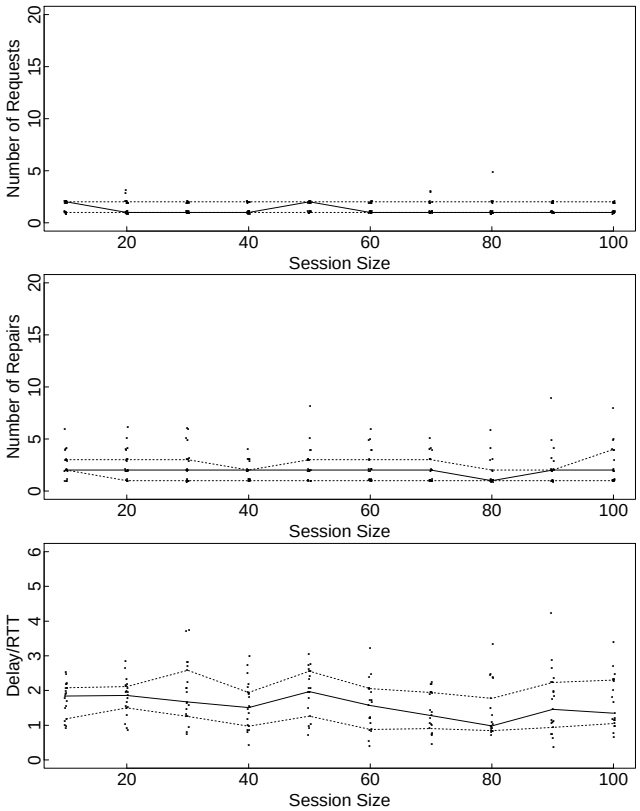

Figure 33: Delay-based adaptive algorithm for a bounded-degree tree, degree 4, 1000 nodes. $AveReqDelay = 1.5$, $AveRepDelay = 3$, $AveDups = 2$.

Figure 33 shows the delay-based adaptive algorithm with parameters $AveReqDelay = 1.5$, $AveRepDelay = 3$, and $AveDups = 2$. Comparing Figure 33 with Figure 32 shows that the looser delay goals in this figure results in higher delay, but also in improved worst-case values for the number of duplicate requests.

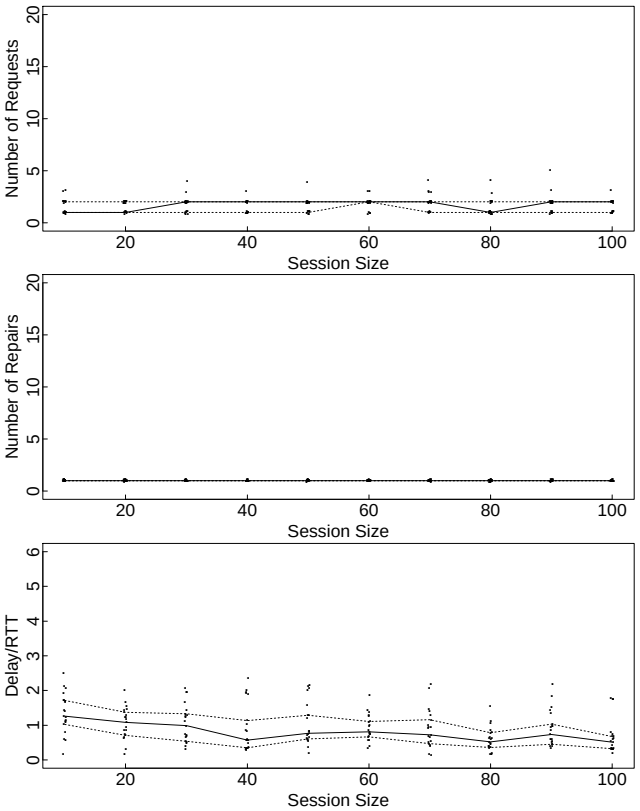

Figure 34: Delay-based adaptive algorithm for a bounded-degree tree, degree 4, 1000 nodes, with a congested link adjacent to the source. $AveReqDelay = 1$, $AveRepDelay = 2$, $AveDups = 3$.

