VIETNAM NATIONAL UNIVERSITY HO CHI MINH CITY HO CHI MINH CITY UNIVERSITY OF TECHNOLOGY FACULTY OF COMPUTER SCIENCE AND ENGINEERING



SPECIALIZED PROJECT

STUDYING AND DEVELOPING NONBLOCKING DISTRIBUTED MPSC QUEUES

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Disclaimers

I affirm that this specialized project is the product of my original research and experimentation. Any references, resources, results which this project is based on or a derivative work of have been given due citations and properly listed in the footnotes and the references section. All original contents presented are the culmination of my dedication and perserverance under the close guidance of my supervisors, Mr. Thoại Nam and Mr. Diệp Thanh Đăng, from the Faculty of Computer Science and Engineering, Ho Chi Minh City University of Technology. I take full responsibility for the accuracy and authenticity of this document. Any misinformation, copyright infrigment or plagiarism shall be faced with serious punishment.



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Chapter I Introduction

The demand for computation power has always been increasing relentlessly. Increasingly complex computation problems arise and accordingly more computation power is required to solve them. Much engineering efforts have been put forth towards obtaining more computation power. A popular topic in this regard is distributed computing: The combined power of clusters of commodity hardware can surpass that of a single powerful machine. To fully take advantage of the potential of distributed computing, specialized algorithms and data structures need to be devised.

Noticeably, multi-producer single-consumer (MPSC) is one of those data structures that are utilized heavily in distributed computing, forming the backbone of many applications. Consequently, an MPSC can easily present a performance bottleneck if not designed properly. A desirable distributed MPSC should be able to exploit the highly concurrent nature of distributed computing. Currently, in the literature, most distributed data structures are designed from the ground up, completely disregarding the any existing data structures developed in the shared memory area, e.g. [2]. This is partly due to the historical differences between the programming models utilized in these two areas. However, since the introduction of specialized networking hardware RDMA and the improved support of the remote memory access (RMA) programming model in MPI-3, this gap has been bridged. Thus, it has opened up a lot new research ([3]) on reusing the principles in the shared memory literature to distributed computing. One favorable characteristic of concurrent data structures that has been heavily researched in the shared memory literature, which is also equally important in distributed computing, is the property of non-blocking, or in particular, lock-freedom. Lock-freedom guarantees that if some processes suspend or die, other processes can still complete. This provides both progress guarantee and fault-tolerance, especially in distributed computing where nodes can fail any time. Thus, the rest of this document concerns itself with investigating and devising efficient non-blocking distributed MPSCs. Interestingly, we choose to adapt current MPSC algorithms in the shared-memory literature to distributed context, which enables a wealth of accumulated knowledge in this literature.

1.1 Motivation

Lock-free MPSC and other FIFO variants, such as multi-producer multi-consumer (MPMC), concurrent single-producer single-consumer (SPSC), are heavily studied in the shared memory literature, dating back from the 1980s-1990s [4], [5], [6] and more recently [7], [8]. It comes as no surprise that algorithms in this domain are highly developed and optimized for performance and scalability. However, most research about MPSC or FIFO algorithms in general completely disregard the available state-of-the-art algorithms in the shared memory literature. With the new RDMA networking hardware support and capabilities added to MPI-3 RMA API: lock-free shared-memory algorithms can be straightforwardly ported to distributed context using this programming model. This presents an opportunity to make use of the highly accumulated research in the shared



memory literature, which if adapted and mapped properly to the distributed context, may produce comparable results to algorithms exclusively devised within the distributed computing domain. Therefore, we decide to take this novel route to developing new non-blocking MPSC algorithms: Port and adapt potential lock-free shared-memory MSPCs to distributed context using the MPI-3 RMA programming model. If this approach proves to be effective, a huge intellectual reuse of shared-memory MSPC algorithms into the distributed domain is possible. Consequently, there may be no need to develop distributed MPSC algorithms from the ground up.

1.2 Objective

This thesis aims to:

- Investigate state-of-the-art shared-memory MPSCs.
- Select and appropriately modify potential MPSC algorithms so they can be implemented in popular distributed programming environments.
- Port MPSC algorithms using MPI-3 RMA.
- Evaluate various theoretical aspects of ported MPSC algorithms: Correctness, progress guarantee, time complexity analysis.
- Benchmark the ported MPSC algorithms and compare them with current distributed MPSCs in the literature.
- Discover distributed-environment-specific optimization opportunities for ported MPSC algorithms.

1.3 Scope

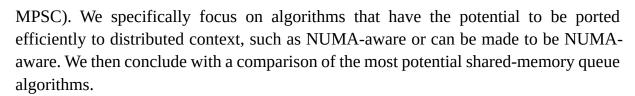
- For related works on shared-memory MPSCs, we only focus on linearizable MPSCs that support at least lock-free enqueue and dequeue operations.
- Any implementation details, benchmarking and optimizations assume MPI-3 settings.
- For optimizations, we focus on performance-related metrics, e.g. time-complexity (theoretically), throughput (empirically).

