

Scalable Communication Endpoints for MPI+Threads Applications

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Abstract—Hybrid MPI+threads programming is gaining prominence as an alternative to the traditional “MPI everywhere” model to better handle the disproportionate increase in the number of cores compared with other on-node resources. Current implementations of these two models represent the two extreme cases of communication resource sharing in modern MPI implementations. In the MPI-everywhere model, each MPI process has a dedicated set of communication resources (also known as endpoints), which is ideal for performance but is resource wasteful. With MPI+threads, current MPI implementations share a single communication endpoint for all threads, which is ideal for resource usage but is hurtful for performance.

In this paper, we explore the tradeoff space between performance and communication resource usage in MPI+threads environments. We first demonstrate the two extreme cases—one where all threads share a single communication endpoint and another where each thread gets its own dedicated communication endpoint (similar to the MPI-everywhere model) and showcase the inefficiencies in both these cases. Next, we perform a thorough analysis of the different levels of resource sharing in the context of Mellanox InfiniBand. Using the lessons learned from this analysis, we design an improved resource-sharing model to produce *scalable communication endpoints* that can achieve the same performance as with dedicated communication resources per thread but using just a third of the resources.

Index Terms—multiple endpoints, hybrid MPI, multithreading, InfiniBand, scalable endpoints

I. INTRODUCTION

The Message-Passing Interface (MPI) is the most commonly used model for programming large-scale parallel systems today. The traditional model for using MPI hitherto has been the “MPI everywhere” model in which the application launches an MPI process on each core of the supercomputer and executes by ignoring the fact that some of the MPI processes reside on different cores of the same node while some execute on different nodes. The MPI implementation then internally optimizes communication within the node by using shared memory or other techniques.

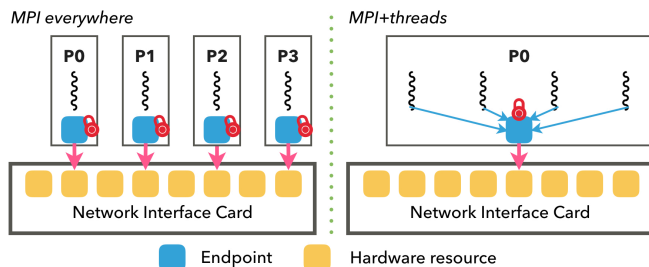


Fig. 1. Endpoint configuration of MPI-everywhere and MPI+threads models.

While the MPI-everywhere model of parallelism has served applications well for several decades, scaling applications in this model is becoming increasingly difficult. The biggest reason for this difficulty in scaling is that not all on-node resources scale at the same rate. Specifically, the number of cores available on a node is increasing rapidly. Other on-node resources such as memory, cache, TLB space, and network resources, however, scale much more slowly. Since the MPI-everywhere model uses a separate MPI process for each core, it inadvertently leads to a static split of all on-node resources, resulting in underutilization and wastage of resources. While optimizations such as *MPI shared memory* [13] address sharing a subset of resources (in particular, memory), these optimizations are not a generic solution for all on-node resources. Consequently, researchers have been increasingly looking at hybrid MPI+threads programming (e.g., MPI+OpenMP) as an alternative to the traditional MPI-everywhere model [16].

Current implementations of these two models—MPI everywhere and MPI+threads—represent the two extreme cases of communication resource sharing in modern MPI implementations. Figure 1 contrasts these two models in state-of-the-art MPI implementations, such as MPICH [7], that use one communication endpoint per MPI process [16]. A communication endpoint is a set of communication resources that allows the software to interface with the network hardware to send messages over the network.

In the MPI-everywhere model, multiple communication endpoints exist per node where each MPI process communicates using its own endpoint. This allows each MPI process to communicate completely independently of other processes, thus providing a direct and contention-free path to the network hardware and leading to the best-achievable communication performance (assuming that the MPI implementation is sufficiently optimized). In the MPI+threads model, on the other hand, all threads within an MPI process communicate using a single endpoint, which causes the MPI implementation to use locks on the endpoint for serialization. This model hurts communication throughput; more important, the available network-level parallelism remains underutilized. But, this model uses the least possible amount of communication resources.

A straightforward way to achieve maximum communication path independence between threads in the MPI+threads model is to dedicate a separate context each containing an endpoint with its own set of resources to each thread. This emulates the

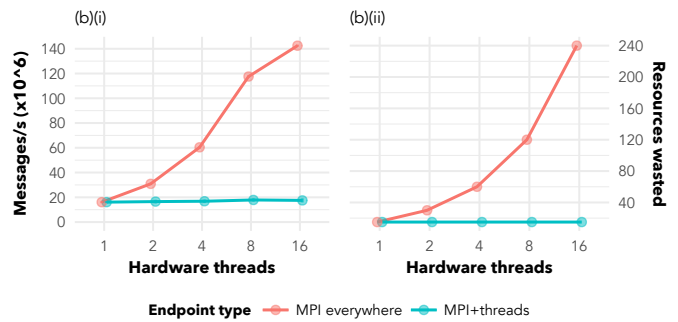
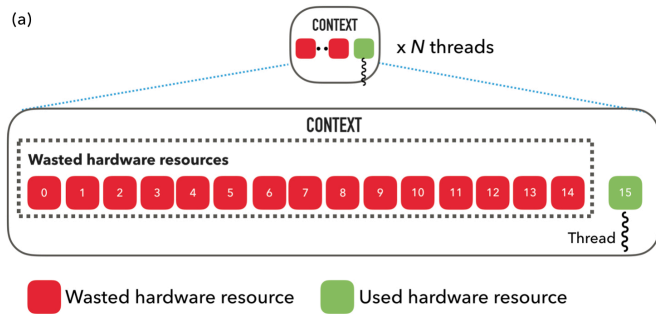


Fig. 2. (a) Demonstration of 93.75% hardware resource wastage per context in the naïve approach. (b) (i) Throughput (higher is better) and (ii) number of wasted hardware resources (lower is better) with state-of-the-art endpoints on Mellanox’s ConnectX-4 adapter.

endpoint configuration in the MPI-everywhere model where each MPI process has its own context. Although such a naïve approach can achieve the maximum throughput for a given number of threads, it wastes the hardware’s limited resources. Figure 2(a) shows how this naïve approach translates to 93.75% hardware resource wastage on a modern Mellanox mlx5 InfiniBand device. In order to achieve maximum resource efficiency, multiple threads can share just one endpoint, which is the case for the MPI+threads model in state-of-the-art MPI implementations. Doing so, however, drastically impacts communication throughput. Figure 2(b) shows the tradeoff between throughput and hardware resource wastage in a multithreaded environment that emulates state-of-the-art endpoints in the MPI-everywhere and MPI+threads models.