Figure 34 shows the delay-based adaptive algorithm for a bounded-degree tree with a congested link adjacent to the source. This compares favorably to both Figure 15, which shows the non-adaptive algorithm, and Figure 21, which shows the duplicates-based adaptive algorithm.

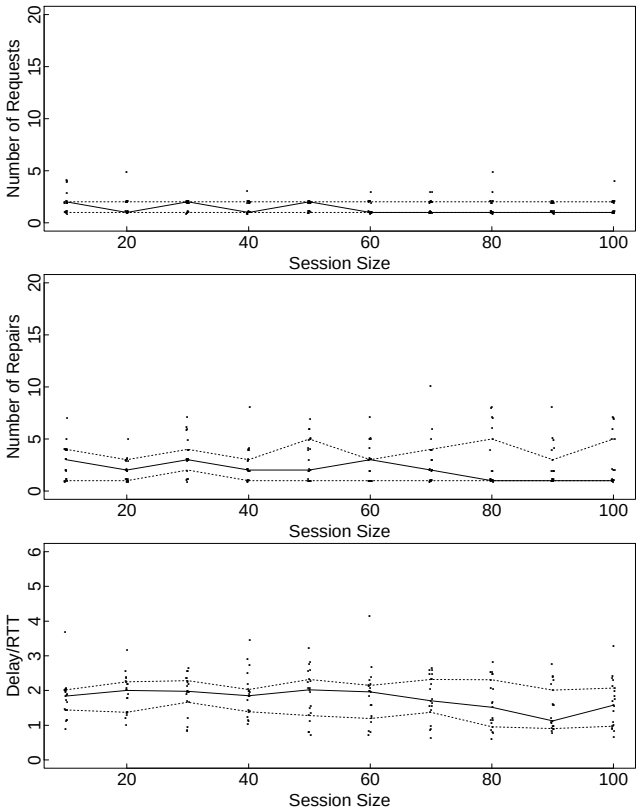

Figure 35: Delay-based adaptive algorithm for a bounded-degree tree, degree 10, 2000 nodes. $AveReqDelay = 1$, $AveRepDelay = 2$, $AveDups = 3$.

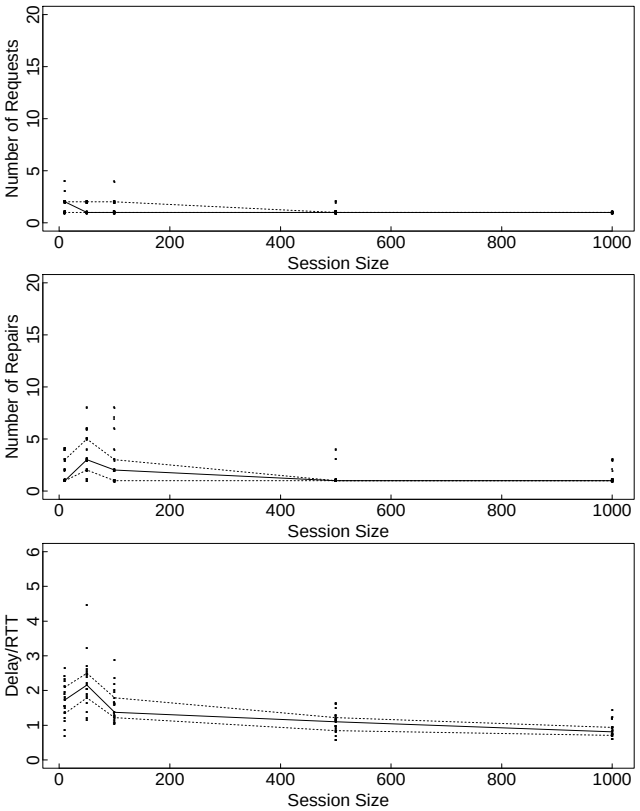

Figure 36: Delay-based adaptive algorithm for a bounded-degree tree, degree 4, 5000 nodes. $AveReqDelay = 1$, $AveRepDelay = 2$, $AveDups = 3$.

