

DeTail: Reducing the Flow Completion Time Tail in Datacenter Networks

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ABSTRACT

Web applications have now become so sophisticated that rendering a typical page may require hundreds of intra-datacenter flows. At the same time, web sites must meet strict page creation deadlines of 200-300ms to satisfy user demands for interactivity. Long-tailed flow completion times make it challenging for web sites to meet these constraints. They are forced to choose between rendering a subset of the complex page, or delay its rendering, thus missing deadlines and sacrificing either quality or responsiveness. Either option leads to potential financial loss.

In this paper, we present a new cross-layer network stack aimed at reducing the long tail of flow completion times. The approach exploits cross-layer information to reduce packet drops, prioritize latency-sensitive flows, and evenly distribute network load, effectively reducing the long tail of flow completion times. We evaluate our approach through NS-3 based simulation and Click-based implementation demonstrating our ability to consistently reduce the tail across a wide range of workloads. We commonly achieve reductions of over 50% in 99.9th percentile flow completion times without significantly impacting the median.

1. INTRODUCTION

Web sites have grown complex in their quest to provide increasingly rich and dynamic content. A typical Facebook page consists of a timeline-organized “wall” that is writable by the user and her friends, a real-time cascade of friend event notifications, a chat application listing friends currently on-line, and of course, advertisements selected by displayed content. Modern web pages such as these are made up of many components, generated by independent subsystems and “mixed” together to provide a rich presentation of information.

Building such systems is not easy. They exploit high-level parallelism to assemble the independent page parts in a timely fashion, and present these incrementally, subject to deadlines to provide an interactive response. The final mixing system must wait for all subsystems to deliver some of their content, potentially sacrificing responsiveness if a small number of subsystems are delayed. Alternatively, it must present what

it has at the deadline, sacrificing page quality and wasting resources consumed in creating parts of a page that a user never sees.

In this paper, we investigate how the network complicates such application construction, because of the high variation in performance of the network flows underlying their distributed workflows. By improving the statistics of network flow completion, in particular *by reducing the long flow completion tail*, the application gains better worst case performance from the network. Applying the end-to-end principle, while the mixer software must still deal with subsystems that fail to respond by the deadline, an underlying network that yields better flow statistics reduces the conservativeness of time-outs while reducing the frequency with which they are triggered. The ultimate application-layer result is better quality and responsiveness of presented pages.

Deadlines are an essential constraint on how these systems are constructed. Experiments at Amazon [25] demonstrated that failing to provide a highly interactive web site leads to significant financial loss. Increasing page presentation times by as little as 100 ms significantly reduces user satisfaction. To meet these demands, web sites seek to meet deadlines of 200-300 ms 99.9% of the time [13, 33].

Highly variable flow completion times complicate the meeting of interactivity deadlines. Application workflows that span the network depend on the performance of the underlying network flows. Packet arrival pacing is dictated by roundtrip delays and congestion can significantly affect performance. While datacenter network roundtrip times can be as low as $250\mu s$, in the presence of congestion, these delays can grow by two orders of magnitude, forming a long tail distribution [13]. Average roundtrip times of hundreds of microseconds can occasionally take tens of milliseconds, with implications for how long a mixer application must wait before timing-out on receiving results from its subsystems.

Flash congestion is the culprit and it cannot be managed through conventional transport-layer means. Traffic bursts commonly cause packet losses and retransmissions [13]. Un-even load balancing often causes a subset of flows to experience unnecessarily high congestion [12]. The absence of traffic prioritization causes latency-sensitive foreground

flows to wait behind latency-insensitive background flows [33]. Each contributes to increasing the long tail of flow completion. While partial solutions exist [12, 13, 29, 33], no existing approach solves the whole problem. Fortunately, datacenter networks already contain the key enabling technology to reduce the long flow completion tail. They employ high-speed links and a scaled-out network topology, providing multiple paths between every source and destination [11, 22, 23].

Flash congestion can be reduced if it can be detected and if network-layer alternatives can be exploited quickly enough. We address this challenge by constructing a *cross-layer network stack that quickly detects congestion at lower network layers, to drive upper layer routing decisions, to find alternative lower-congestion paths to destinations*.

In this paper, we present the implementation and experimental evaluation of *DeTail*. DeTail is a cross-layer network stack design to reduce long-tailed flow completions in datacenter networks. It provides an effective network foundation for enabling mixer applications to assemble their complex content more completely and within responsiveness time constraints. Our key contributions of this work are:

- Quantification of the impact of long-tailed flow completion times on different datacenter workflows;
- Assessment of the causes of long-tailed flow completion times;
- A cross-layer network stack that addresses them;
- Implementation-validated simulations demonstrating DeTail’s reduction of 99.9th percentile flow completion times by over 50% for a range of workloads without significantly increasing the median

In the following section, we analyze how long-tailed flow completion times affect workflows’ interactive deadlines. In Section 3, we describe the causes of long-tailed flow completion times and the inadequacy of partial solutions. In Section 4, we introduce the cross-layer network-based approach DeTail uses to overcome these issues. In Section 5, we describe the NS-3-based simulation [8] and Click-based implementation [26] with which we evaluate DeTail. We evaluate DeTail in Section 6, demonstrating reduced flow completion times for a wide range of workloads. We discuss various aspects of DeTail in Section 7. We describe how DeTail compares with prior work in Section 8 and conclude in Section 9.

2. IMPACT OF THE LONG TAIL

In this section, we begin by analyzing datacenter network traffic measurements, describing the phenomenon of the long tail. Next, we present two workflows commonly used by page creation subsystems and quantify the impact of the long flow completion time tail on their ability to provide rich, interactive content. We compare this with the performance that

could be achieved with shorter-tailed distributions. We conclude this section with a discussion of how to quantify the long tail.

2.1 Traffic Measurements

Recently, Microsoft researchers [13] published datacenter traffic measurements for production networks performing services like web search. These traces captured three traffic types: (i) soft real-time queries, (ii) urgent short messages, and (iii) large deadline-insensitive background updates. Figure 1 reproduces graphs from [13], showing the distribution of measured round-trip-times (RTTs) from worker nodes to aggregators. The former typically communicate with mid-level aggregators (MLAs) located on the same rack. This graph represents the distribution of intra-rack RTTs.

Figure 1 shows that while the measured intra-rack RTTs are typically low, congestion causes them to vary by two orders of magnitude, forming a long-tail distribution. In this particular environment, intra-rack RTTs take as little as $61\mu s$ and have a median duration of $334\mu s$. But, in 10% of the cases, RTTs take over 1ms. In fact, RTTs can be as high as 14ms. Since these RTTs are the measured time between the transmission of a TCP packet and the receipt of its acknowledgment, *the variation in RTTs is caused primarily by congestion*.

For comparison, Figure 1 includes a synthetic distribution of RTTs following a normal distribution. While we set this distribution to have a median value that is 50% higher than that of the measured one, it has a much shorter tail.

As a measured distribution of datacenter flow completion times is unavailable, we conservatively assume that each flow takes one RTT.

2.2 Impact on Workflows

Here we introduce the partition-aggregate and sequential workflows commonly used by page creation subsystems. For both workflows, we compare the impact of the long-tailed measured distribution with a shorter-tailed one. For this comparison, we focus on 99.9th percentile performance as this is the common metric used for page creation [13, 33]. We see that a long-tailed distribution performs significantly worse than a shorter-tailed distribution, even when the latter has a higher median. We conclude this analysis with the key takeaways.

2.2.1 Partition-Aggregate

Partition-aggregate workflows are used by subsystems such as web search. Top-level aggregators (TLAs) receive requests. They divide (partition) the computation required to perform the request across multiple mid-level aggregators (MLAs), who further partition computation across worker nodes. Worker nodes perform the computation in parallel and send their results back to their MLA. Each MLA combines the results it receives and forwards them on to the TLA.