Note that the MPI+threads model itself does not force the extreme of using a single endpoint for communication by all threads. That is simply how the state-of-the-art MPI libraries implement it. Unlike MPI everywhere, the MPI+threads environment allows for any arbitrary level of sharing of communication resources between the different threads. The question that we really need to answer is, *what level of resource sharing is ideal?* As is the case with any computer science question that is worth its salt, the answer is, *it depends*. If one is looking for the least amount of resources to use without losing any performance compared with the MPI-everywhere model, a certain set of resources can be shared while others cannot. If a small percentage of performance loss is acceptable, a different division of shared vs. dedicated resources would be ideal. If resource efficiency is the most important criterion and additional performance loss is acceptable, yet another division of shared vs. dedicated resources would be ideal. Understanding this tradeoff space between performance and resource usage is the primary goal of this paper.

To that end, this paper makes the following contributions.

- (1) We demonstrate the two extreme cases—one where all threads share a single communication endpoint and another where each thread gets its own dedicated endpoint. We showcase the inefficiencies in both these cases.
- (2) We explore the tradeoff space between performance (communication throughput) and communication resource usage in a multithreaded environment. In Section III, we first discuss the communication resources

of an endpoint. In Section V, we thoroughly analyze the different levels of resource sharing in MPI+threads environments in the context of Mellanox InfiniBand, the most popular high-speed interconnect on the TOP500 and also the preferred interconnect for both artificial intelligence and high-performance computing (HPC) [5].

- (3) Using the lessons learned from our analysis, we design efficient resource-sharing models in Section VI to provide *scalable communication endpoints*. Scalable endpoints provide a wide range of resource-sharing models, ranging from fully independent to fully shared communication paths. Our evaluation on scalable endpoints in Section VII shows that fully independent communication paths can achieve performance as high as MPI-everywhere endpoints by using 3.2x fewer resources.

II. BACKGROUND

InfiniBand (IB) is the popular choice among high-speed interconnects. Mellanox Technologies is the most renowned IB vendor, powering 216 systems (both IB and Ethernet) on the TOP500 [5]. Hence, we study the mlx5 provider of Verbs, the IB software stack. Mellanox’s Connect-IB adapter and its ConnectX series, starting from ConnectX-4, are mlx5 devices.

A. InfiniBand Resources

The software bidirectional communication portal in IB is the queue pair (QP): a pair of send and receive FIFO queues, to which work queue entries (WQEs), IB’s message descriptors, are posted. Each QP is associated with a completion queue (CQ) that contains completion queue entries (CQEs) corresponding to the completion of signaled WQEs. To create a QP, we need at least one memory buffer (BUF), device context (CTX), protection domain (PD), and CQ. A memory region (MR) is required if the NIC needs direct access to memory. Chapter 10 of the IB specification details the IB resources [2]. Additionally, we can assign QPs to thread domains (TDs) to provide single-threaded access hints to the QPs in a TD.

The CTX is the container of all IB resources and is also a slice of the network hardware, containing a subset of the NIC’s hardware resources. In mlx5 devices, the hardware resources are part of the user access region (UAR) of the NIC’s address space. Each UAR page consists of two micro UARs (uUARs). By default, a CTX contains eight UARs (UAR pages) and,

hence, 16 uUARs. The user’s QPs are mapped to one of the *statically allocated* uUARs unless a QP is part of a TD in which case the QP is mapped to a uUAR in a UAR that was *dynamically allocated* during TD creation. [10] details these resources and describe mlx5’s uUAR-to-QP assignment policy.

B. InfiniBand Operational Features

To send a message on InfiniBand, the application calls `ibv_post_send`. What follows is a series of coordinated operations between the CPU and the NIC to fetch the WQE (DMA read), read its payload (DMA read), and signal its completion (DMA write). [10] portrays the operations involved.

The NIC is typically a PCIe device and hence, the overhead of the operations is multiple PCIe round-trip latencies. Naturally, reducing the number of round-trip latencies for small messages impacts throughput significantly. *Inlining*, *Postlist*, *Unsignaled Completions*, and *BlueFlame* are IB’s operational features that help reduce this overhead. We describe them below considering the depth of the QP to be n .

Postlist. Instead of posting only one WQE per `ibv_post_send`, IB allows the application to post a linked list of WQEs with just one call to `ibv_post_send`. It can reduce the number of *DoorBell* rings from n to 1.

Inlining. Here, the CPU copies the data into the WQE. Hence, with its first DMA read for the WQE, the NIC gets the payload as well, eliminating the second DMA read for the payload.

Unsignaled Completions. Instead of signaling a completion for each WQE, IB allows the application to turn off completions for WQEs provided that at least one out of every n WQEs is signaled. Turning off completions reduces the DMA writes of CQEs by the NIC. Additionally, the application polls fewer CQEs, reducing the overhead of making progress.

BlueFlame. *BlueFlame* is Mellanox’s terminology for programmed I/O—it writes the WQE along with the *DoorBell*, cutting off the first DMA read. With *BlueFlame*, the UAR pages are mapped as write-combining (WC) memory. Hence, the WQEs sent using *BlueFlame* are buffered through the CPU’s WC buffers. Note that *BlueFlame* is not used with *Postlist*; the NIC will DMA-read the WQEs in the linked list.

Using both *Inlining* and *BlueFlame* for small messages eliminates two PCIe round-trip latencies. While the use of *Inlining* and *BlueFlame* is dependent on message size, the use of *Postlist* and *Unsignaled Completions* is reliant primarily on the user’s design choices and application semantics.

III. COMMUNICATION RESOURCES

To send messages across the network, the software (CPU) coordinates with the hardware (NIC) to *initiate* a transfer and confirm its *completion*. This coordination occurs through three communication resources: a software transmit queue, a software completion structure, and a NIC’s hardware resource. The three interact using the mechanisms described in [10] and features described in Section II-B. In IB, the transmit queue is the QP, the completion structure is the CQ, and the hardware resource is the uUAR contained within a UAR page. The QP,

TABLE I
BYTES USED BY MLX5 VERBS RESOURCES

CTXs	PDs	MRs	QPs	CQs	Total
256K	144	144	80K	9K	345K

UAR, and uUAR make up the *initiation* interface; the CQ is the *completion* interface.

The threads of a MPI+threads application eventually map to QPs, and the QPs eventually map to a uUAR on a UAR of the NIC. As seen in Section II-A, the interconnect’s driver dictates the mapping between the transmit queues and the hardware resources while the user decides the mapping between the transmit queues and completion structures. Multiple QPs could share the same CQ, or each could have its own.