Figure 35 shows the delay-based adaptive algorithm for a bounded-degree tree, degree 10, with 2000 nodes. This compares favorably to Figure 16, which shows the non-adaptive algorithm for this scenario. Figure 35 differs from Figure 22, which shows the duplicates-based adaptive algorithm, mainly in the absence of outlier simulations with a Delay/RTT ratio of five or more.

Figure 36 shows the delay-based adaptive algorithm for a bounded-degree tree, degree 4, with 5000 nodes. This compares favorably to Figure 18, which shows the non-adaptive algorithm. In comparison with Figure 27, which shows the duplicates-based adaptive algorithm, the delay is improved, and there is some difference in the number of duplicates.

E Local recovery

In this section we use simulations to explore the optimal behavior that could be achieved from the local recovery algorithms described in Section 6.2. First we examine networks were all links have a link threshold of one, and next we examine networks with a range of values for the link thresholds.

The simulations suggest that even assuming an optimal execution of the local recovery algorithms, the effectiveness of local recovery in limiting the unnecessary use of bandwidth depends heavily on the topology of the underlying network. A simple hop-count-based local recovery algorithm in an arbitrary topology is not always effective. However, for a topology with naturally-defined *neighborhoods*, and some mechanism for restricting request and repair messages to these neighborhoods, local recovery algorithms can be quite effective. In this section we explore local recovery based on *threshold-defined neighborhoods*, where the neighborhood is defined by a high-threshold link separating the neighborhood from the rest of the multicast network. Similar results should apply to local recovery based on neighborhoods defined by administrative scoping, or on neighborhoods defined by the boundaries of hierarchical multicasting.

To explore the optimal possible performance of TTL-based local recovery algorithms, we assume that the loss neighborhood is stable, and that members have some method for estimating h_1 and h_2, where h_1 is the minimum TTL needed to reach all members in the loss neighborhood, and h_2 is the minimum TTL needed to reach some member not in the loss neighborhood. Further, we assume that for each loss recovery event, the request/repair algorithms exhibit their optimal behavior. That is, we assume that there is a single request and a single repair, and that both come from the members closest to the point of failure. We restrict attention to scenarios where the loss neighborhood contains at most 1/10-th of the session members.

In the current Mbone, each link (more precisely, each interface or tunnel) is assigned a *threshold*, with a default threshold of one. The threshold is the minimum TTL required for a multicast packet to be forwarded on that link, and is used to control the scope of multicast packets. Every multicast router decrements the TTL of a forwarded packet by one.

Figure 37 shows the results of an optimal execution of the local recovery algorithms in a random tree with link thresholds of one, where all network nodes are session members. For this scenario, local recovery is quite effective in reducing the unnecessary use of bandwidth. In the top graph of Figure 37, the x-axis shows the session size, and the y-axis shows the fraction of session members reached by the repair. In the bottom graph of Figure 37, the y-axis shows the number of session members in the *repair neighborhood*, that is, the number of session members reached by the repair, as a multiple of the number of members in the loss neighborhood. For each session size, twenty simulations are run, each with

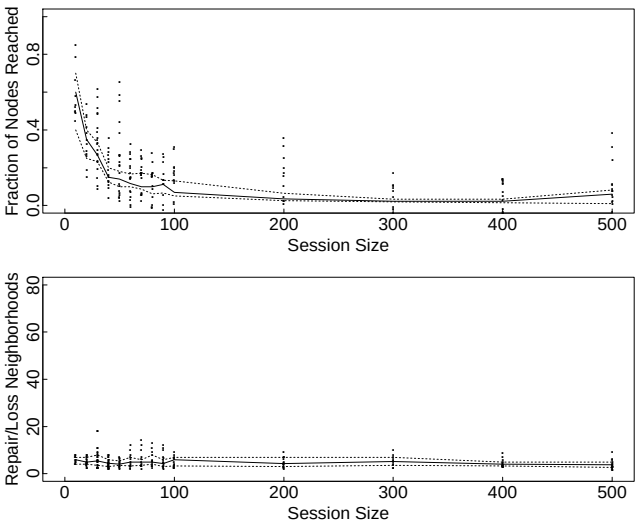

Figure 37: Local recovery in random trees, all nodes as session members, thresholds of one.

a different random tree (if applicable), session membership, and source and congested link for the dropped packet. The results of each simulation are represented by a jittered dot. The three lines indicate the first, second, and third quartiles.