To ensure that the response is provided in a timely manner,

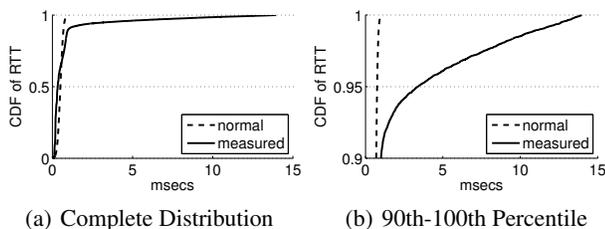


Figure 1: CDF of RTTs from the worker to the aggregator. We compare the Microsoft’s measured distribution [13] with a synthetic normal one having a 50% larger median.

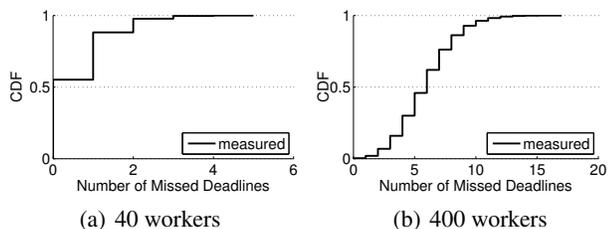


Figure 2: Probability that a workflow will have a certain number of workers miss their 10ms deadline. All workers would meet their deadlines if RTTs followed the normal distribution.

it is common practice to give worker nodes as little as 10ms to perform their computation and deliver their result [13]. If a worker node does not meet its deadline, its results are typically discarded, ultimately degrading the quality of the response.

To assess the impact of the measured RTT distribution on partition-aggregate workers meeting such deadlines, we analyze two hypothetical workflows. One has 40 workers while the other has 400. In Figure 2, we show the probability that a workflow will have a certain number of workers miss their deadlines. We assigned completion times to each worker by sampling from the measured RTT. Those with completion times greater than 10ms were considered to have missed their deadlines. We performed this calculation 10,000 times. Under the measured distribution, at the 99.9th percentile, a 40-worker workflow has 4 workers (10%) miss their deadlines, while a 400-worker workflow has 14 (3.50%) miss theirs. Had RTTs followed the normal distribution, no workers would have missed their deadlines. This is despite the distribution having a 50% higher median than the measured one. This shows the hazard of designing for the median rather than long-tail performance.

These results assume that worker nodes do not spend any time computing the result they transmit. As the pressure for workers to perform more computation increases, the fraction of workers missing their deadlines will increase as well.

2.2.2 Sequential

In sequential workflows, a single front-end server fetches data from back-end servers (datastores) for every page cre-

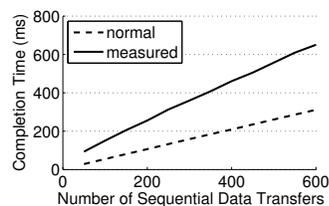


Figure 3: 99.9th percentile completion times of sequential workflows. A web site could use twice as many sequential request per page creation under a shorter-tailed distribution.

ation. Future requests depend on the results of previous ones.

To quantify the impact of the long tail, we generated sequential workflows with varying numbers of data retrievals. We assumed that each data retrieval would use one flow and obtained values for retrievals by sampling from the appropriate distribution in Figure 1. We took the completion time of sequential workflows to be the sum of the randomly generated data retrieval times. We performed this calculation 10,000 times.

In Figure 3, we report 99.9th percentile completion times for different RTT distributions. Under the measured RTT distribution, to meet 200ms page creation deadlines, web sites are limited to less than 150 sequential data retrievals per page creation. Had RTTs followed the normal distribution, web sites could employ more than 350 sequential data retrievals per page. This is despite the distribution having a 50% higher median than the measured one. Again, designing for the median rather than long-tail performance is a mistake.

2.2.3 Takeaways

Long-tailed RTT distributions make it challenging for workflows used by page creation subsystems to meet interactivity deadlines. *While events at the long tail occur rarely, workflows use so many flows, that it is likely that several will experience long delays for every page creation.* Hitting the long tail is so significant that workflows actually perform better under distributions that have higher medians but shorter tails.

The impact is likely to be even greater than that presented here. Our analysis did not capture packet losses and retransmissions that are likely to cause more flows to hit the long tail.

Facebook engineers tell us that the long tail of flow completions forces their applications to choose between two poor options. They can set tight data retrieval timeouts for retrying requests. While this increases the likelihood that they will render complete pages, long tail flows generate spurious requests that increase server load. Alternatively, they can use conservative timeouts that avoid spurious requests, but limit complete web page rendering by waiting too long for retrievals that never arrive. *A network that reduces the flow completion time tail allows such applications to use tighter timeouts to render more complete pages without increasing*

server load.

2.3 Quantifying the Tail

Median flow completion time is an insufficient indicator of workflow performance. However, determining the right metric is challenging. Workflows only requiring 10 flows are much less likely to be affected by 99.9th percentile flow completion times versus those with 1000 flows. To capture the effect of the long tail on a range of different workflow sizes, we report both 99th and 99.9th percentile flow completion times.

3. CAUSES OF LONG TAILS

Section 2 showed how the long tail of flow completion times impacts page creation workflows. As mentioned earlier, flash congestion aggravates three problems that lead to long-tailed flow completion times: packet losses and retransmissions, absence of prioritization, and uneven load balancing. In this section, we describe these problems. We conclude by discussing why current solutions fall short.

3.1 Packet Losses and Retransmissions

[13, 15, 31] study the effect of packet losses and retransmissions on network performance in datacenters. Packet losses often lead to flow timeouts, particularly in short flows where window sizes are not large enough to perform fast recovery. In datacenters, these timeouts are typically set to 10 ms [13, 31]. Since datacenter RTTs are commonly of the order of $250\mu s$, just one timeout guarantees that the short flow will hit the long tail. It will complete too late, making it unusable for page creation. Using shorter timeouts may mitigate this problem, but it increases the likelihood of spurious retransmissions that increase network and server load.

Additionally, partition-aggregate workflows make incast more likely [13, 33]. Workers performing computation typically respond simultaneously to the same aggregator, sending it short flows. This sometimes leads to correlated losses that cause many flows to timeout and hit the long tail.

3.2 Absence of Prioritization

Datacenter networks represent a shared environment where many flows have different sizes and timeliness requirements. The traces from Section 2 show us that datacenters must support both latency-sensitive and latency-insensitive flows, with sizes that typically range from 2KB to 100MB [13].

During periods of flash congestion, short latency-sensitive flows can become enqueued behind long latency-insensitive flows. This increases the likelihood that latency-sensitive flows will hit the long tail and miss their deadlines. Approaches that do not consider different flow requirements can harm latency-sensitive flows.

3.3 Uneven Load Balancing

Modern datacenter networks have scaled out, creating many paths between every source and destination [11, 22, 23]. Flow

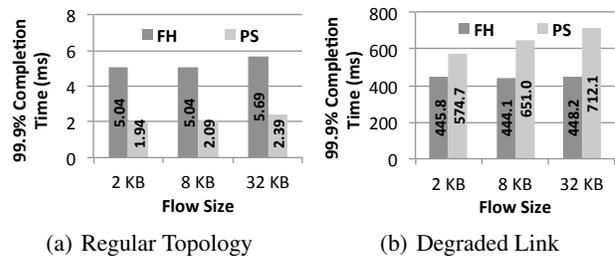


Figure 4: Simulated 99th percentile flow completion times of flow hashing (FH) and packet scatter (PS) in a bursty workload.

hashing is typically used to spread load across these paths while maintaining the single-path assumption commonly employed by transport protocols. Imperfect hashing, as well as varying flow sizes often lead to uneven flow assignments. These flows are unnecessarily assigned to a more congested path, despite the availability of less congested ones. This increases the likelihood that they will hit the long tail.