1.4 Structure

The rest of this report is structured as follows:

Chapter II discusses the theoretical foundation this thesis is based on and the technical terminology that's heavily utilized in this domain. As mentioned, this thesis investigates state-of-the-art shared-memory MPSCs. Therefore, we discuss the theory related to the design of concurrent algorithms such as lock-freedom and linearizability, the practical challenges such as the ABA problem and safe memory reclamation problem. We then explore the utilities offered by C++11 to implement concurrent algorithms and MPI-3 to port shared memory algorithms.

Chapter III surveys the shared-memory literature for state-of-the-art queue algorithms, specifically MPSC and SPSC algorithms (as SPSC can be modified to implement



Chapter IV documents distributed-versions of potential shared-memory MPSC algorithms surveys in Chapter III. It specifically presents our adaptation efforts of existing algorithms in the shared-memory literature to make their distributed implementations feasible.

Chapter V discusses various interesting theoretical aspects of our distributed MPSC algorithms in Chapter IV, specifically correctness (linearizability), progress guarantee (lock-freedom and wait-freedom), performance model. Our analysis of performance model helps back our empirical findings in Chapter VI, together, they work hand-in-hand to help us discover optimization opportunities.

Chapter VI introduces our benchmarking setup, including metrics, environments, benchmark/microbenchmark suites and conducting methods. We aim to demonstrate some preliminary results on how well ported shared-memory MPSCs can compare to existing distributed MPSCs. Finally, we discuss important factors that affect the runtime properties distributed MPSC algorithm, which have partly been explained by our theoretical analysis in Chapter V.

Chapter VII concludes what we have accomplished in this thesis and considers future possible improvements to our research.



2.1 Irregular applications

Irregular applications are a class of programs particularly interesting in distributed computing. They are characterized by:

- Unpredictable memory access: Before the program is actually run, we cannot know which data it will need to access. We can only know that at run time.
- Data-dependent control flow: The decision of what to do next (such as which data tp accessed next) is highly dependent on the values of the data already accessed. Hence the unpredictable memory access property because we cannot statically analyze the program to know which data it will access. The control flow is inherently engraved in the data, which is not known until runtime.

Irregular applications are interesting because they demand special treatments to achieve high performance. One specific challenge is that this type of applications is hard to model in traditional MPI APIs. The introduction of MPI RMA (remote memory access) in MPI-2 and its improvement in MPI-3 has significantly improved MPI's capability to express irregular applications comfortably.

2.2 Multiple-producer, single-consumer (MPSC)

Multiple-producer, single-consumer (MPSC) is a specialized concurrent first-in first-out (FIFO) data structure. A FIFO is a container data structure where items can be inserted into or taken out of, with the constraint that the items that are inserted earlier are taken out of earlier. Hence, it's also known as the queue data structure. The process that performs item insertion into the FIFO is called the producer and the process that performs items deletion (and retrieval) is called the consumer. In concurrent queues, multiple producers and consumers can run in parallel. Concurrent queues have many important applications, namely event handling, scheduling, etc. One class of concurrent FIFOs is MPSC, where one consumer may run in parallel with multiple producers. The reasons we're interested in MPSCs instead of the more general multiple-producer, multiple-consumer data structures (MPMCs) are that (1) high-performance and high-scalability MPSCs are much simpler to design than MPMCs while (2) MPSCs are noticeably as powerful as MPMCs - its consensus number equals the number of producers [9]. Thus, MPSCs can see as many use cases as MPMCs while being easily scalable and performant.

2.3 Progress guarantee

Many concurrent algorithms are based on locks to create mutual exclusion, in which only some processes that have acquired the locks are able to act, while the others have to wait. While lock-based algorithms are simple to read, write and verify, these algorithms are said to be blocking: One slow process may slow down the other faster processes, for example, if the slow process successfully acquires a lock and then the operating system (OS) decides to suspends it to schedule another one, this means until the process



is awaken, the other processes that contend for the lock cannot continue. Lock-based algorithms introduces many problems such as:

- Deadlock: There's a circular lock-wait dependencies among the processes, effectively prevent any processes from making progress.
- Convoy effect: One long process holding the lock will block other shorter processes contending for the lock.
- Priority inversion: A higher-priority process effectively has very low priority because it has to wait for another low priority process.

Furthermore, if a process that holds the lock dies, this will halt the whole program. This consideration holds even more weight in distributed computing because of a lot more failure modes, such as network failures, node falures, etc.

Therefore, while lock-based algorithms are easy to write, they do not provide **progress** guarantee because deadlock or livelock can occur and its use of mutual exclusion is unnecessarily restrictive. These algorithms are said to be **blocking**. An algorithm is said to be **non-blocking** if a failure or slow-down in one process cannot cause the failure or slow-down in another process. Lock-free and wait-free algorithms are to especially interesting subclasses of non-blocking algorithms. Unlike lock-based algorithms, they provide progress guarantee.