For a random tree, the median ratio between the repair and the loss neighborhoods ranges between 4 to 6. Thus, while for this scenario the use of local recovery does not completely *eliminate* the unnecessary use of bandwidth, it does considerably *reduce* the unnecessary use of bandwidth.

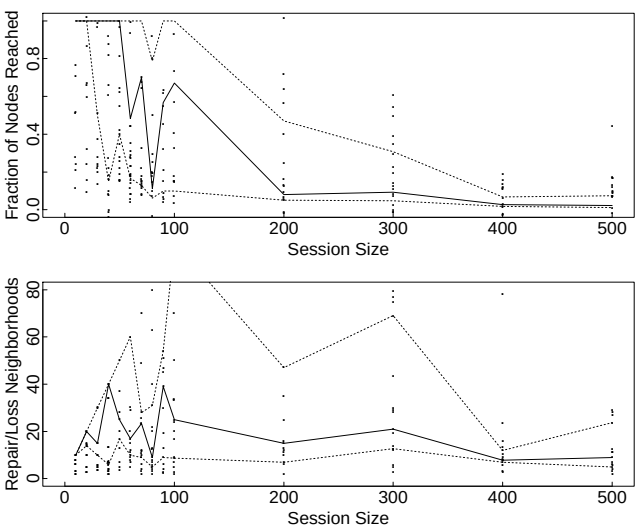

Figure 38: Local recovery in bounded-degree trees with 1000 nodes, thresholds of one.

In contrast, Figure 38 shows the results of local recovery in a large bounded-degree network of degree four, with link

thresholds of one. This scenario differs from that in Figure 37 in two respects: the underlying network is considerably larger than the session itself, and in addition the underlying network is of high average degree. In this case, when the session itself is fairly small relative to the size of the underlying network, the benefits of pure hop-count-based local recovery are small. For sessions with 100 members, for example, the median ratio between the repair neighborhood and the loss neighborhood is 25, and the median repair neighborhood contains two-thirds of the nodes in the multicast session.

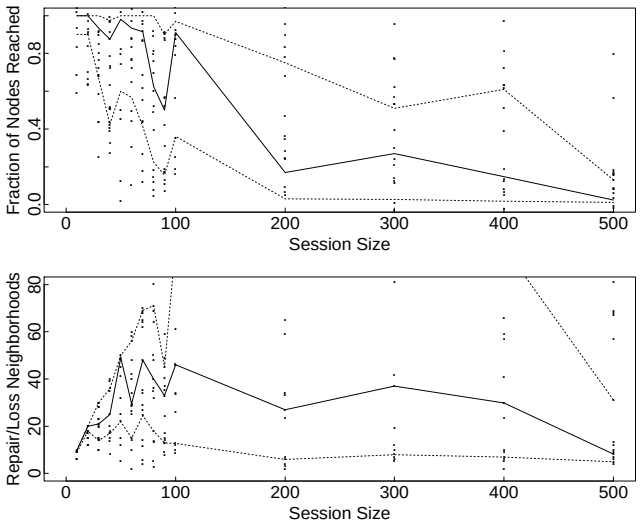

Figure 39: Local recovery in a mesh-based graph with 1000 nodes, thresholds of one.

To introduce a network with a range of link thresholds, we create a mesh-based network as in Figure 40, with a central mesh and subtrees attached to each node in the mesh. We chose this topology because it seems to match the topology of the Mbone, which can be considered as consisting of naturally-defined neighborhoods separated by transcontinental links and by backbone links within the US. One way to construct a local recovery scheme that allows the local recovery to follow these naturally-defined neighborhoods is to use a TTL-based local recovery scheme in a topology where neighborhoods are separated by high-threshold links. As mentioned above, other ways to use neighborhoods would be to use administrative scoping for local recovery, or to make use of hierarchical multicast routing algorithms.