This phenomena has been observed before for large flow sizes [12, 29]. Here we show it is also a problem for the short flows common in page creation. We present a simulation on a 128-server FatTree topology with a moderate oversubscription factor of four (two from top-of-rack to aggregate switches and two from aggregate to core switches). For this experiment, we ran an all-to-all workload consisting solely of high-priority, uniformly chosen 2KB, 8KB, and 32KB flows. These sizes span the range of latency-sensitive flows common in datacenter networks [13].

In Figure 4(a), we compare the performance of flow hashing and a simple multipath approach; *packet scatter*. Packet scatter randomly picks the output port on which to send packets when multiple shortest paths are available. To factor out transport-layer effects, we used infinitely large switch buffers and also disabled rate-limiting and packet retransmission mechanisms. We see that packet scatter significantly outperforms traditional flow hashing, cutting the 99.9th percentile flow completion time in half. As we have removed transport-layer effects, these results show that *single path approaches reliant of flow hashing significantly underperform multipath ones*.

Multipath approaches that do not dynamically respond to congestion, like packet scatter, may perform significantly worse than flow hashing for topological asymmetries. Consider a common type of failure, where a 1 Gbps link between a core and aggregate switch has been degraded and now operates at 100 Mbps [29]. Figure 4(b) shows that for the same workload, packet scatter can perform over 50% worse than flow hashing. As we will see in Section 6, flow hashing itself performs poorly.

Topological asymmetries occur for a variety of reasons. Datacenter network failures are common [17]. Asymmetries can be caused by incremental deployments or network re-configurations. Both static approaches (packet scatter and flow hashing) are unaware of the different capabilities of different paths and cannot adjust to these environments. An

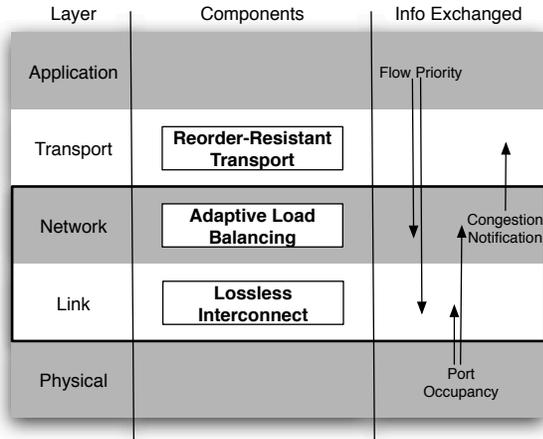


Figure 5: The DeTail network stack uses cross-layer information to address the sources of long tail in flow completion times.

adaptive multipath approach would be able to manage such asymmetries.

3.4 Current Solutions Insufficient

DCTCP and D^3 [13, 33] are two single path solutions recently proposed to reduce the completion times of latency-sensitive flows. The datacenter bridging effort [3] has proposed fairness and congestion control protocols. These reduce packet losses and prioritize latency-sensitive flows. But they do not address the uneven load balancing caused by flow hashing, and hence suffer the performance loss illustrated in Figure 4(a).

Recently two solutions have been proposed to more evenly balance flows across multiple paths. Hedera monitors link state and periodically remaps flows to alleviate hotspots [12]. Since Hedera remaps flows every five seconds and focuses on flows taking more than 10% of link capacity, it cannot improve performance for the short flows common in page creation.

The other is MPTCP [29]. MPTCP launches multiple TCP subflows and balances traffic across them based on congestion. MPTCP uses standard TCP congestion detection mechanisms that have been shown by DCTCP to be insufficient for preventing packet drops and retransmissions [13]. Also, while MPTCP is effective for flow sizes larger than 70KB, it is worse than TCP for flows with less than 10 packets [29]. As small flows typically complete in just a few RTTs, host-based solutions do not have sufficient time to react to congested links and rebalance their load. *Current multipath-aware solutions cannot support the short flows common in page creation workflows.*

4. DETAIL

In this section, we first provide an overview of DeTail’s functionality and discuss how it addresses the causes of long-tailed flow completion times. We then describe the mecha-

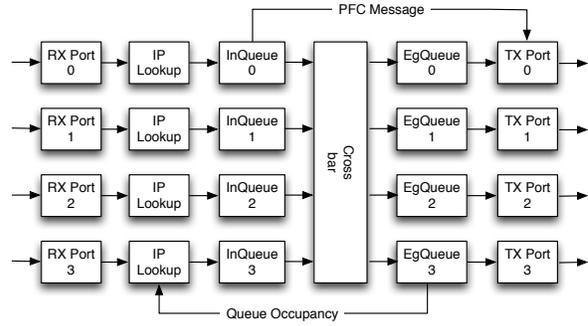


Figure 6: Assumed CIOQ Switch Architecture

nisms DeTail uses to achieve this functionality and their parameterization.

4.1 Overview

DeTail is a cross-layer network-based approach for reducing the long flow completion time tail. The major components of the DeTail stack and the cross-layer information exchanged is depicted in Figure 5.

At the link layer, DeTail uses port buffer occupancies to construct a *lossless interconnect*. By responding quickly, lossless interconnects ensure that packets are never dropped due to flash congestion. They are only dropped due to hardware errors and/or failures. Preventing congestion-related losses reduces the number of flows that experience long completion times.

At the network layer, DeTail performs per-packet adaptive load balancing of packet routes. At every hop, switches use the congestion information obtained from port buffer occupancies to dynamically pick a packet’s next hop. This approach evenly smooths network load across available paths, reducing the likelihood of encountering a congested portion of the network. Since it is adaptive, it performs well even given topologic asymmetries.

DeTail’s choices at the link and network layers have implications for transport. Since packets are no longer lost due to congestion, our transport protocol relies upon congestion notifications derived from port buffer occupancies. Since routes are load balanced one packet at a time, out-of-order packet delivery cannot be used as an early indication of congestion to the transport layer.

Finally, DeTail allows applications to specify flow priorities. Applications typically know which flows are latency-sensitive foreground flows and which are latency-insensitive background flows. By allowing applications to set these priorities, and responding to them at the link and network layers, DeTail ensures that high-priority packets do not get stuck behind low-priority ones. This assumes that applications are trusted, capable of specifying which flows are high priority. We believe that this assumption is appropriate for the kind of environment targeted by DeTail.

4.2 DeTail’s Details

Now we discuss the detailed mechanisms DeTail uses to realize the functionality presented earlier. We begin by describing our assumed switch architecture. Then we go up the stack, discussing what DeTail does at every layer. We conclude with a feasibility analysis for datacenter networks.

4.2.1 Assumed Switch Architecture

In Figure 6, we depict a simplified four-port representation of a Combined Input/Output Queue (CIOQ) Switch. The CIOQ architecture is commonly used in today’s switches [2,28]. While we discuss DeTail’s mechanisms in the context of this architecture, they should be implementable on others as well. This architecture employs both ingress and egress queues, which we denote as InQueue and EgQueue, respectively. A crossbar moves packets between these queues.

When a packet arrives at an input port (e.g., RX Port 0), it is passed to the forwarding engine (IP Lookup). The forwarding engine determines on which output port (e.g., TX Port 2) the packet should be sent. Once the output port has been determined, the packet is stored in the ingress queue (i.e., InQueue 0) until the crossbar becomes available. When this happens, the packet is passed from the ingress queue to the egress queue corresponding to the desired output port (i.e., InQueue 0 to EgQueue 2). Finally, when the packet reaches the head of the egress queue, it is transmitted on the corresponding output port (i.e., TX Port 2).

To ensure that high-priority packets do not wait behind those with low-priority, the switch’s ingress and egress queues perform strict priority queueing. Switches are typically capable of performing strict priority queueing between eight different priorities [6]. We use strict prioritization at both ingress and egress queues.