2.3.1 Lock-free algorithms

Lock-free algorithms provide the following guarantee: Even if some processes are suspended, the remaining processes are ensured to make global progress and complete in bounded time. This property is invaluable in distributed computing, one dead or suspended process will not block the whole program, providing fault-tolerance. Designing lock-free algorithms requires careful use of atomic instructions, such as Fetch-and-add (FAA), Compare-and-swap (CAS), etc.

2.3.2 Wait-free algorithms

Wait-freedom is a stronger progress guarantee than lock-freedom. While lock-freedom ensures that at least one of the alive processes will make progress, wait-freedom guarantees that any alive processes will finish in bounded time. Wait-freedom is useful to have because it prevents starvation. Lock-freedom still allows the possibility of one process having to wait for another indefinitely, as long as some still makes progress.

2.4 Correctness - Linearizability

Correctness of concurrent algorithms is hard to defined, especially when it comes to the semantics of concurrent data structures like MPSC. One effort to formalize the correctness of concurrent data structures is the definition of linearizability. A method call on the FIFO can be visualized as an interval spanning two points in time. The starting point is called the **invocation event** and the ending point is called the **response event**. **Linearizability** informally states that each method call should appear to take effect instantaneously at some moment between its invocation event and response event [10].



The moment the method call takes effect is termed the **linearization point**. Specifically, suppose the followings:

- We have n concurrent method calls $m_1, m_2, ..., m_n$.
- Each method call m_i starts with the **invocation event** happening at timestamp s_i and ends with the **response event** happening at timestamp e_i . We have $s_i < e_i$ for all $1 \le i \le n$.
- Each method call m_i has the **linearization point** happening at timestamp l_i , so that $s_i \leq l_i \leq e_i$.

Then, linerizability means that if we have $l_1 < l_2 < ... < l_n$, the effect of these n concurrent method calls $m_1, m_2, ..., m_n$ must be equivalent to calling $m_1, m_2, ..., m_n$ sequentially, one after the other in that order.

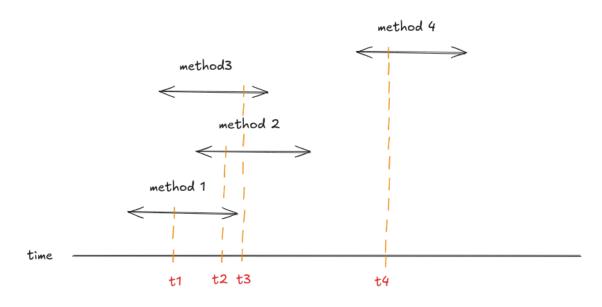


Figure 1: Linerization points of method 1, method 2, method 3, method 4 happens at $t_1 < t_2 < t_3 < t_4$, therefore, their effects will be observed in this order as if we call method 1, method 2, method 3, method 4 sequentially

2.5 Common issues when designing lock-free algorithms

2.5.1 ABA problem

In implementing concurrent lock-free algorithms, hardware atomic instructions are utilized to achieve linearizability. The most popular atomic operation instruction is compare-and-swap (CAS). The reason for its popularity is (1) CAS is a **universal atomic instruction** - it has the **concensus number** of ∞ - which means it's the most powerful atomic instruction [11] (2) CAS is implemented in most hardware (3) some concurrent lock-free data structures such as MPSC can only be implemented using powerful atomic instructions such as CAS. The semantic of CAS is as follows. Given the instruction CAS(memory location, old value, new value), atomically compares the value at memory location to see if it equals old value; if so, sets the value at memory location to



new value and returns true; otherwise, leaves the value at memory location unchanged and returns false. Concurrent algorithms often utilize CAS as follows:

- 1. Read the current value old value = read(memory location).
- 2. Compute new value from old value by manipulating some resources associated with old value and allocating new resources for new value.
- 3. Call CAS(memory location, old value, new value). If that succeeds, the new resources for new value remain valid because it was computed using valid resources associated with old value, which has not been modified since the last read. Otherwise, free up new value because old value is no longer there, so its associated resources are not valid.

This scheme is susceptible to the notorious ABA problem:

- 1. Process 1 reads the current value of memory location and reads out A.
- 2. Process 1 manipulates resources associated with A, and allocates resources based on these resources.
- 3. Process 1 suspends.
- 4. Process 2 reads the current value of memory location and reads out A.
- 5. Process 2 CAS(memory location, A, B) so that resources associated with A are no longer valid.
- 6. Process 3 CAS(memory location, B, A) and allocates new resources associated with A.
- 7. Process 1 continues and CAS(memory location, A, new value) relying on the fact that the old resources associated with A are still valid while in fact they aren't.

To safe-guard against ABA problem, one must ensure that between the time a process reads out a value from a shared memory location and the time it calls CAS on that location, there's no possibility another process has CAS the memory location to the same value. Some notable schemes are **monotonic version tag** ([6]) and **hazard pointer** ([12]).