The QP and CQ are associated with circular buffers that contain their WQEs and CQEs, respectively. The CPU writes to the QP’s buffer, and the NIC DMA-reads it when *Inlining* is not used. The NIC DMA-writes the CQ’s buffer and the CPU reads it when polling for progress. Both buffers are pinned by the operating system during resource creation.

The QP and CQ occupy memory with their circular buffers. So, every time we create a QP or a CQ, we impact memory consumption. Table I shows the memory used by each type of a Verbs resource (for mlx5) that is required to open a QP. Creating one endpoint requires at least 354 KB of memory, with the CTX occupying 74.2% of it.

However, the memory usage of the QP and the CQ is on the order of kilobytes, whereas the memory on the nodes of clusters and supercomputers is typically on the order of hundreds of gigabytes. Hence, we will notice a formidable impact on memory consumption only when the number of the Verbs resources is on the order of thousands. The impact of creating a QP or a CQ on memory is not of immediate concern.

On the other hand, the limit on the hardware resource is much smaller: 8K UAR pages on the ConnectX-4 NIC with only two uUARs per UAR. The situation is similar for other interconnects such as Intel Omni-Path, where the maximum number of hardware contexts on its NIC is 160 [4]. The 8K UARs on ConnectX-4 translates to a maximum of 907 CTXs, considering that the user creates a TD-assigned QP contained within its own CTX for each thread. Each CTX contains a total of 18 uUARs—the 16 static ones plus the two from the TD’s dynamically allocated UAR (see Section II-A). The resource wastage of this approach is a staggering 94% since it uses only one uUAR out of 18. Arguably, we will not run out of hardware resources even if we create one endpoint per core on existing processors with this approach, but eliminating this huge wastage would enable vendors to significantly reduce the power and cost of their NICs. Such high wastage translates to requiring a second NIC on the node after only marginally utilizing the resources on the first.

IV. EVALUATION SETUP

To evaluate the impact of resource sharing on performance, we write a multithreaded “sender-receiver,” RDMA-write mes-

sage rate benchmark. We choose RDMA writes to eliminate any receiver-side processing on the critical path.

We conduct our study on the Joint Laboratory for System Evaluation’s Gomez cluster (each node has quad-socket Intel Haswell processors with 16 cores/socket and one hardware thread/core) using a patched rdma-core [11] library that contains the infrastructure to allow for maximally independent paths and disabled mlx5 locks as described in Section V-B. The two nodes are connected via a switch, and each node hosts a single-port Mellanox ConnectX-4 NIC. We ensure that each thread is bound to its own core. For repeatable and reliable measurements, we disable the processor’s turbo boost and set the CPU frequency to 2.5 GHz.

The design of our message-rate benchmark is adopted from `perftest` [3]. The loop of a thread iterates until all its messages are completed. In each iteration, the thread posts WQEs on a QP of depth, d in multiples of `Postlist`, p requesting for one signaled completion every q WQEs, where q is the value of `Unsignaled Completions`. In each poll on the CQ, the thread requests for $c = d/q$ completions, namely, all possible completions in an iteration. The depth of the CQ is c .

`Postlist` and `Unsignaled Completions` control the rate and amount of interaction between the CPU and NIC. Empirically, we find that setting $p = 32$ and $q = 64$ achieves the maximum throughput for 16 threads; hence, we use them as our default values. Note that we define the values of `Postlist` and `Unsignaled Completions` with respect to the threads, not to their associated QPs.

To study the effect of an IB feature, we remove that feature while using others, referring to this case as “All w/o f ,” where f is the feature of interest. To disable `BlueFlame`, we set the `MLX5_SHUT_UP_BF` environment variable. To enable `Inlining`, we set the `IBV_SEND_INLINE` flag on the send-WQE. We use “w/o `Postlist`” to mean $p = 1$, and similarly “w/o `Unsignaled`” to mean $q = 1$.

Figure 3 shows the scalability* of communication throughput across features and communication resource usage of endpoints created with one TD-assigned QP per context per thread for 2-byte RDMA writes. We observe that the number of QPs and CQs is an identity function of the number of threads and increases their memory consumption from 89 KB with one thread to 1.39 MB with 16 threads. The usage of UARs and uUARs also grows by a factor of 9 and 18, respectively. The reason is that each CTX containing one TD allocates 9 UARs and each UAR consist of two uUARs.

V. RESOURCE-SHARING ANALYSIS

From an analytical perspective, a thread can map to the hardware resources in four possible ways. Figure 4(b) demonstrates the four ways described below.

- (1) *Maximum independence* – There is no sharing of any hardware resource between the threads; each is assigned to its own UAR page (used in MPI everywhere).

*The NIC is attached only to the first socket; cross-socket NIC behavior is out of the scope of this work.

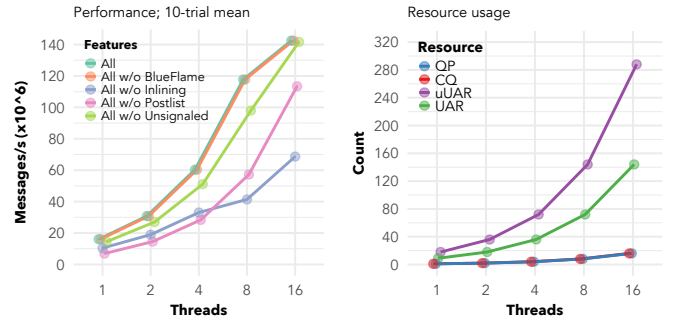


Fig. 3. Scalability using a TD-assigned-QP per CTX per thread. Left: Throughput across features. Right: Resource usage.

- (2) *Shared UAR* – The threads are assigned to distinct uUARs sharing the same UAR page (mlx5 default for multiple TDs described in Section V-B).
- (3) *Shared uUAR* – Although the threads have their own QPs, the distinct QPs share the same uUAR (medium-latency uUARs in [10]). A lock is needed on the shared uUAR for concurrent `BlueFlame` writes.
- (4) *Shared QP* – The threads share the same QP (used in state-of-the-art MPI+threads), in which case a lock on the QP is needed for concurrent device WQE preparation. The lock on the QP also protects concurrent `BlueFlame` writes on the uUAR since the lock is released only after a `BlueFlame` write.

Sharing software and hardware communication resources at different levels improves resource efficiency but can hurt throughput. Below, we explore the tradeoff space between resource efficiency and communication throughput from the perspective of the Mellanox IB user while considering the various IB features described in Section II-B. The user allocates and interacts with the communication resources through the IB resources shown in Figure 4(a). Each of those objects represents a level of sharing between threads. Hence, we analyze the impact of sharing each IB resource on performance and resource usage. We verify our analyses for 16 threads using the setup described in Section IV.