DeTail requires that the switch provide per-priority ingress and egress queue occupancies to higher layers in the stack. Each queue maintains a *drain bytes* counter per priority. This is the number of bytes of equal or higher priority in front of a newly arriving packet. The switch maintains this value by incrementing/decrementing the counters for each arriving/departing packet.

Having higher layers continuously poll the counter values of each queue may be prohibitively expensive. To address this issue, the switch associates a signal with each counter. Whenever the value of the counter is below a pre-defined threshold, the switch asserts the associated signal. These signals enable higher layers to quickly select queues without having to obtain the counter values from each. When multiple thresholds are used, a signal per threshold is associated with each counter. We describe how these thresholds are set in Section 4.3.2.

4.2.2 Link Layer

At the link layer, DeTail employs flow control to create a lossless interconnect. While many variants of flow control exist [10], we chose to use the one that is becoming part of the Ethernet standard: Priority Flow Control (PFC) [9].

PFC has already been adopted by vendors and is available on newer Ethernet switches [6].

The switch monitors ingress queue occupancy to detect congestion. When the drain byte counters of an ingress queue pass a threshold, the switch reacts by sending a Pause message informing the previous hop that it should stop transmitting packets with the specified priorities. When the drain byte counters reduce, it sends an Unpause message to the previous hop asking it to resume transmission of packets with the selected priorities¹.

During periods of persistent congestion, the buffers at the previous hop fill, forcing it to generate its own Pause message. In this way, flow control messages can propagate back to the source, quenching it.

We chose to generate Pause/Unpause messages based on ingress queue occupancies because packets stored in these queues are attributed to the port on which they arrived. By sending Pause messages to the corresponding port when an ingress queue fills, DeTail ensures that the correct source postpones transmission.

Our choice of using PFC is based on the fact that packets in lossless interconnects can experience head-of-line blocking. With traditional flow control mechanisms, when the previous hop receives a Pause message, it must stop transmitting all packets on the link, not just those contributing to congestion. As a result, packets at the previous hop that are not contributing to congestion may be unnecessarily delayed. By allowing eight different priorities to be paused individually, PFC reduces the likelihood that low-priority packets will delay high priority ones. We describe how packet priorities are set in Section 4.2.5.

4.2.3 Network Layer

At the network layer, DeTail makes congestion-based load balancing decisions. Since datacenter networks have many paths between the source and destination, multiple shortest path options exist. When a packet arrives at a switch, it is forwarded on to the shortest path that is least congested.

DeTail monitors the egress queue occupancies described in Section 4.2.1 to make congestion-based decisions. Unlike traditional Ethernet, egress queue occupancies provide an indication of the congestion being experienced downstream. As congestion increases, flow control messages are propagated towards the source, causing the queues at each of the switches in the path to fill. By reacting to local egress queue occupancies we make globally-aware hop-by-hop decisions without additional control messages.

We would like to react by picking an acceptable port with the smallest drain byte counter at its egress queue for every packet. However, with the large number of ports in today’s switches, the computational cost of doing so is prohibitively high. We leverage the threshold-based signals described in Section 4.2.1. By concatenating all the signals for a given

¹PFC messages specify the duration for which packet transmissions should be delayed. We use them here in an on/off fashion.

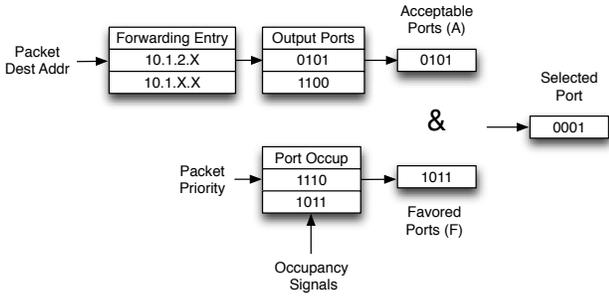


Figure 7: Performing Adaptive Load Balancing - A packet’s destination IP address is used to determine the bitmap of Acceptable Ports (A). Additionally, the packet’s priority and port buffer occupancy signals are used to find the bitmap of the lightly loaded Favored Ports (F). A bitwise AND (&) of these two bitmaps gives the set of Selected Ports from which one is chosen.

priority, we obtain a bitmap of the favored ports (F), which are lightly loaded.

DeTail relies on forwarding engines to obtain the set of available shortest paths to a destination. We assume that associated with each forwarding entry is a bitmap of acceptable ports (A) that lead to shortest paths for matching packets².

As shown in Figure 7, when a packet arrives, DeTail sends its destination IP address to the forwarding engine to determine which entry it belongs to and obtains the associated bitmap of acceptable ports (A). It then performs a bitwise AND (&) of this bitmap (A) and the set of favored ports (F) matching the packet’s priority, to obtain the set of lightly loaded ports that the packet can use. DeTail randomly chooses from one of these remaining ports and forwards the packet³.

During periods of high congestion, the set of favored ports may be empty. In this case, DeTail performs the same operation with a second, larger threshold. If that does not yield results either, DeTail randomly picks a port from the bitmap. We describe how to set these thresholds in Section 4.3.2.

4.2.4 Transport Layer

A transport-layer protocol must address two issues to run on our load-balanced, lossless interconnect. It must be resistant to packet reordering and it must react to a form of congestion notification other than packet loss.

Our lossless interconnect simplifies developing a transport protocol that is robust to out-of-order packet delivery. The lossless interconnect ensures that packets will only be lost due to relatively infrequent hardware errors/failures. As packet drops are now much less frequent, it is not necessary that the transport protocol respond agilely to them. We simply need to disable the monitoring and reaction to out-of-order packet delivery. For TCP NewReno, we do this by

²Bitmaps can be obtained with the TCAM and RAM approach described by [11].

³In cases where obtaining a random number is too costly, round robin can be used instead

disabling fast recovery and fast retransmit. While this leads to increased buffering at the end host, this is an acceptable tradeoff given the large amount of memory available on modern servers.

Obtaining congestion information from a lossless interconnect is more difficult. Traditionally, transport protocols monitor packet drops to determine congestion information. As packet drops no longer happen due to congestion, another approach is necessary. To enable TCP NewReno to operate effectively with DeTail, we task each source node with monitoring its egress buffer occupancy. When its drain byte counters rise above a threshold, it reacts by setting outgoing packets’ ECN flags, forcing TCP to reduce its rate. By placing this functionality at the source, we ensure that the transport layer reacts only to persistent congestion.

4.2.5 Application Layer

DeTail depends upon applications to properly specify flow priorities based on how latency-sensitive they are. Applications express these priorities to DeTail through the sockets interface. They set each flow (and hence the packets belonging to it) to have one of eight different priorities. As the priorities are relative, applications need not use all of them. In our evaluation, we only use two.

Applications must also react to extreme congestion events where the source has been quenched for a long time (Section 4.2.2). They need to determine how to reduce network load while minimally impacting the user.

4.2.6 Feasibility Analysis

DeTail depends on the ability of datacenter networks to use non-commodity switches. This assumption is appropriate for modern-day datacenters because operators are willing to influence and adopt new designs [33].

Building custom switches is less of a hurdle than previously. Switches with enhanced functionality can be built from reasonably priced commodity ASICs. A PCI-E board with a 4-port switch ASIC costs as little as \$400 [27]. Furthermore, ASIC manufacturers are showing a willingness to adopt new protocols. For example, the FM 6000 ASIC already incorporates the recently standardized datacenter bridging protocols [6].

We recognize that it is advantageous to make as few changes to switches as possible to minimize the cost per port. Our choice of using PFC to perform link-layer flow control and our simple ALB scheme both reflect this consideration.

4.3 Choice of Settings

Now that we have described the mechanisms employed by DeTail, we provide analysis of how to choose their parameters. We also assess how end-host parameters should be chosen when running DeTail.