2.5.2 Safe memory reclamation problem

The problem of safe memory reclamation often arises in concurrent algorithms that dynamically allocate memory. In such algorithms, dynamically-allocated memory must be freed at some point. However, there's a good chance that while a process is freeing memory, other processes contending for the same memory are keeping a reference to that memory. Therefore, deallocated memory can potentially be accessed, which is erroneneous. Solutions ensure that memory is only freed when no other processes are holding references to it. In garbage-collected programming environments, this problem can be conveniently push to the garbage collector. In non-garbage-collected programming environments, however, custom schemes must be utilized. Examples include using a reference counter to count the number of processes holding a reference to some memory and hazard pointer [12] to announce to other processes that some memory is not to be freed.

2.6 MPI-3

MPI stands for message passing interface, which is a **message-passing library interface specification**. Design goals of MPI includes high availability across platforms, efficient communication, thread-safety, reliable and convenient communication interface while still allowing hardware-specific accelerated mechanisms to be exploited [1].

2.6.1 MPI-3 RMA

RMA in MPI RMA stands for remote memory access. As introduced in the first section of Section Chapter II, RMA APIs is introduced in MPI-2 and its capabilities are further extended in MPI-3 to conveniently express irregular applications. In general, RMA is intended to support applications with dynamically changing data access patterns where the data distribution is fixed or slowly changing [1]. In such applications, one process, based on the data it needs, knowing the data distribution, can compute the nodes where the data is stored. However, because data acess pattern is not known, each process cannot know whether any other processes will access its data.

Using the traditional Send/Receive interface, both sides need to issue matching operations by distributing appropriate transfer parameters. This is not suitable, as previously explain, only the side that needs to access the data knows all the transfer parameters while the side that stores the data cannot anticipate this.

2.6.2 MPI-RMA communication operations

RMA only requires one side to specify all the transfer parameters and thus only that side to participate in data communication.

To utilize MPI RMA, each process needs to open a memory window to expose a segment of its memory to RMA communication operations such as remote writes (MPI_PUT), remote reads (MPI_GET) or remote accumulates (MPI_ACCUMULATE, MPI_GET_ACCUMULATE, MPI_FETCH_AND_OP, MPI_COMPARE_AND_SWAP) [1]. These remote communication operations only requires one side to specify.

2.6.3 MPI-RMA synchronization

Besides communication of data from the sender to the receiver, one also needs to synchronize the sender with the receiver. That is, there must be a mechanism to ensure the completion of RMA communication calls or that any remote operations have taken effect. For this purpose, MPI RMA provides **active target synchronization** and **passive target synchronization**. In this document, we're particularly interested in **passive target synchronization** as this mode of synchronization does not require the target process of an RMA operation to explicitly issue a matching synchronization call with the origin process, easing the expression of irregular applications [13].

In **passive target synchronization**, any RMA communication calls must be within a pair of MPI_Win_lock/MPI_Win_unlock or MPI_Win_lock_all/MPI_Win_unlock_all. After the unlock call, those RMA communication calls are guaranteed to have taken effect.



One can also force the completion of those RMA communication calls without the need for the call to unlock using flush calls such as MPI_Win_flush or MPI_Win_flush_local.

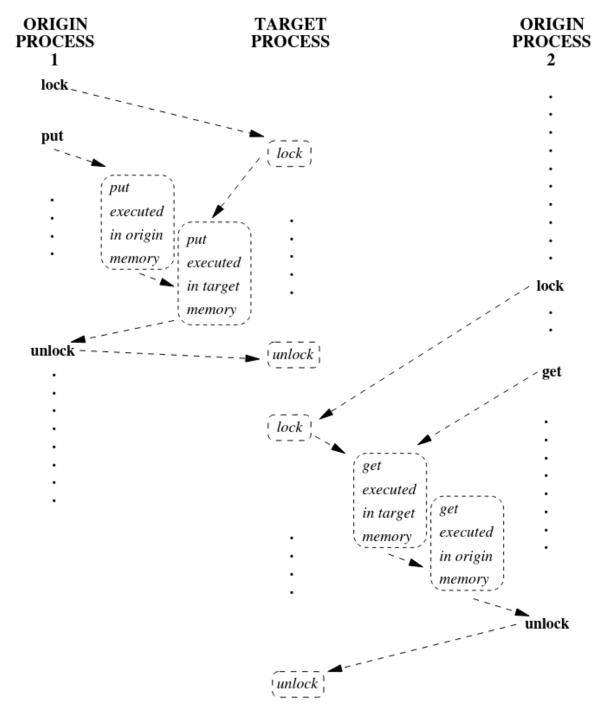


Figure 2: An illustration of passive target communication. Dashed arrows represent synchronization (source: [1])

2.7 Pure MPI approach of porting shared memory algorithms

In pure MPI, we use MPI exclusively for communication and synchronization. With MPI RMA, the communication calls that we utilize are:

• Remote read: MPI_Get

• Remote write: MPI_Put



• Remote accumulation: MPI_Accumulate, MPI_Get_accumulate, MPI_Fetch_and_op and MPI_Compare_and_swap.

For lock-free synchronization, we choose to use passive target synchronization with MPI_Win_lock_all/MPI_Win_unlock_all.