In the figures below, x -way sharing means the resource of interest is being shared x ways. For example, 8-way sharing means the resource is shared between 8 threads (two instances of the shared resource). Moreover, we are interested in the change in throughput with increasing sharing rather than the absolute throughput obtained by using certain features.

Starting with naïve endpoints—each thread driving its own set of resources using a TD-assigned QP—we move down each level of IB resource sharing according to the hierarchical relation shown in Figure 4(a). Figure 3 shows the performance and resource usage of this approach for 16 threads.

A. Memory Buffer Sharing

The highest level of sharing is the non-IB resource: memory buffer. We define the BUF to be the pointer to the payload of the message. If the payload size is small enough, it can be inlined within the WQE; that is, the CPU will read it. By default, the maximum message size that can be inlined on

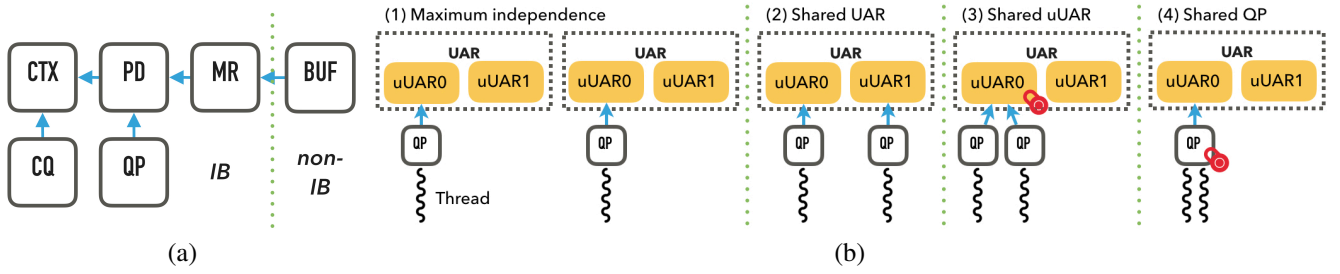


Fig. 4. (a) Hierarchical relation between the various Verbs resources (the arrow points to the parent); each resource can have multiple children but only one parent. (b) Four levels of thread-to-uUAR mapping in mlx5 between independent threads.

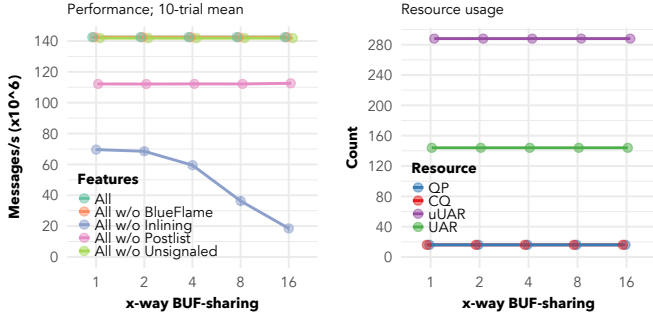


Fig. 5. Message rate (left) and communication resource usage (right) with increasing BUF sharing across 16 threads.

ConnectX-4 exposed through Verbs is 60 bytes. Therefore, for any larger message, the NIC must DMA-read the payload. **Performance.** When the CPU reads the payload, sharing this BUF between the threads is safe since concurrent reads to the same memory location in a CPU are harmless. When the NIC reads the payload, however, its TLB design is important since a virtual-to-physical address translation is imperative for the DMA read. The NIC typically has a multirail TLB design that handles multiple transactions in parallel in order to sustain the high speed of the NIC’s ASIC. The load is distributed across the TLBs by using a hash function. If this hash function is based on the cache line, concurrent DMA reads to the same cache line will hit the same translation engine, serializing the reads. With a shared BUF, the WQEs of multiple threads would point to the same cache line, serializing the DMA reads.

Figure 5 indeed shows that the throughput decreases with increasing BUF sharing without *Inlining* that is, when the NIC reads the payload. To further validate our analysis, Figure 6(a) shows that independent 2-byte buffers without 64-byte cache alignment also hurt performance since all 16 buffers are on the same cache line. While the total number of PCIe reads (measured using PMU tools [9]) with and without cache alignment is equal, Figure 6(b) shows that the rate of these PCIe reads is much slower when the buffers are on the same cache line.

Resource usage. The BUF is a non-IB resource. Hence, it does not affect the usage of any of the communication resources, as we can see in Figure 5.

B. Device Context Sharing

We note that the Verbs user gets maximally independent (level 1 in Figure 4(b)) paths without CTX sharing since the QPs naturally get assigned to uUARs on different UARs.

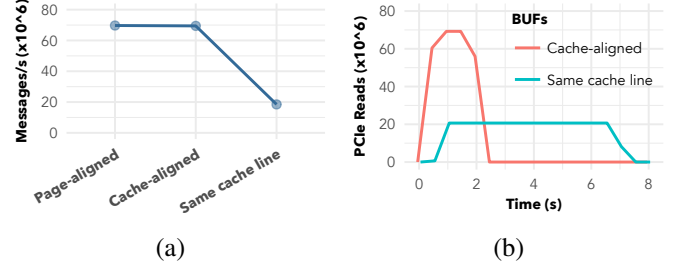


Fig. 6. Effects on (a) message rate and (b) PCIe reads with and without cache-aligned buffers.

Within a shared CTX, however, the user has no way to explicitly request maximally independent paths for multiple QPs. When the user creates multiple TDs, the mlx5 provider can assign the threads to a uUAR using either the first or the second level of sharing, as shown in Figure 4(b). Currently, the mlx5 provider is hardcoded to use the second level of sharing for multiple TDs, restricting the user from creating maximally independent QPs within a CTX. More abstractly, the Verbs users today have no way to request a sharing level for the QPs/TDs they create. The number of levels of sharing is provider specific.

To overcome this Verbs design limitation, we propose a variable, *sharing*, in the TD initialization attributes (`struct ibv_td_init_attr`) that are passed during TD creation. The higher the value of *sharing*, the higher is the amount of hardware resource sharing between multiple TDs. A *sharing* value of 1 refers to maximally independent paths. In mlx5, only two levels of sharing exist for TDs, corresponding to (1) and (2) in Figure 4(b).

Note that the second uUAR of the UAR dedicated to a maximally independent TD is wasted. Since the number of hardware resources is limited, the user can request only a certain maximum number of independent hardware resources within a CTX. This would be half of the maximum number of UARs the user can dynamically allocate using TDs. In mlx5, the maximum number of maximally independent paths is 256.

Furthermore, we note that when the user assigns a QP to a TD, the lock on the QP is still obtained. The mlx5 provider currently removes only the lock on the uUAR that the TD is assigned to. Since the user guarantees no concurrent access from multiple threads to a QP assigned to a TD, the lock on the QP itself can be disabled. We optimize the mlx5 provider for this case [6].