4.3.1 Link Layer Flow Control

A key parameter is the threshold for triggering PFC messages. Pausing a link early allows congestion information to

be propagated more quickly, making DeTail’s adaptive load balancing more agile. At the same time, it increases the number of control messages. As PFC messages take time to be sent and responded to, setting the Unpause threshold too low can lead to buffer underflow, reducing link utilization.

To strike a balance between these competing concerns, we must first calculate the time to generate PFC messages. We use the same approach described in [9] to obtain this value.

For GigE, it may take up to $36.456 \mu s$ for a PFC message to take effect⁴. 4,557 bytes may arrive after a switch generates a PFC message. As we pause every priority individually, this can happen for all eight priorities. We must leave $4,557B \times 8 = 36,456B$ of buffer space for receiving packets after PFC generation. Assuming 128 KB buffers, this implies a maximum Pause threshold of $(131,072B - 36,456B)/8 = 11,827$ drain bytes per priority. Setting the threshold any higher can lead to packet loss.

Calculating the Unpause threshold is more challenging because the specifics of congestion causes queues to drain at different rates. In our calculations, we assume a drain rate of 1 Gbps. To ensure that the ingress queues do not underflow, our Unpause threshold must be at least 4,557B. In certain situations, ingress queues may drain faster or slower than 1 Gbps. If they drain slower, additional control messages may have to be sent, re-pausing the priority. If they drain faster, our egress queues reduce the likelihood of link underutilization.

These calculations establish the minimum and maximum threshold values to prevent packet loss and buffer underflow. Between the desire for agility and reduced control message overhead, we set the Unpause threshold to the minimum value of 4,557 drain bytes and the Pause threshold to 8,192 drain bytes (halfway between the minimum and the maximum). When fewer priorities are used, the Pause threshold can be raised without suffering packet loss. Given the desire to agilely respond to congestion, we leave it unmodified.

4.3.2 Adaptive Load Balancing

When performing threshold-based adaptive-load balancing, we must determine how many thresholds to have for a given priority (i.e., most favored, favored, and least favored ports) as well as what these thresholds should be. Clearly, increasing the number of thresholds increases complexity, so the benefits of each additional threshold must outweigh the complexity cost.

Through a simulation-based exploration of the design space with the other parameters as described above, we determined that having two thresholds of 16 KB and 64 KB yields favorable results.

⁴We do not consider jumbo frames. Also, PFC is only defined for 10 GigE. We use GigE in this paper for manageable simulation times. We base PFC response times on the time specified for Pause Frames. This is appropriate because 10 GigE links are given the same amount of time to respond to PFC messages as they are to Pause Frames.

4.3.3 Explicit Congestion Notification

The threshold for setting ECN flags represents a trade-off. Setting it too low reduces host queue occupancy but increases the chance high-priority flows will be forced to reduce their rate. Setting it too high has the opposite effect. Through experiments, we determined that a threshold of 64KB drain bytes appropriately makes this tradeoff.

4.3.4 End-Host Timers

Setting end host timers too low leads to spurious retransmissions that waste network resources. Setting them too high leads to long response-times when packets are dropped.

Traditionally, transport-layer protocols recover from packet drops caused by congestion and hardware failures. Congestion occurs frequently, so responding quickly to packet drops is important for achieving high throughput. However, DeTail ensures that packet drops only occur due to relatively infrequent hardware errors/failures. Therefore, it is more important for the timeouts to be larger to avoid spurious retransmissions.

To determine a robust timeout value for DeTail, we simulated all-to-all incast 25 times with varying numbers of servers and timeout values. During every incast, one server receives a total of 1 MB from the remaining servers. We saw that timeouts 10 ms and larger effectively avoid spurious retransmissions.

In this simulation, there is only one switch between all of the servers. For the experiments with DeTail in Section 6, we use a timeout of 200 ms to reflect that datacenters topologies typically have multiple hops.

5. EXPERIMENTAL SETUP

Here we describe the NS-3 based simulator [8] and Click-based implementation [26] we use to evaluate DeTail.

5.1 NS-3 Simulation

Our NS-3 based simulation closely follows the switch design depicted in Figure 6. Datacenter switches typically have 128 - 256 KB buffers per port [13]. To meet this constraint, we chose per-port ingress and egress queues of 128 KB.

Network simulators typically assume that nodes are infinitely fast at processing packets, this is inadequate for evaluating DeTail. We extended NS-3 to include real-world processing delays. Switch delays of $25 \mu s$ are common in datacenter networks [13]. We rely upon published specifications to break-down this delay as follows, providing explanations where possible:

- $12.24 \mu s$ transmission delay of a full-size 1530B Ethernet frame on a 1 GigE link.
- $3.06 \mu s$ crossbar delay when using a speedup of 4. Crossbar speedups of 4 are commonly used to reduce head of line blocking [28].
- $0.476 \mu s$ propagation delay on a copper link [9].

- $5\mu s$ transceiver delay (both ends of the link) [9].
- $4.224\mu s$ forwarding engine delay (the remainder of the $25\mu s$ budget).

We incorporate the transceiver delay into the propagation delay. The other delays are implemented individually, including the response time to PFC messages.

Packet-level simulators are known to have scalability issues, in terms of topology size and simulation duration [29]. We evaluated the feasibility of also developing a flow-level simulator, but concluded that it would be unable to shed light on the packet-level dynamics that are the focus of this paper.

5.2 Click-based Implementation

To validate our approach, we implemented DeTail in Click [26]. In general, our implementation mirrors the design decisions specified in Section 4 and portrayed in Figure 6. Here we describe the salient differences and analyze the impact they have on our parameters.

5.2.1 Design Differences

Unlike hardware switches, software routers typically do not emulate a CIOQ switch architecture. Instead, the forwarding engine places packets directly into the output queue. This output-queued approach is poorly suited to DeTail because we rely on ingress queues to determine when to send PFC messages.

To address this difference, we modified Click to have both ingress and egress queues. When packets arrive, the forwarding engine simply annotates them with the desired output port and places them in the ingress queue corresponding to the port on which they arrived. Crossbar elements then pull packets from the ingress queue to the appropriate egress queue. Finally, when the output port become free, it pulls packets from its egress queue.

Software routers also typically do not have direct control over the underlying hardware. For example, when Click *sends* a packet, it is actually enqueued in the driver’s ring buffer. The packet is then DMAed to the NIC where it waits in another buffer until it can be placed on the wire. In Linux, the driver’s ring buffer alone can contain hundreds of packets. It is difficult for the software router to assess how congested the output link is when performing load balancing. Also, hundreds of packets may be transmitted between the time when the software router receives a PFC message and it takes effect.

To address this issue, we add a rate limiter in Click before every output port. They clock out packets based on the bandwidth of the link. They effectively reduce packet buildup in the driver’s and NIC’s buffers, instead keeping those packets in Click’s queues for a longer duration.

5.2.2 Parameter Modifications

The limitations of our software router impact our parameter choices. It does not have hardware support for PFC mes-

sages. Consequently it takes more time for these messages to be generated and responded to.

Also, our rate limiter allows batching up to 6 KB of data to ensure efficient DMA use. This may cause PFC messages to be enqueued for longer before they are placed on the wire and additional data may be transmitted before a PFC message takes effect. This also hurts high-priority packets. High priority packets will suffer additional delays if they arrive just after a batch of low priority packets has been passed to the driver.

To address these limitations, we increased our Pause / Unpause thresholds. However, instead of increasing ingress queue sizes, we opted to ensure that only two priorities were used at a time. This approach allows us to provide a better assessment of the advantages of DeTail in datacenter networks.