In the MPI-3 specification [1], these functions are specified as follows:

Operation	Usage	
MPI_Win_lock_all	Starts and RMA access epoch to all processes in a memory	
	window, with a lock type of MPI_LOCK_SHARED. The calling	
	process can access the window memory on all processes in	
	the memory window using RMA operations. This routine is	
	not collective.	
MPI_Win_unlock_all	Matches with an MPI_Win_lock_all to unlock a window	
	previously locked by that MPI_Win_lock_all.	

Table 1: Specification of MPI_Win_lock_all and MPI_Win_unlock_all

The reason we choose this is 3-fold:

- Unlike active target synchronization, passive target synchronization does not require the process whose memory is being accessed by an MPI RMA communication call to participate in. This is in line with our intention to use MPI RMA to easily model irregular applications like MPSCs.
- Unlike active target synchronization, MPI_Win_lock_all and MPI_Win_unlock_all do not need to wait for a matching synchronization call in the target process, and thus, is not delayed by the target process.
- Unlike passive target synchronization with MPI Win lock/MPI Win unlock, multiple calls of MPI_Win_lock_all can succeed concurrently, so one process needing to issue MPI RMA communication calls do not block others.

An example of our pure MPI approach with MPI_Win_lock_all/MPI_Win_unlock_all, inspired by [13], is illustrated in the following:



```
MPI_Win_lock_all(0, win);
MPI_Get(...); // Remote get
MPI_Put(...); // Remote put
MPI_Accumulate(..., MPI_REPLACE, ...); // Atomic put
MPI_Get_accumulate(..., MPI_NO_OP, ...); // Atomic get
MPI_Fetch_and_op(...); // Remote fetch-and-op
MPI_Compare_and_swap(...); // Remote compare and swap
MPI_Win_flush(...); // Make previous RMA operations take effects
MPI_Win_flush_local(...); // Make previous RMA operations take
effects locally
MPI_Win_unlock_all(win);
```

Listing 1: An example snippet showcasing our synchronization approach in MPI RMA



Figure 3: An illustration of our synchronization approach in MPI RMA



Chapter III Related works

There exists numerous research into the design of lock-free shared memory MPMCs and SPSCs. Interestingly, research into lock-free MPSCs are noticeably scarce. Although in principle, MPMCs and SPSCs can both be adapted for MPSCs use cases, specialized MPSCs can usually yield much more performance. In reality, we have only found 4 papers that are concerned with direct support of lock-free MPSCs: LTQueue [7], DQueue [9], WRLQueue [14] and Jiffy [8]. Table 2 summarizes the charateristics of these algorithms.

MPSCs	LTQueue	DQueue	WR-	Jiffy
			LQueue	
ABA solution	Load-link/	Incorrect	Custom	Custom
	Store-con-	custom	scheme	scheme
	ditional	scheme (*)		
Memory reclamation	Custom	Incorrect	Custom	Custom
	scheme	custom	scheme	scheme
		scheme (*)		
Progress guarantee of dequeue	Wait-free	Wait-free	Blocking	Wait-free
			(*)	
Progress guarantee of enqueue	Wait-free	Wait-free	Wait-free	Wait-free
Number of elements	Un-	Un-	Un-	Un-
	bounded	bounded	bounded	bounded

Table 2: Characteristic summary of existing shared memory MPSCs. The cell marked with (*) indicates that our evaluation contradicts with the author's claims

LTQueue [7] is the earliest wait-free shared memory MPSC to our knowledge. This algorithm is wait-free with $O(\log n)$ time complexity for both enqueues and dequeues, with n being the number of enqueuers. Their main idea is to split the MPSC among the enqueuers so that each enqueuer maintains a local SPSC data structure, which is only shared with the dequeuer. This improves the MPSC's scalability as multiple enqueues can complete the same time. The enqueuers shared a distributed counter and use it to label each item in their local SPSC with a specific timestamp. The timestamps are organized into nodes of a min-heap-like tree so that the dequeuer can look at the root of tree to determine which local SPSC to dequeue next. The min-heap property of the tree is preserved by a novel wait-free timestamp-refreshing operation. Memory reclamation becomes trivial as each MPSC entry is only shared by one enqueuer and one dequeuer in the local SPSC. The algorithm avoids ABA problem by utilizing load-link/storeconditional (LL/SC). This, on the other hand, presents a challenge in directly porting LTQueue as LL/SC is not widely available as the more popular CAS instruction.

DQueue [9] focuses on optimizing performance. It aims to be cache-friendly by having each enqueuer batches their updates in a local buffer to decrease cache misses. It also try to replace expensive atomic instructions such as CAS as many as possible. The MPSC is represented as a linked list of segments (which is an array). To enqueue, the enqueuer reserves a slot in the segment list and enqueues the value into the local buffer. If the local buffer is full, the enqueuer flushes the buffer and writes it onto every reserved slot in the segment list. The producer dequeues the values in the segment list in order, upon encountering a reserved but empty slot, it helps all enqueuers flush their local buffers. For memory reclamation, DQueue utilized a dedicated garbage collection thread that reclaims all fully dequeued segments. However, their algorithm is flawed and a segment maybe freed while some process is holding a reference to it.