Performance. For maximally independent threads, sharing

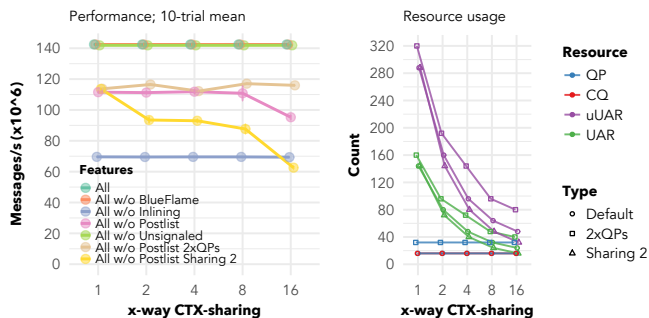


Fig. 7. Message rate (left) and communication resource usage (right) with increasing CTX sharing across 16 threads.

the CTX should not affect performance since we emulate the thread-to-uUAR mapping in the MPI-everywhere model. Sharing a CTX with the second level of sharing between threads could hurt performance—the uUARs on the same UAR could be sharing the same set of the NIC’s registers, negatively impacting throughput. Additionally, the CPU architecture’s implementation of flushing write combining memory can impact performance in the second level of sharing since the memory attribute of the uUARs is set at the page-level granularity by using the Page Attribute Table (PAT) [8].

Figure 7 shows that sharing the CTX does not hurt performance except when we do not use *Postlist* that is, when we use *BlueFlame* writes. For example, we notice a 1.15x drop in performance going from 8-way to 16-way CTX sharing even with maximally independent TDs. While the engineers at Mellanox are able to reproduce this drop even on the newer ConnectX-5, the cause for the drop is unknown. We discovered that creating twice the number of maximally independent TDs but using only half of them (even or odd ones) can eliminate this drop, as seen in the “All w/o Postlist 2xQPs” line. Additionally, from the “All w/o Postlist Sharing 2” line, we can see the harmful effects of sharing a UAR when the mlx5 provider is hardcoded to use the second sharing level for assigning TDs within a shared CTX to uUARs.

While this evaluation validates the need for maximally independent paths, it does not explain the decline in throughput when there are concurrent *BlueFlame* writes to distinct uUARs sharing the same UAR page. Finding the precise reason for this behavior is hard since the hardware-software interaction is dependent on the aforementioned proprietary technologies. **Resource usage.** Sharing the CTX is critical for hardware resource usage, as seen in Figure 7. The reason is that a maximally independent TD within a shared CTX adds only 1 UAR as opposed to 9 UARs when it is created within its own independent CTX. Also, the 16 uUARs and 8 UARs statically allocated by the mlx5 provider during CTX creation (see Section II-A) are wasted only once. Nonetheless, maximally independent TDs will waste one uUAR per thread. While sharing the CTX does not impact QP and CQ usage, it does reduce the overall memory consumption. For example, when shared between 16 threads, it can reduce the overall memory consumption by 9x (from 5.15 MB to 0.35 MB).

Creating twice as many TDs (“2xQPs” in Figure 7) in-

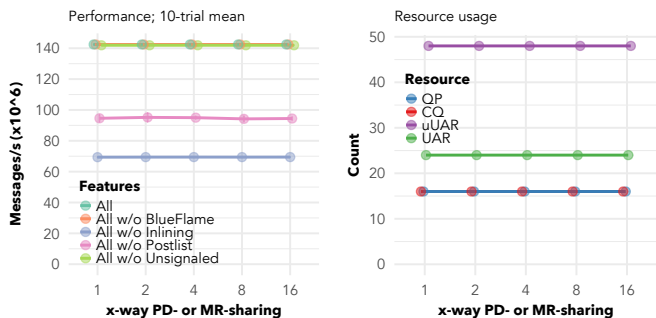


Fig. 8. Message rate (left) and communication resource usage (right) with increasing PD or MR sharing across 16 threads.

creases resource usage since each the extra 16 maximally independent TDs allocates their own QP and UAR. The second level of sharing that mlx5 is hardcoded to use consumes 2x fewer UARs than do maximally independent TDs.

C. Protection Domain Sharing

The protection domain is just a means of isolating a collection of IB resources. Resources contained under different PDs cannot interact with each other.

Performance. The software PD object is not accessed on the critical data-path; the protection checks occur in the NIC. Hence, from a performance perspective, sharing a PD between multiple threads would be harmless, as observed in Figure 8.

Resource usage. The PD does not impact the usage of any of the communication resources, as we can see in Figure 8. The uUAR and UAR values reflect those of one CTX since the PD can be shared only within a CTX.

D. Memory Region Sharing

The MR is an object that pins memory in the virtual address space of the user with the OS and prepares it for DMA accesses from the NIC.

Performance. Sharing the MR between threads will have no impact on performance since the MR is just an object that points to a registered memory region. The MR may span multiple contiguous BUFs. Sharing an MR containing only one BUF means that the threads are sharing the BUF as well, which implies the same effects of BUF sharing. Figure 8 confirms that sharing the MR does not affect performance as long as the threads have independent cache-aligned buffers.

Resource usage. The MR does not control the allocation of any of the communication resources. Hence, sharing it will have no impact, as we can see in Figure 8.

E. Completion Queue Sharing

The Verbs user can map multiple QPs to the same CQ, allowing for CQ-sharing between threads. In a latency-bound application, the user actively polls the CQ on the critical data-path to confirm progress in communication.

Performance. The CQ has a lock that a thread will acquire before polling it. Hence, the threads sharing a CQ will contend on its lock. Additionally, if QP i and QP j share a CQ, then thread i driving QP i can read QP j ’s completions. Hence, the completion counter for any thread i requires atomic

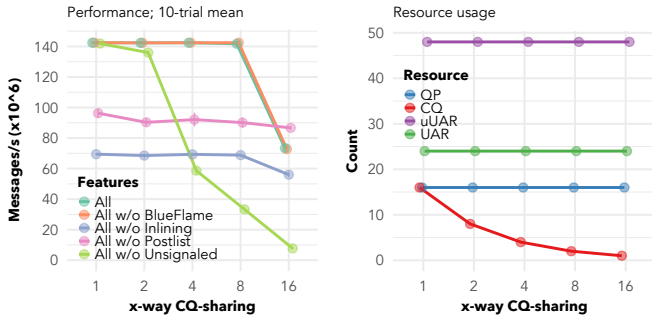


Fig. 9. Message rate (left) and communication resource usage (right) with increasing CQ sharing across 16 threads.