6. EXPERIMENTAL RESULTS

In this section, we evaluate DeTail through extensive simulation and implementation, demonstrating its ability to reduce the flow completion time tail for a wide range of workloads. We begin with an overview describing our traffic workloads and touching on key results. Next, we compare simulation and implementation results, validating our simulator. Later, we subject DeTail to a wide range of workloads under a larger topology than permitted by the implementation and investigate its scaled-up performance.

6.1 Overview

To evaluate DeTail’s ability to reduce the flow completion time tail, we focus on the following scenarios:

Flow Hashing (FH): Switches employ flow-level hashing.

This is the status quo and is our baseline for comparing the performance of DeTail.

Lossless Packet Scatter (LPS): Switches employ packet scatter (as already explained in Section 3) along with Priority Flow Control (PFC). While not being an industry standard, *LPS* is a naive multipath approach that can be deployed in current datacenters, as both packet scatter and PFC exist in the latest Ethernet switches. We use *LPS* to highlight the advantage of Adaptive Load Balancing (ALB) employed by DeTail over existing mechanisms.

DeTail: As already explained in previous sections, switches employ PFC and ALB.

All three cases use strict priority queueing and use TCP NewReno as the transport layer protocol. For *FH*, we use a TCP timeout of 10 ms, as suggested by prior work [13, 31]. Since, *LPS* and DeTail use PFC to avoid packet losses, we use the standard TCP timeout of 200 ms (as discussed in Section 4.3.4). Also, we use reorder buffers at the end-hosts to deal with out-of-order packet delivery.

We evaluate DeTail against *LPS* only in Section 6.4. For all other workloads, *LPS* shows similar improvements as DeTail and has been omitted for space constraints.

In NS-3, we were unable to simulate explicit congestion notification (as discussed in Section 4.2.4) employed by DeTail due to the lack of ECN support in the TCP model of NS-3. However, we implemented it and our results show that it minimally affects high-priority short flows.

Traffic Model: Our traffic model consists primarily of high-priority data retrievals. For each retrieval, a server sends a 10-byte request to and obtains a variable sized response (i.e., data) from another server. The size of the data (henceforth referred to as *retrieval data size*) is randomly chosen to be 2 KB, 8 KB, or 32 KB, with equal probability. We chose discrete data sizes for more effective analysis of 99th and 99.9th percentile performance. The rate of generation of these data retrievals (henceforth called *data retrieval rate*) is defined by the traffic workload. Where specified, we also run low-priority background data transfers.

Key results: Throughout our evaluation, we focus on 99th and 99.9th percentile completion times of data retrievals to assess DeTail’s effectiveness. We use the percentage reduction in the completion times provided by DeTail over Flow Hashing as the metric of improvement. Our key results are:

- DeTail completely avoids congestion-related losses, reducing 99.9th percentile completion times of data retrievals in bursty all-to-all workloads by up to 83% over Flow Hashing.
- DeTail effectively moves packets away from congestion hotspots that may arise due to disconnected links, reducing 99.9th percentile completion times by up to 94% over Flow Hashing. *LPS* can perform worse than *FH* in this environment.
- Reductions in individual data retrievals translate into improvements for partition-aggregate and sequential workflows, reducing their 99.9th percentile completion times by 31% and 59%, respectively.

6.2 Simulator Verification

To validate our simulator, we ran our Click-based implementation on Emulab [5]. We constructed a 36-node, 16-server FatTree topology. Oversubscription is common in datacenter networks [4]. To model the effect of a moderate oversubscription factor of four, we rate-limited the ToR-to-aggregate links to 500 Mbps and the aggregate-to-core links to 250 Mbps.

We designated half of the servers to be front-end (web-facing) servers and half to be back-end servers. Each front-end server periodically (according to Poisson distribution) issues a high-priority data retrieval to a randomly chosen back-end server. Data retrievals are generated at different steady rates of 100 to 1500 retrievals/second.

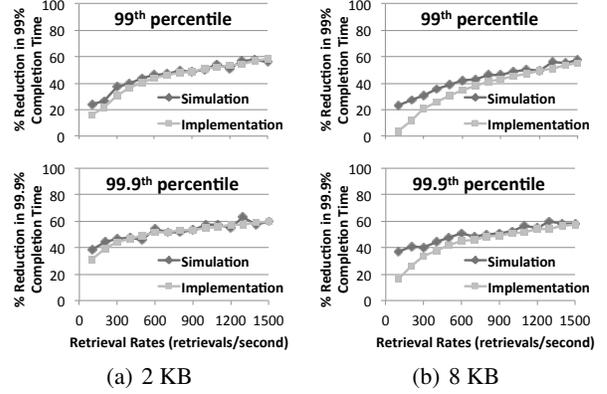


Figure 8: **Comparison of simulation and implementation results** - Reduction under DeTail over *FH* in 99% and 99.9% completion times of 2 KB and 8 KB data retrievals

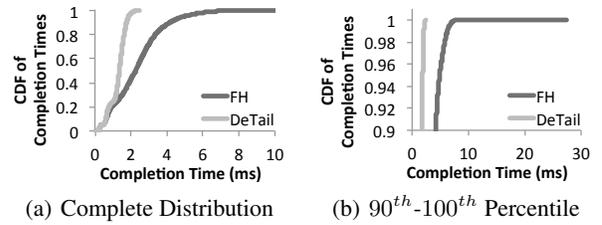


Figure 9: CDF of completion times of 8 KB data retrievals under a steady workload of 2000 retrievals/second.

We simulated the same workload and topology, with parameters matched with that of the implementation. Figure 8 compares the simulation results with the implementation measurements. For rates ranging from 500 to 1500 retrievals/sec, the reduction in completion time predicted by the simulator is closely matched by implementation measurements, with the difference in the percentage being within 8% (results for 32 KB data retrievals and *LPS* have been omitted for space constraints). Note that this difference increases for lower rates. We hypothesize that this is due to end-host processing delays that are present only in the implementation (i.e., not captured by simulation) dominating completion times during light traffic loads.

This demonstrates that our simulator is a good predictor of performance that one may expect in a real implementation. Next, we use this simulator to evaluate larger topologies and a wider range of workloads.

6.3 Microbenchmarks

We evaluate the performance of DeTail on a larger Fat-Tree topology with 128 servers. The servers are distributed into four pods having four ToR switches and four aggregate switches each. The four pods are connected to eight core switches. This gives a over-subscription factor of four in the network (two from top-of-rack to aggregate switches and two from aggregate to core switches). We subject this network to all-to-all data retrieval workloads. Each server

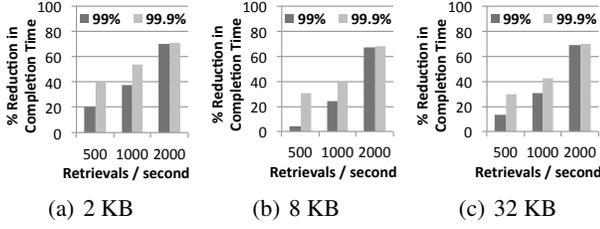


Figure 10: **Steady Workload** - Reductions provided by DeTail over *FH* in 99% and 99.9% completion times of 2 KB, 8 KB and 32 KB data retrievals

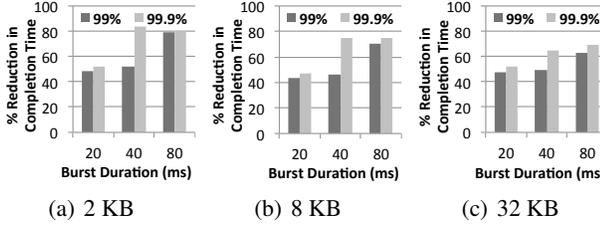


Figure 11: **Bursty Workload** - Reductions provided by DeTail over *FH* in 99% and 99.9% completion times of 2 KB, 8 KB and 32 KB data retrievals

periodically (according to a Poisson distribution) retrieves data from a randomly selected destination server. In addition, each server has on average, one 1 MB low-priority background flow. The rate of retrievals is varied to create either a steady or a bursty workload. Using a wide range of such workloads, we illustrate how ALB and PFC employed in DeTail reduces the tail of completion times as compared to *FH*.