WRLQueue [14] is a lock-free MPSC for embedded real-time system. Its main purpose is to avoid excessive modification of storage space. WRLQueue is simplfy a pair of buffer, one is worked on by multiple enqueuers and the other is work on by the dequeuer. The enqueuers batch their enqueues and write multiple elements onto the buffer once at a time. The dequeuer upon invocation will swap its buffer with the enqueuer's buffers to dequeue from it. However, this requires the dequeuer to wait for all enqueue operations to complete in their buffer. If an enqueue suspends or dies, the dequeuer will have to wait forever, this clearly violates the property of non-blocking.

Jiffy [8] is a fast and memory-efficient wait-free MPSC by avoiding excessive allocation of memory. Like DQueue, Jiffy represents the queue as a linked list of segments. Each enqueue reserves a slot in the segment, extends the linked-list as appropriately, writes the value into the slot and sets a per-slot flag to indicate that the slot is ready to be dequeued. To dequeue, the dequeuer repeatedly scan all the slots to find the first-ready-to-bedequeue slot. Jiffy shows significant good memory usage and throughput compared to other previous state-of-the-art MPMCs.



Chapter IV Distributed MPSCs

Based on the MPSC algorithms we have surveyed in Chapter III, we propose two waitfree distributed MPSC algorithms:

- LTQueuev1 (Section 4.3) is a direct modification of LTQueue [7] without any usage of LL/SC.
- LTQueueV2 (Section 4.4) is inspired by the timestamp-refreshing idea of LTQueue [7] and repeated-rescan of Jiffy [8]. Although it still bears some resemblance to LTQueue, we believe it to be more optimized for distributed context.

MPSC	LTQueueV1	LTQueueV2
Correctness	Linearizable	Linearizable
Progress guarantee of dequeue	Wait-free	Wait-free
Progress guarantee of enqueue	Wait-free	Wait-free
Worst-case time complexity of dequeue	$O(\log n) R + O(\log n) A$	constant R + $O(n)$ A
Worst-case time complexity of enqueue	$O(\log n) R + O(\log n) A$	constant R + constant A
ABA solution	Unique timestamp	No harmful ABA problem
Memory reclamation	Custom scheme	Custom scheme
Number of elements	Unbounded	Unbounded

Table 3: Characteristic summary of our proposed distributed MPSCs. *n* is the number of enqueuers, R stands for **remote operation** and A stands for **atomic operation**

In this section, we present our proposed distributed MPSCs in detail. Any other discussions about theoretical aspects of these algorithms such as linearizability, progress guarantee, time complexity are deferred to Chapter V.

In our description, we assume that each process in our program is assigned a unique number as an identifier, which is termed as its **rank**. The numbers are taken from the range of [0, size - 1], with size being the number of processes in our program.

4.1 Distributed primitives in pseudocode

Although we use MPI-3 RMA to implement these algorithms, the algorithm specifications themselves are not inherently tied to MPI-3 RMA interfaces. For clarity and convenience in specification, we define the following distributed primitives used in our pseudocode.

remote<T>: A distributed shared variable of type T. The process that physically stores the variable in its local memory is referred to as the **host**. This represents data that can be accessed or modified remotely by other processes.



void aread_sync(remote<T> src, T* dest): Issue a synchronous read of the distributed variable src and stores its value into the local memory location pointed to by dest. The read is guaranteed to be completed when the function returns.

void aread_sync(remote<T*> src, int index, T* dest): Issue a synchronous read of the element at position index within the distributed array src (where src is a pointer to a remotely hosted array of type T) and stores the value into the local memory location pointed to by dest. The read is guaranteed to be completed when the function returns.

void awrite_sync(remote<T> dest, T* src): Issue a synchronous write of the value at the local memory location pointed to by src into the distributed variable dest. The write is guaranteed to be completed when the function returns.

void awrite_sync(remote<T*> dest, int index, T* src): Issue a synchronous write of the value at the local memory location pointed to by src into the element at position index within the distributed array dest (where dest is a pointer to a remotely hosted array of type T). The write is guaranteed to be completed when the function returns.

void aread_async(remote<T> src, T* dest): Issue an asynchronous read of the distributed variable src and initiate the transfer of its value into the local memory location pointed to by dest. The operation may not be completed when the function returns.