To create the 1000-node mesh-based network, first a mesh of 19 nodes is created, with three rows of nodes. Each node in the mesh is then made the root of a 53-node bounded-degree tree where each node has degree four within the tree. The last node in the mesh has a slightly smaller attached tree to make 1000 nodes in all. Figure 39 shows the results of local recovery in a mesh-based graph with 1000 nodes and link thresholds of one; the results are similar to those of bounded-degree trees in Figure 38.

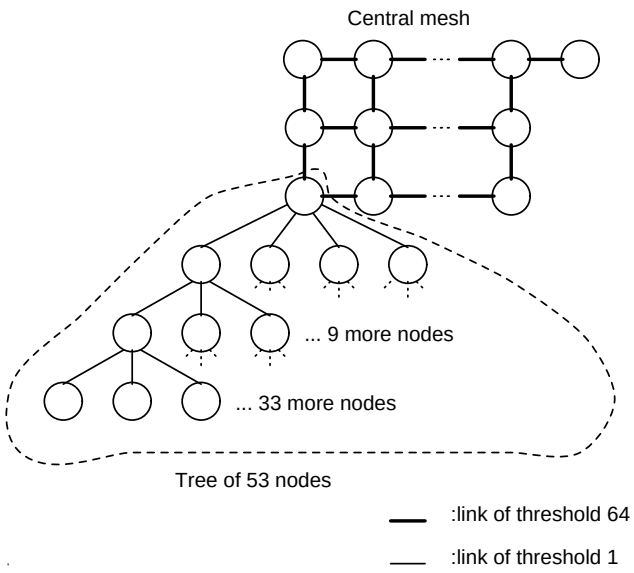

Figure 40: A mesh-based graph with thresholds of one and 64.

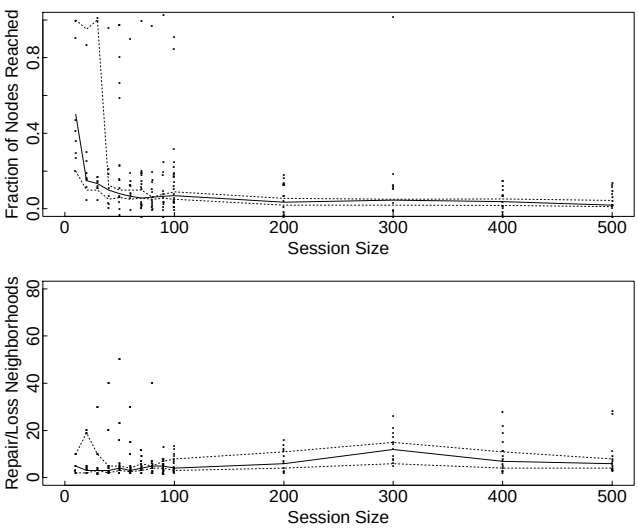

Figure 41: Local recovery in a mesh-based graph with 1000 nodes, thresholds of one and 64.

Figure 40 shows a mesh-based graph with thresholds of 64 for links in the central mesh. In contrast to Figure 39, Figure 41 shows that local recovery can be particularly effective in such an environment, where regions of the network are partitioned from each other by paths with a high threshold.

We call a region a *threshold-defined neighborhood* if the diameter of the neighborhood (in terms of the TTL required to get from one node to another) is less than the threshold of the links separating that neighborhood from the rest of the

network. Given a threshold-defined neighborhood, local recovery can be used to ensure that request and repair messages for losses within the neighborhood do not "leak out" and affect session members outside of the neighborhood. Note that this is only effective if the threshold-defined neighborhood contains not only the entire loss neighborhood, but also some session member that is not in the loss neighborhood.