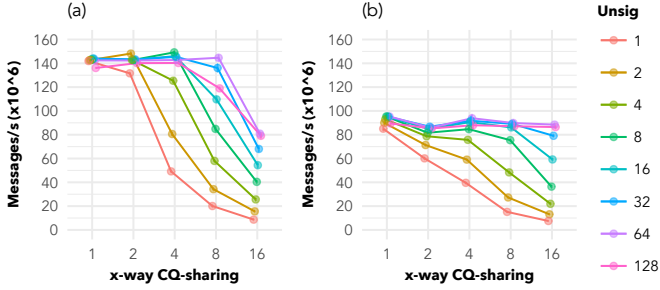


Fig. 10. (a) Postlist size of 32, (b) Postlist size of 1.

updates. Atomics and locks are obvious sources of contention when sharing CQs between threads. Figure 9 demonstrates these hurtful effects of CQ sharing. The effects are most noticeable in 16-way sharing because there exists a tradeoff space between the benefits of *Unsignaled Completions* and the overheads of CQ sharing. Figure 10(a) portrays this tradeoff space. Lower values of *Unsignaled Completions* imply that the thread reads more completions from the CQ than for higher values, translating to longer hold-time of the shared CQ’s lock. Thus, the impact of lock contention is most visible in “All w/o Unsignaled.” For higher *Unsignaled Completion*-values, we see a drop only after a certain level of CQ sharing because the benefits of *Postlist* outweigh the impact of contention. Removing *Postlist* shows a linear decrease in throughput with increasing contention in Figure 10(b).

We note that even if the Verbs user can guarantee single-thread access to a CQ, the standard CQ does not allow the user to disable the lock on the CQ. The extended CQ, on the other hand, allows the user to do so during CQ creation (`ibv_create_cq_ex`) with the `IBV_CREATE_CQ_ATTR_SINGLE_THREADED` flag.

Resource usage. Sharing the CQ translates to fewer circular buffers, and hence it reduces the memory consumption of the completion communication resource. But it does not affect hardware resource usage, as we can see in Figure 9. The uUAR and UAR usage shown corresponds to that of one CTX since a CQ can be shared only within a CTX.

F. Queue Pair Sharing

Ultimately, the user can choose to share the queue pair between threads to achieve maximum resource efficiency. This is the case in state-of-the-art MPI implementations.

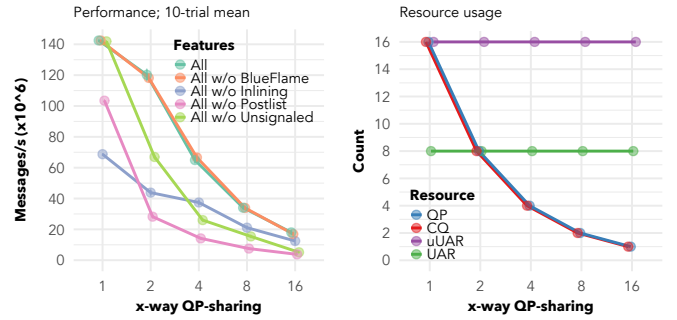


Fig. 11. Message rate (left) and communication resource usage (right) with increasing QP sharing across 16 threads.

Performance. The QP has a lock that a thread needs to acquire before posting on it. Hence, threads will contend on a shared QP’s lock. Additionally, the threads need to coordinate to post on the finite QP-depth of the shared QP using atomic fetch and decrement the QP-depth value. These locks and atomics are sources of contention when sharing QPs. Most important, the NIC’s parallel capabilities are not utilized with shared QPs since each QP is assigned to only one hardware resource through which the messages of multiple threads are serialized. Figure 11(a) shows the expected decline in throughput with increasing QP-sharing. Removing *Postlist* is more detrimental than removing *Unsignaled Completion* because the contention on the QP’s lock without *Postlist* is higher.

Resource usage. Sharing the QP means fewer circular buffers for the WQEs and hence lower memory consumption. It does not affect hardware resource usage, as we can see in Figure 11(b). QP sharing reduces the number of both QPs and CQs, reducing the total memory consumption of the software communication resources by 16x with 16-way sharing.

Summary. Below are the lessons learned from our analysis.

- Each thread must have its own cache-aligned buffer to prevent a performance drop.
- CTX-sharing is the most critical for the usage of hardware resources. With 16-way sharing, “2xQPs” can achieve the same performance as independent CTXs using 80 uUARs instead of 288. If 20% less performance is acceptable, we can use maximally independent TDs that use 6x fewer resources. If 50% less performance is acceptable, we can use “Sharing 2” that uses 9x fewer resources.
- Sharing the PD or the MR will not hurt performance, while keeping them independent will not utilize any communication resource.
- Only QP- and CQ-sharing affects the memory consumption of the software resources. However, the reduction in memory usage by sharing them is not as critical as the consequent drop in performance. For example, 16-way sharing of the CQ improves memory usage by 1.1x but can result in an 18x drop in performance.

VI. DESIGNING SCALABLE ENDPOINTS

Building on our analysis, we define the *scalable endpoints* resource sharing model that concretely categorizes the design space of multiple communication endpoints into six categories.

Below we describe the design of the initiation interface in each category, state how the user can create it, discuss what occurs internally in the IB stack, and discuss its implications on performance and resource usage. For simplicity, we maintain a separate CQ for each QP.

MPI everywhere. This category emulates the endpoint configuration when multiple ranks run on a node. It represents level 1 in Figure 4(b). The user creates this by creating a separate CTX for each thread, each containing its own QP and CQ. Within each CTX, the `mlx5` driver assigns the QP to a low-latency uUAR. Since each CTX contains 8 UARs, consecutive QPs naturally get assigned to distinct UAR pages. The performance of this category is the closest to the best possible since there is no sharing of resources. It is not the best since the lock on the QP is still taken even though no other thread contends for it. The resource usage of this category is high: every CTX allocates 8 UARs. Additionally, it is wasteful since only 1 of the 16 allocated uUARs is used per thread. The memory consumption increases linearly with the number of threads since the number of QPs and CQs is an identity function of the number of threads.

2xDynamic. This category also represents a 1-to-1 mapping between a uUAR and a thread. Unlike MPI everywhere, however, the user creates only one CTX for all the threads and creates twice as many TD-assigned-QPs as threads. The threads use only the even or odd QPs. The `mlx5` provider dynamically allocates a new UAR page for each TD and assigns the first uUAR to the TD, enabling a 1-to-1 mapping. This category delivers the best performance. Since the number of QPs is twice the number of threads, however, each thread wastes 1 dynamically allocated UAR, 3 uUARs, and 1 QP. The memory consumption of QPs and CQs is twice that of MPI everywhere. The statically allocated hardware resources are wasted regardless of the number of threads.