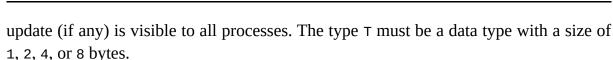
void aread_async(remote<T*> src, int index, T* dest): Issue an asynchronous read of the element at position index within the distributed array src (where src is a pointer to a remotely hosted array of type T) and initiate the transfer of its value into the local memory location pointed to by dest. The operation may not be completed when the function returns.

void awrite_async(remote<T> dest, T* src): Issue an asynchronous write of the value at the local memory location pointed to by src into the distributed variable dest. The operation may not be completed when the function returns.

void awrite_async(remote<T*> dest, int index, T* src): Issue an asynchronous write of the value at the local memory location pointed to by src into the element at position index within the distributed array dest (where dest is a pointer to a remotely hosted array of type T). The operation may not be completed when the function returns.

void flush(remote<T> src): Ensure that all read and write operations on the distributed variable src (or its associated array) issued before this function call are fully completed by the time the function returns.

bool compare_and_swap_sync(remote<T> dest, T old_value, T new_value): Issue a synchronous compare-and-swap operation on the distributed variable dest. The operation atomically compares the current value of dest with old_value. If they are equal, the value of dest is replaced with new_value; otherwise, no change is made. The operation is guaranteed to be completed when the function returns, ensuring that the



bool compare_and_swap_sync(remote<T*> dest, int index, T old_value, T new_value): Issue a synchronous compare-and-swap operation on the element at position index within the distributed array dest (where dest is a pointer to a remotely hosted array of type T). The operation atomically compares the current value of the element at dest[index] with old_value. If they are equal, the element at dest[index] is replaced with new_value; otherwise, no change is made. The operation is guaranteed to be completed when the function returns, ensuring that the update (if any) is visible to all processes. The type T must be a data type with a size of 1, 2, 4, or 8.

T fetch_and_add_sync(remote<T> dest, T inc): Issue a synchronous fetch-and-add operation on the distributed variable dest. The operation atomically adds the value inc to the current value of dest, returning the original value of dest (before the addition) to the calling process. The update to dest is guaranteed to be completed and visible to all processes when the function returns. The type T must be an integral type with a size of 1, 2, 4, or 8 bytes.

4.2 A simple distributed SPSC

The two algorithms we propose here both utilize a distributed SPSC data structure, which we will present first. For implementation simplicity, we present a bounded SPSC, effectively make our proposed algorithms support only a bounded number of elements. However, one can trivially substitute another distributed unbounded SPSC to make our proposed algorithms support an unbounded number of elements, as long as this SPSC supports the same interface as ours.

Placement-wise, all shared data in this SPSC is hosted on the enqueuer.

Types

| data_t = The type of data stored

Shared variables

First: remote<uint64_t>

The index of the last undequeued entry. Hosted at the enqueuer.

Last: remote<uint64 t>

The index of the last unenqueued entry. Hosted at the enqueuer.

Data: remote<data_t*>

An array of data_t of some known capacity. Hosted at the enqueuer.

Enqueuer-local variables

Capacity: A read-only value indicating

the capacity of the SPSC

First_buf: The cached value of First

Last_buf: The cached value of Last

Dequeuer-local variables

Capacity: A read-only value indicating

the capacity of the SPSC

First_buf: The cached value of First

Last_buf: The cached value of Last

Enqueuer initialization

Dequeuer initialization

```
Initialize First and Last to 0
Initialize Capacity
Allocate array in Data
First_buf = Last_buf = 0
```

```
| Initialize Capacity
| First_buf = Last_buf = 0
```

The procedures of the enqueuer are given as follows.

Procedure 2: bool spsc_enqueue(data_t v)

spsc_enqueue first computes the new Last value (line 1). If the queue is full as indicating by the difference the new Last value and First-buf (line 2), there can still be the possibility that some elements have been dequeued but First-buf hasn't been synced with First yet, therefore, we first refresh the value of First-buf by fetching from First (line 3). If the queue is still full (line 4), we signal failure (line 5). Otherwise, we proceed to write the enqueued value to the entry at Last_buf % Capacity (line 6), increment Last (line 7), update the value of Last_buf (line 8) and signal success (line 9).

Procedure 3: bool spsc_readFront_e(data_t* output)

```
if (First_buf >= Last_buf)

return false

read_sync(First, &First_buf)

if (First_buf >= Last_buf)

return false

return false

read_sync(Data, First_buf % Capacity, output)

return true
```

spsc_readFront_e first checks if the SPSC is empty based on the difference between First_buf and Last_buf (line 10). Note that if this check fails, we signal failure immediately (line 11) without refetching either First or Last. This suffices because



Last cannot be out-of-sync with Last_buf as we're the enqueuer and First can only increase since the last refresh of First_buf, therefore, if we refresh First and Last, the condition on line 10 would return false anyways. If the SPSC is not empty, we refresh First and re-perform the empty check (line 12-14). If the SPSC is again not empty, we read the queue entry at First_buf % Capacity into output (line 15) and signal success (line 16).

The procedures of the dequeuer are given as follows.

Procedure 4: bool spsc_dequeue(data_t* output)

spsc_dequeue first computes the new First value (line 15). If the queue is empty as indicating by the difference the new First value and Last-buf (line 16), there can still be the possibility that some elements have been enqueued but Last-buf hasn't been synced with Last yet, therefore, we first refresh the value of Last-buf by fetching from Last (line 17). If the queue is still empty (line 18), we signal failure (line 19). Otherwise, we proceed to read the top value at First_buf % Capacity (line 20) into output, increment First (line 21) - effectively dequeue the element, update the value of First_buf (line 22) and signal success (line 23).