For the simulations in Figure 41, when the loss occurs within a subtree and both the loss neighborhood and some node not in the loss neighborhood are contained in the subtree, then the requests and repairs are naturally restricted to the threshold-defined neighborhood (i.e., the subtree) through the operations of the local recovery algorithm. Because the TTL required to get from one node to another within the subtree is at most 31, local-recovery requests for this scenario are naturally sent with TTL at most 31 (following the TTL-based local recovery algorithms), local-recovery repairs are sent with TTL at most 62, and local-recovery requests and repairs are contained within the subtree. Note that this does not require that session members have any knowledge of either the topology or of the existence of the subtree.

For an environment where members have explicit knowledge of the existence of a threshold-defined neighborhood, the local recovery algorithm can be simplified. If past history indicates both that losses are likely to have occurred within the threshold-defined neighborhood and that the threshold-defined neighborhood also is likely to contain a member capable of answering requests, then all that is required is that both the local-recovery requests and the local-recovery repairs are sent with a TTL at least one less than the threshold separating the neighborhood from the rest of the network. In this case, the request and repair traffic will be contained within the threshold-defined neighborhood. Similarly simple algorithms are possible if members have explicit knowledge of the existence of an administratively-defined neighborhood, or of a neighborhood defined by a hierarchical multicast routing algorithm.

Measurements of packet losses for multicast audio traffic in the current Mbone [YKT95] suggest that packet loss in multicast traffic is most likely to occur not in the "backbone", but in the "edges" on the multicast network. This is exactly the scenario for which local recovery would be of most benefit.

E.1 Determining membership in naturally-defined neighborhoods

For a local-recovery algorithm based on the members' explicit knowledge of naturally-defined neighborhoods (e.g., neighborhoods defined by separating links of threshold 64, or neighborhoods defined by administrative scope), then member A needs some method to determine if member B is contained within the same naturally-defined neighborhood. As an example, for a topology with neighborhoods of diameter (in terms of hop counts) at most 31 defined by separating links of threshold 64, members could occasionally send session messages with a TTL of 31 to identify other members in the same threshold-defined neighborhood.

For threshold-based local recovery, session and control packets should contain not only the modifiable TTL field, but also the original unmodified TTL. A member A answering a request from member B needs to estimate h_3, the original TTL of that request. This is easy if requests carry the original unmodified TTL. A member A, on receiving a request from member B, needs to estimate the minimum number of hops needed to reach B. Again, assuming paths with symmetric hop counts, this can be done by comparing the modified TTL in the request packet with the original unmodified TTL.

A member before sending a request needs to estimate h_1, the minimum TTL to reach all other members in the loss neighborhood, and h_2, the minimum TTL to reach some member not in the loss neighborhood. As an example, consider the mesh-based graph in Figure 40, where a TTL of 31 is sufficient for a path between any two nodes within the same subtree, and members occasionally send session messages with a TTL of 31. Then if the loss neighborhood and a member not in the loss neighborhood are contained within the same subtree, the local-recovery request will be sent with a TTL of at most 31, the repair will be sent with a TTL at most twice that of the original request, and the request and repair will both be contained within the subtree.

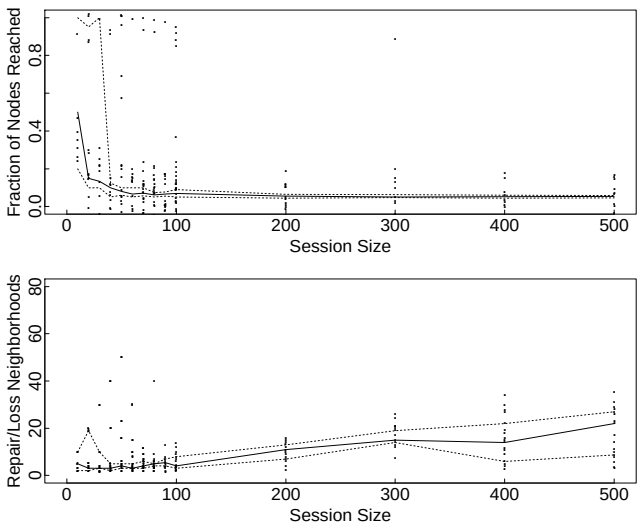

Figure 42: Local recovery in a mesh-based graph with 1000 nodes, thresholds of one and 64, and a TTL of 31 for local-recovery requests.