Dynamic. This category also represents a 1-to-1 mapping between a uUAR and a thread, but the number QPs equals the number of threads. The user creates this configuration similar to “2xDynamic” by creating only as many QPs as threads. According to Section V-B, this configuration hurts communication throughput. In terms of resource usage, however, only one uUAR is wasted per thread. The 8 statically allocated UARs are naturally wasted; none of the dynamically allocated UARs are wasted. The memory consumption of QPs and CQs is half of that in “2xDynamic” and same as MPI everywhere.

Shared Dynamic. This category represents level 2 in Figure 4(b). The user creates this configuration using a shared CTX, similar to the way in “Dynamic,” but assigns each QP to a TD with the second level of sharing. The `mlx5` driver will dynamically allocate UARs only for the even TDs and map the even TDs to the first uUAR and the odd TDs to the second uUAR of the allocated UAR. According to Section V-B, sharing the UAR will hurt performance. The hardware resource usage is less than with “Dynamic” since only half as many UARs and uUARs as threads are allocated. Apart from the 8 statically allocated UARs and uUARs, none of the dynamically allocated resources are wasted. The memory consumption of

QPs and CQs is equivalent to that of “Dynamic.”

Static. The user uses the statically allocated resources within a CTX, resulting in a many-to-one mapping between the threads and uUARs (and UARs). To do so, the user simply creates a QP for each thread within a shared CTX without any TDs. The final state of the mapping for a given number of QPs is dependent on the driver’s assignment policy. In `mlx5`, with 16 QPs, we end up with a combination of the second and third level of sharing in Figure 4(b)—the 5th and 16th QP are mapped to the same uUAR (third level), while the others are mapped to the rest of the uUARs using the second level of sharing. The hardware resource usage is the number of statically allocated resources. Resources are wasted only when the number of threads is less than 16. The memory consumption is equivalent to that of “Dynamic.”

MPI+threads. This category represents level 4 in Figure 4(b). The user creates this by creating only 1 CTX, 1 QP, and 1 CQ. The `mlx5` driver assigns the one QP to a low-latency uUAR. The performance of this category is the worst possible since the communication of all the threads is bottlenecked through one QP. The resource usage of this category is not a function of the number of threads and hence is the best possible. All threads allocate only 8 UARs, 16 UARs, 1 QP, and 1 CQ.

Note that the CQ can be shared in any manner in the above categories and its impact is orthogonal to the effects of the initiation interface.

VII. EVALUATING SCALABLE ENDPOINTS

We evaluate the performance and resource usage of scalable endpoints described in Section VI on two benchmarks, namely, global array and 5-point stencil on our two-node evaluation setup[†]. We limit our evaluation to conservative application semantics—those that do not allow *Postlist* and *Unsignaled Completions* and focus on *BlueFlame* writes instead of *Door-Bells* since they are latency oriented.

Global array benchmark. The pattern of fetching and writing tiles from and to a global array is at the core of many scientific applications such as NWChem [17], which constitutes a multidimensional double-precision matrix multiply (DGEMM).

We implement a DGEMM benchmark ($A \times B = C$), where the global matrices A , B , and C reside on a server node and a client node performs the DGEMM using Verbs for internode communication. We design the benchmark such that all the QPs share the same PD but each has three BUFs and three MRs—one for each of the three tiles from A , B , and C .

Figure 12 shows the performance and resource usage of scalable communication endpoints for 16 threads. Performance decreases with lower resource usage. For RDMA writes, for example, we observe that using maximally independent TDs with twice the number of QPs (2xDynamic) gives us 108% of the performance of dedicated endpoints (MPI everywhere) while using only 31.25% as many hardware resources. Maximally independent paths with as many QPs as threads (Dy-

[†]Thread domains are supported only kernel 4.16 onward; the latest stable kernel was 4.17.2, hence, only two nodes since the combination of a `mlx5` device along with the latest stable kernel was a rarity.

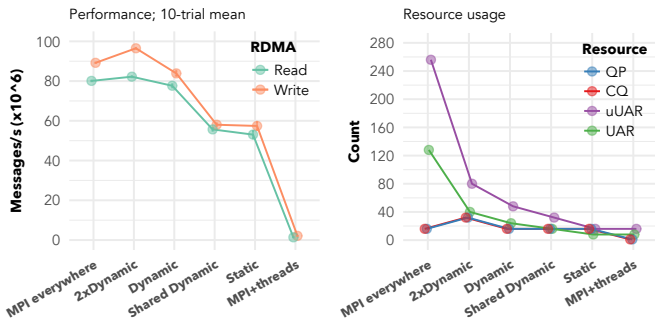


Fig. 12. Scalable endpoints for the global array kernel with 16 threads. Left: Communication throughput. Right: Communication resource usage

dynamic) gives us 94% of the performance of MPI everywhere while using 18.75% as many hardware resources. Sharing the UAR (Shared Dynamic) gives us 65% of the performance using 12.5% of the hardware resources. Sharing the uUAR (Static) gives us 64% of the performance using 6.25% as many hardware resources. We observe only a minimal drop in performance in Static since only two threads share the uUAR in Static; the rest share the UAR (see Section VI), and hence we observe performance similar to Shared Dynamic. Finally, sharing the QP results in only 3% of the performance while still using 6.25% as many hardware resources.

The memory consumption of QPs and CQs is the same for all categories except 2xDynamic and MPI+threads. While the number of QPs and CQs in 2xDynamic is twice that of MPI everywhere, the overall memory usage in the former is 3.27x lower (1.64 MB vs 5.39 MB; see Section III) since MPI everywhere has 16 CTXs while 2xDynamic has only one. The memory consumption is the lowest in MPI+threads with only one QP and one CQ.

Stencil benchmark. Stencil codes are at the heart of various application domains such as computational fluid dynamics, image processing, and partial differential equation solvers. We evaluate scalable endpoints on a 5-pt stencil benchmark with a 1D partitioning of the grid. Figure 13 shows the design of our benchmark. We vary the number of ranks per node and threads per rank such that the total number of hardware threads engaged is 16, the number of cores in a socket. Each rank gets its tile from the grid, and each thread gets a corresponding subtile. Each thread requires two QPs, one for each of its neighbors. We map the two QPs to one CQ. Hence the number of QPs is twice the number of CQs for all cases. Figure 14 shows the performance,[‡] and resource usage of scalable endpoints for the different hybrid scenarios.

For each category, a higher number of processes performs better than a lower one. For MPI-everywhere endpoints, for example, the fully hybrid approach (1.16) performs 1.4x worse than the processes-only approach (16.1). The reason for this behavior is that the number of messages with processes only is 16x higher, while 16 threads per rank can exchange the halo only 7.67x faster than with one thread per rank.