Steady Workload: In this workload, all the servers retrieve data at a steady rate (similar to the workload used for verifying the simulator). This rate was varied from 500 to 2000 retrievals/second, which corresponded to load factors⁵ of approximately 0.17 to 0.67.

Figure 9 illustrates the effectiveness of DeTail in reducing the tail, by presenting the cumulative distribution of completion times of 8 KB data retrievals under a rate of 2000 retrievals/second. While the 99th and 99.9th percentile completion times under *FH* were 6.2 ms and 7.2 ms, respectively, DeTail reduced them to 2.0 ms and 2.3 ms; a reduction of about 67% in both cases. Even the median completion time improved by about 40% from 2.2 ms to 1.3 ms. Furthermore, the worst case completion time was 27 ms under *FH* compared to 2.5 ms, which reinforces the phenomenon discussed in Section 2. Flow completion times can increase by an order of magnitude due to congestion and mechanisms employed by DeTail are essential for ensuring tighter bounds on network performance.

Figure 10 presents the reductions in completion times for three data sizes at three retrieval rates. DeTail provided up to 70% reduction at the 99th percentile (71% at 99.9th percentile) completion times. Specifically, the 99.9th percentile

⁵load factor is the approximate utilization of the aggregate-to-core links by high-priority traffic only

completion times for all sizes were within 3.6 ms compared to 11.9 ms under *FH*. Due to the steady nature of the load, less than 0.1% of the flows suffered timeouts in *FH*. This indicates that PFC (which avoids timeouts) had less role to play, and packet-level load balancing by ALB provided most benefits. This further reinforces our arguments in Section 3 that any flow-hashing limits the effectiveness of single path approaches.

Within each data size, higher rates have greater improvement. This is because higher traffic load exacerbates the effect of uneven load balancing caused by *FH* giving a greater opportunity to ALB to balance the load.

Bursty Workload: Network traffic is traditionally characterized as bursty. Hence, we evaluated DeTail under a bursty workload, where the rate of data retrievals on all servers are periodically increased to generate bursts of traffic. In every 100 ms interval, servers generate all-to-all retrievals at 2500 retrievals/second for a specific *burst duration*. In the rest of the interval, retrievals are generated at low rate of 100 retrievals/second. The burst duration was varied from 20 ms to 80 ms, which corresponded to load factors of approximately 0.19 to 0.67.

Figure 11 shows that DeTail provided from 43% to 79% reduction at 99th percentile (47% to 83% at 99.9th percentile) completion times. The 99.9th percentile completion times for 8 KB data retrievals under all three burst durations were within 10 ms for DeTail compared to 41 ms for *FH*. Unlike the steady workload, more than 3% of the flows suffered timeouts under *FH*. Since these timeouts were eliminated under DeTail, a significant fraction of the improvement was due to PFC. As expected, higher burst durations (i.e., higher traffic loads) have greater chances of packet drops and timeouts under *FH* and therefore see greater improvements.

Long Background Flows: DeTail achieves these improvements for data retrievals (i.e., high-priority, short flows) without hurting the performance of long, low-priority background flows. To illustrate this, we evaluated the average completion time of such flows. Since these flows are deadline-insensitive, we focus on the average (which reflects throughput) rather than the 99.9th percentile. We used the 16-server implementation (described in the previous subsection) that performs ECN marking based on end-host buffer occupancies. We ran a steady workload where each server generates high-priority data retrievals at 300 retrievals/second. Alongside, each server is continuously engaged in a low-priority background flow with another randomly selected server, whose size is randomly chosen to be 1 MB, 16 MB or 64 MB with equal probability. Figure 14 shows that DeTail provided a 37% to 58% reduction over *FH* in the average completion time, which indicates that PFC and ALB employed in DeTail can even improve the performance of long flows. A detailed evaluation of DeTail’s impact on long flows is left for future work.

6.4 Topological Asymmetries

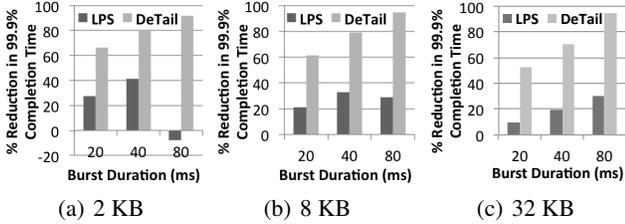


Figure 12: **Bursty Workload with Disconnected Link** - Reductions provided by *LPS* and DeTail over *FH* in 99.9% completion times of 2 KB, 8 KB and 32 KB data retrievals

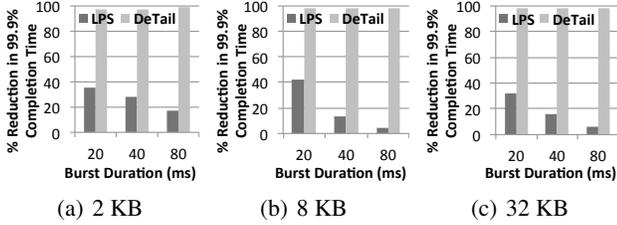


Figure 13: **Bursty Workload with Degraded Link** - Reductions provided by *LPS* and DeTail over *FH* in 99.9% completion times of 2 KB, 8 KB and 32 KB data retrievals

As discussed in Section 3.3, a multipath approach must be robust enough to handle topological asymmetries due to network component failures or reconfigurations. We consider two types of asymmetries: disconnected links and degraded links. These asymmetries lead to load imbalance, even with packet scatter. In this section, we present how ALB can adapt to the varying traffic demands and overcome the limitations of packet-level scattering. Besides *FH*, we evaluate DeTail against *LPS* to highlight the strength of ALB over packet scatter used in *LPS*. We assume that the routing protocol used in the network has detected the asymmetry and converged to provide stable multiple routes.

Disconnected Link: We re-evaluated the same bursty workload described in the previous section with the assumption of a disconnected link between a core switch and an aggregate switch. Figure 12 presents the reduction in 99.9th percentile completion for both *LPS* and DeTail (we do not present 99th percentile for space constraints). DeTail provided 52% to 94% reduction, an order of magnitude improvement (17 ms under DeTail compared to 210 ms under *FH* for 2 KB retrievals under 80 ms bursts). *LPS* is unable to match DeTail’s improvement, thus highlighting the effectiveness of ALB at evenly distributing load despite asymmetries in available paths. In fact, *LPS* can potentially lead to worse 99.9% completion times than *FH*, as is evident in case of 2 KB retrievals with 80 ms bursts (refer to Figure 12(a)).

Degraded Link: Instead of disconnecting, links can occasionally be downgraded from 1 Gbps to 100 Mbps. Figure 13

presents the results of the same bursty workload with a degraded core-to-aggregate link. DeTail provided more than 96% reduction compared to *FH*. This dramatic improvement is due to ALB’s inherent capability to route around congestion hotspots (i.e., switches connected to the degraded link) by redirecting traffic to alternate paths. While the 99.9th percentile completion time for 2 KB with 80 ms bursts (refer to Figure 13(a)) under *FH* and *LPS* was more than 2 seconds, it was 33 ms under DeTail.

In both cases of faults, the improvement in the tail comes at the cost of increased median completion times. As we have argued earlier, this trade off between the median and the 99.9% performance is appropriate for consistently meeting deadlines.

6.5 Web Workloads

Next, we evaluate how the improvements in individual data retrievals translate to improvements in the sequential and partition-aggregate workflows used in page creation workflows. Here we assign half the servers to be front-end servers and half to be back-end servers. The front-end servers initiate the workflows to retrieve data from randomly chosen back-end servers. We present the reduction in the 99.9th percentile completion times of these workflows.