Figure 42 shows the performance of the local recovery algorithms with such a rough mechanism for determining the other nodes reachable by a TTL of 31. In this case, even if the loss neighborhood consists of a single member with a

nearby member not in the loss neighborhood, the request will be sent with a TTL of 31, and the repair will reach all of the members in that threshold-defined neighborhood. As Figure 42 shows, the only penalty for using this rough mechanism for determining TTLs, rather than using a more precise mechanism, is that for dense groups local recovery repairs reach all of the members in the subtree, instead of being restricted to only one area of the subtree.

E.2 Estimating TTLs

This section explores in some detail the additional mechanisms that would be needed for TTL-based local recovery for members to determine more precisely the *minimum* TTL needed to reach other members. This would be necessary for TTL-based local recovery if there were no mechanism such as session messages sent with limited scope for determining membership in the threshold-defined neighborhood, or if it was desired for determining more precisely the local neighborhood. In this case member A, on receiving a session message or repair from member B, needs to be able to estimate the minimum TTL needed to reach B.

If all links have a link threshold of one, then from the original unmodified TTL t_0 and the modified TTL field t_1 on a packet received from B, member A can determine precisely the minimum TTL $t_0 - t_1$ needed to reach A from B. However, for networks with a range of link thresholds, this cannot be done. A packet from one hop away, sent with TTL 64, will be received with a TTL of 63 whether the link threshold was 1 or 64.

For networks with a range of link thresholds, some additional mechanisms are necessary if member A needs to determine more precisely the *minimum* TTL needed to reach A from B. As one possibility, the session, repair, and request packets could carry a modifiable field indicating the *minimum TTL* needed to reach the current router from the packet's source. Such a field would not be difficult to update at each router, but it would require the active updating by the multicast routers. For a multicast router that receives a message reporting a current TTL (before being decremented by the router) of t, an original TTL of t_0, and a minimum TTL needed to reach the previous router of min, the "minimum TTL" field is not changed if min is at least $th + (t_0 - t)$, for th the link threshold of the link from the previous router. Otherwise, the "minimum TTL" field is set to $th + (t_0 - t)$. The simulations in Figure 41 use this mechanism to determine the minimum TTL needed to reach A from B.

E.3 Robustness of the local recovery algorithms

So far, we have only explored optimal executions of the local recovery algorithm, where exactly one request is sent, answered by exactly one repair. A non-optimal execution with multiple local-recovery requests or repairs can be somewhat more complex.

One way that the local recovery algorithm can "fail" is if the scope of the original request is not sufficient to reach all members sharing that loss. In this case, there will be duplicate requests, and each request might reach a different group of members. One possibility is that a loss with a large loss neighborhood could result in many local-recovery-based loss recovery events, instead of a single global-scope loss recovery event. In this case the multiple local-recovery-based loss recoveries could end up using more bandwidth than would have been used by a single global loss recovery.

A second way that the local recovery algorithm can "fail" is if the scope of the original request is not sufficient to reach any potential responders. In this case, a second request will be sent with global scope when some member's backed-off request timer subsequently expires. Thus, the penalties of a failed local recovery can include both increased delay and wasted bandwidth in the first iteration of the local recovery algorithm.

Another possible complication is that a node B capable of answering a request could receive multiple requests, each one of which would normally be answered by a repair with a different TTL. In this case node B should send the repair with the largest of the candidate TTLs. Node B could subsequently receive a request from node A that would normally be answered by yet a larger TTL. If the earlier repair sent by node B has a sufficiently large TTL to reach node A, then node B should probably ignore A's request, and wait to give time for its earlier repair to be delivered. If some node backed off its own request timer because of the request from A, and does not receive the earlier repair from node B, then that node can resend a request when its backed-off timer expires.