[‡]The message rates are above 150 million, the maximum reported for ConnectX-4 [1] since a majority of the halo exchanges are intranode. Intranode communication in InfiniBand still involves the NIC.

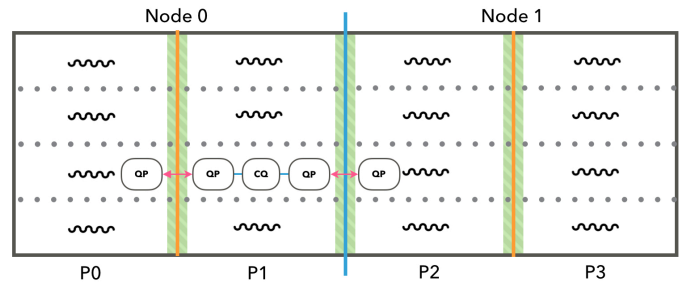


Fig. 13. 1-D partitioning of a grid for the 5-pt stencil between two nodes, four ranks (P0..P3), and four threads per rank showing the QP-CQ connection for each thread using one sample. The shaded regions are the halo regions.

In the processes-only case, there is no resource sharing since each process has only one thread. 2xDynamic, Dynamic, and Shared Dynamic achieve 106% of MPI-everywhere’s performance because of the absence of the lock on the QP. Static produces 100% performance since the lock on its QP exists. MPI+threads achieves 87% of the performance even though there is no contention between threads, because of the overhead of atomics and additional branches associated with QP-sharing. In 16.1, the number of QPs and CQs is the same for all cases except for 2xDynamic, where they are 2x higher. The hardware resource usage is higher in 2xDynamic, Dynamic, and Shared Dynamic since they waste the statically allocated resources in each process, unlike other categories.

For the hybrid cases, we observe a performance trend similar to the global array kernel. 2xDynamic achieves 103% of the performance of MPI everywhere; and with increasing resource-sharing, we improve resource usage but lose performance. In the case of 4.4, of the eight QPs per CTX, the fifth QP uses the first level of sharing in Static, resulting in eight such QPs in total; hence, it performs better than Shared Dynamic wherein all the QPs use only the second level of sharing. Similarly, in 1.16, of the 32 QPs per CTX, 28 use the third level of sharing in Static, hence performing worse than Shared Dynamic. For a given category, the hardware resource usage is lower when the number of processes is smaller since fewer processes mean fewer CTXs, and hence, the total number of statically allocated resources is smaller. Similarly, the number of QPs and CQs is the same for all hybrid cases in all categories except in MPI+threads, where it is a function of the number of processes.

VIII. RELATED WORK

To the best of our knowledge, the resource-sharing analysis in this paper to design a resource-sharing model at the low level of interconnects is the first of its kind. The idea of multiple endpoints for multinode programming models such as MPI and Unified Parallel C (UPC), however, is not new. The research in this domain is motivated by the same problem: loss in communication throughput in hybrid environments.

MPI endpoints. Dinan et al. [12] enable multiple communication endpoints by creating additional MPI ranks that serve as the “MPI endpoints.” The threads within the MPI ranks then map to the MPI endpoints, achieving the same configuration as MPI-everywhere endpoints since each MPI endpoint has

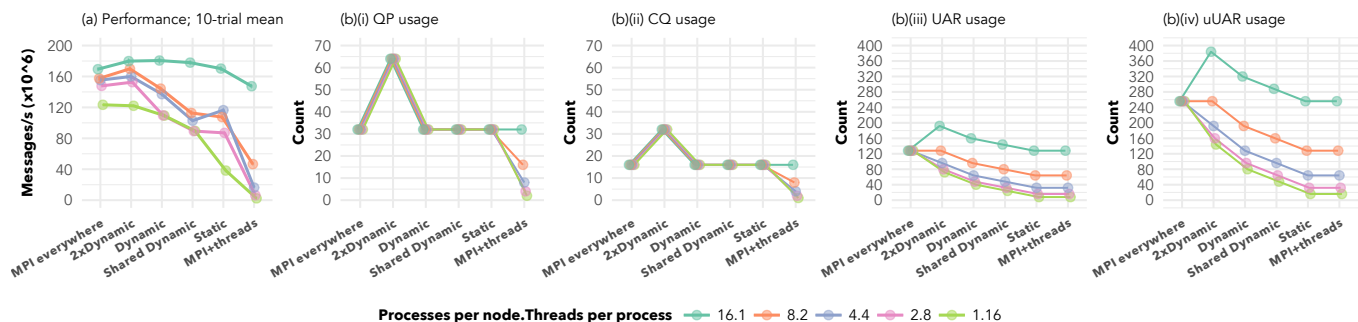


Fig. 14. (a) Message rate and (b) resource usage of (i) QP, (ii) CQ, (iii) UAR, and (iv) uUAR of 16 threads with scalable endpoints for a 5-pt stencil kernel.

its CTX. However, they do not consider the resource usage of their approach. Consequently, the 93.75% wastage of resources still holds with MPI endpoints. Our work explores the tradeoff space between performance and resource usage instead of providing one solution, allowing users to choose the best endpoint for their needs.

PAMI endpoints. Tanase et al. [15] implement multiple endpoints for the IBM xUPEC runtime by assigning contexts to UPC threads with a one-to-one mapping. While their work is a complete solution, it does not demonstrate the indirect impact on the resource usage. We show a holistic picture of the different mappings between threads and hardware resources and discuss the tradeoff between performance and resource usage for each mapping.

UPC endpoints. Luo et al. [14] implement network endpoints for the UPC runtime. However, their work does not consider the mapping between the runtime’s network endpoints and the interconnects network resources. Consequently, their work does not evaluate the hardware-resource utilization of their implementation, which is an essential factor for understanding the scalability of multiple communication endpoints.

IX. CONCLUSIONS

For a given number of hardware threads, state-of-the-art MPI implementations either achieve maximum communication throughput and waste 93.75% of hardware resources using multiple processes or achieve maximum resource efficiency and perform up to 7x worse with multiple threads. In this work, we study the tradeoff space between performance and resource usage that lies in between the two extremes. We do so by first analyzing and evaluating in depth the consequences of sharing network resources between independent threads. In the process, we extend the existing Verbs design to allow for maximally independent paths, for which case we also optimize the mlx5 stack. As a result of our analysis, we describe *scalable communication endpoints*, an efficient resource sharing model for multithreading scenarios at the lowest software level of interconnects. Each category of the model reflects a performance level and its corresponding resource usage that users, such as MPICH, can use to guide their creation of endpoints. The model’s 2xDynamic endpoints, for example, can achieve 108% of the performance of the endpoints in MPI-everywhere while using only 31.25% as many resources.

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