Sequential Workflows: Each sequential workflow initiated by a front-end server consists of 10 data retrievals of size 4 KB, 6 KB, 8 KB, 10 KB, and 12 KB (randomly chosen with equal probability, giving an average of 8 KB [1]). As described in the Section 2, these retrievals need to be performed one after another. Similar to the bursty workload used earlier, every 100 ms, each front-end server generates 40 ms bursts at a rate of 800 workflows/second. In the rest of the interval, they generate workflows at 50 workflows/second. Figure 15 shows that DeTail provided 68% to 72% reduction in the 99.9th percentile completion time of individual data retrievals. In total, there was a 31% improvement in the 99.9th completion time of the sequential workflows, from 45 ms to 31 ms.

Partition-Aggregate Workflows: In each partition-aggregate workflow, a front-end server retrieves data in parallel from 10, 20, or 40 (randomly chosen with equal probability) back-end servers. As characterised in [13], the size of individual data retrievals is set to 2 KB. As before, we evaluate DeTail with a bursty workload; every 100 ms, bursts of 800 workflows/second for 20 ms is followed by 80 ms of 100 workflows/second. As shown in Figure 16, DeTail provided a 52% to 59% reduction in the 99.9th percentile completion time of individual data retrievals as well as workflows. Specifically, the 99.9th completion time of partition-aggregate workflow with 40 servers was 15 ms under DeTail, compared to 36 ms under *FH*.

These results demonstrate that DeTail effectively manages

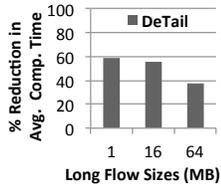


Figure 14: **Long Flows** Reductions in average completion time of long, low-priority flows

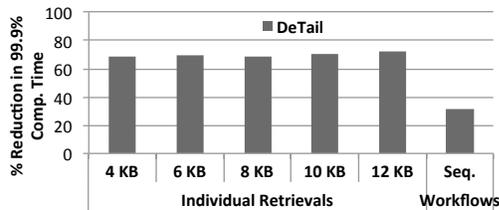


Figure 15: **Sequential Workflows** - Reductions provided by DeTail over *FH* in 99.9% completion time of sequential workflows and their individual data retrievals

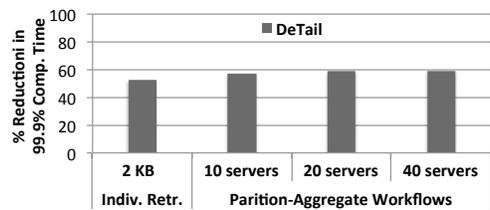


Figure 16: **Partition-Aggregate Workflows** - Reductions provided by DeTail over *FH* in 99.9% comp. time of partition-aggregate workflows and their individual retrievals

network congestion, providing significant improvements in the performance of distributed page creation workflows.

7. DISCUSSION

We first discuss how DeTail can be extended to handle switch errors. Next we present initial ideas on a DeTail-aware transport protocol. Finally, we discuss why the approach taken by DeTail will be advantageous as network bandwidth increases.

DeTail depends on flow control messages to be reliably delivered and processed by the recipient. While these control messages have the highest priority and will not be dropped or delayed, link errors may cause them to be corrupted. Or switch malfunctions may prevent them from being sent or processed in a timely manner.

We can employ a number of techniques to reduce the impact of these events. A switch that sends a flow control message can make sure that the recipient responds. If the recipient does not respond in the expected amount of time, the switch should resend the message. If the recipient continues to ignore flow control messages, the switch can take more extreme actions such as shutting down the link. Given DeTail’s adaptive load balancing and the availability of multiple paths in the datacenter, this should not significantly affect flow completion times.

Even with the additional robustness these mechanisms provide, packet drops may still occur. We rely upon the transport layer to handle these rare events.

The transport layer protocol presented in this paper is a retrofit of TCP NewReno. Delay-based protocols, such as TCP Vegas [14], may be better suited in these environments. Instead of waiting for packet drops that do not occur, they monitor increases in delay. Increased delay is precisely the behavior our lossless interconnect exhibits as congestion rises. We plan to investigate this approach further in the future.

Future increases in network bandwidth are unlikely to eliminate long-tailed flow completion times. Assuming offered load increases with link speeds, long tails will continue to be a challenge. While buffers will drain faster, they will also fill up more quickly, ultimately causing the packet losses and retransmissions that lead to long tails. Prioritization will continue to be important as background flows are likely to remain the dominant fraction of traffic. Finally, load imbalances due to topological asymmetries will continue to be a

problem.

8. RELATED WORK

In this section, we discuss prior work and how it relates to DeTail in three areas: Internet protocols, datacenters, and HPC interconnects, discussing each in turn.

8.1 Internet Protocols

The Internet was initially designed as a series of independent layers [16] with a focus on placing functionality at the end-hosts [30]. This approach explicitly sacrificed performance for generality. Improvements to this design, in terms of TCP modifications such as NewReno, Vegas, and SACK [14, 19, 24] and in terms of buffer management such as RED and Fair Queuing [18, 20] were proposed. All of these approaches focused on improving the notification and response of end-hosts. Consequently, they operate at coarse-grained timescales inappropriate for our workload.

DeTail differs from this work by taking a more agile in-network approach that breaks the single path assumption to reduce the flow completion time tail.

8.2 Datacenter Networks

Relevant datacenter work has focused on two areas: topologies and traffic management protocols. New topologies such as FatTrees, VL2, BCube, and DCell [11, 21–23] were proposed to increase bisection bandwidth. All of these approaches focused on increasing the number of paths between the source and destination because increasing link speeds was seen as impossible or prohibitively expensive.

Prior work has also focused on traffic management protocols for datacenters. DCTCP proposed mechanisms to improve flow completion time by reducing buffer occupancies [13]. D^3 sought to allocate flow resources based on application-specified deadlines [33]. And, the recent industrial effort known as Datacenter Bridging extends Ethernet to support traffic from other protocols that have different link layer assumptions [3]. All of these approaches focus on single-path mechanisms that are bound by the performance of flow hashing.

Datacenter protocols focused on spreading load across multiple paths have been proposed. Hedera performs periodic flow re-mapping of elephant flows [12]. MPTCP takes a step further, making TCP aware of multiple paths [29]. While these approaches provide multipath support, they operate at

timescales that are too coarse-grained to improve the short flow completion time tail.

8.3 HPC Interconnects

DeTail borrows some ideas from HPC interconnects. Credit-based flow control has been extensively studied and is often deployed to create lossless interconnects [10]. Adaptive load balancing algorithms such as UGAL and PAR have also been proposed [10]. To the best of our knowledge, these mechanisms have not been evaluated for web-facing datacenter networks focused on reducing the flow completion tail.

A commodity HPC interconnect, Infiniband, has made its way into datacenter networks [7]. While Infiniband provides a priority-aware lossless interconnect, it does not perform Adaptive Load Balancing (ALB). Without ALB, hotspots can occur, leading a subset of flows to hit the long tail. Host-based approaches to performing load-balancing, such as [32] have been proposed. But these approaches are limited because they are not sufficiently agile.

9. CONCLUSION

In this paper, we presented DeTail, a cross-layer networking stack designed to reduce the tail of flow completion times. DeTail is designed to reduce packet losses and retransmissions, prioritize latency-sensitive flows, and evenly balance traffic across multiple paths, making its flow completion statistics robust to flash congestion. By taking this approach, DeTail reduces 99.9th percentile flow completion times by over 50% for a wide range of workloads. This reduction enables web sites to deliver richer content while still meeting interactivity deadlines.

10. ACKNOWLEDGEMENTS

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