Procedure 5: bool spsc_readFrontd(data_t* output)

```
24 if (First_buf >= Last_buf)
25  | aread_sync(Last, &Last_buf)
26  | if (First_buf >= Last_buf)
27  | return false
28 aread_sync(Data, First_buf % Capacity, output)
29 return true
```

spsc_readFront_d first checks if the SPSC is empty based on the difference between First_buf and Last_buf (line 24). If this check fails, we refresh Last_buf (line 25) and recheck (line 26). If the recheck fails, signal failure (line 27). If the SPSC is not empty,



we read the queue entry at First_buf % Capacity into output (line 28) and signal success (line 29).

4.3 LTQueueV1 - Modified LTQueue without LL/SC

This algorithm presents our most straightforward effort to port LTQueue [7] to distributed context. The main challenge is that LTQueue uses LL/SC as the universal atomic instruction and also an ABA solution, but LL/SC is not available in distributed programming environments. We have to replace any usage of LL/SC in the original LTQueue algorithm. Compare-and-swap is unavoidable in distributed MPSCs, so we use the wellknown monotonic timestamp scheme to guard against ABA problem.

4.3.1 Structure

The structure of our modified LTQueue is shown as in Image 1.

We differentiate between 2 types of nodes: **enqueuer nodes** (represented as the rectangular boxes at the bottom of Image 1) and normal tree nodes (represented as the circular boxes in Image 1).

Each enqueuer node corresponds to an enqueuer. Each time the local SPSC is enqueued with a value, the enqueuer timestamps the value using a distributed counter shared by all enqueuers. An enqueuer node stores the SPSC local to the corresponding enqueuer and a min_timestamp value which is the minimum timestamp inside the local SPSC.

Each tree node stores the rank of an enqueuer process. This rank corresponds to the enqueuer node with the minimum timestamp among the node's children's ranks. The tree node that's attached to an enqueuer node is called a leaf node, otherwise, it's called an internal node.

Note that if a local SPSC is empty, the min_timestamp variable of the corresponding enqueuer node is set to MAX and the corresponding leaf node's rank is set to a DUMMY rank.

Placement-wise:

- The **enqueuer nodes** are hosted at the corresponding **enqueuer**.
- All the **tree nodes** are hosted at the **dequeuer**.
- The distributed counter, which the enqueuers use to timestamp their enqueued value, is hosted at the **dequeuer**.

4.3.2 Pseudocode

Below is the types utilized in LTQueueV1.

Types

data_t = The type of the data to be stored spsc_t = The type of the SPSC, this is assumed to be the distributed SPSC in Section 4.2



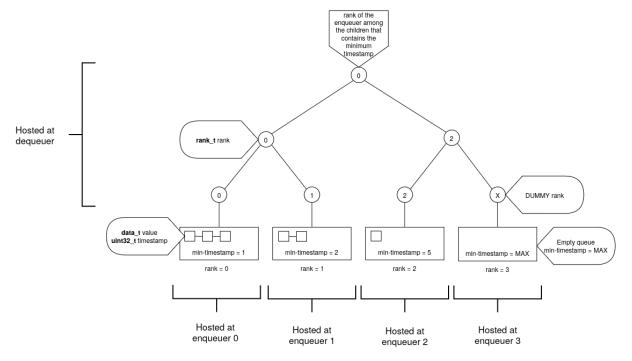


Image 1: Modified LTQueue's structure

rank_t = The type of the rank of an enqueuer process tagged with a unique timestamp (version) to avoid ABA problem

The shared variables in our LTQueue version are as followed.

Note that we have described a very specific and simple way to organize the tree nodes in LTQueue in a min-heap-like array structure hosted on the sole dequeuer. We will resume our description of the related tree-structure procedures parent() (Procedure 7), children() (Procedure 8), leafNodeIndex() (Procedure 9) with this representation in mind. However, our algorithm doesn't strictly require this representation and can be



subtituted with other more-optimized representations & distributed placements, as long as the similar tree-structure procedures are supported.

Shared variables

Counter: remote<uint64_t>

A distributed counter shared by the enqueuers. Hosted at the dequeuer.

Tree_size: uint64_t

A read-only variable storing the number of tree nodes present in the LTQueue.

Nodes: remote<node_t>

An array with Tree_size entries storing all the tree nodes present in the LTQueue shared by all processes.

Hosted at the dequeuer.

This array is organized in a similar manner as a min-heap: At index 0 is the root node. For every index i > 0, $\left\lfloor \frac{i-1}{2} \right\rfloor$ is the index of the parent of node i. For every index i > 0, 2i + 1 and 2i + 2 are the indices of the children of node i.

Dequeuer_rank: uint32_t

The rank of the dequeuer process. This is read-only.

Timestamps: A read-only array [0..size - 2] of remote<timestamp_t>, with size being the number of processes.

The entry at index i corresponds to the Min_timestamp distributed variable at the enqueuer with an order of i.

Similar to the fact that each process in our program is assigned a rank, each enqueuer process in our program is assigned an **order**. The following procedure computes an enqueuer's order based on its rank:

Procedure 6: uint32_t enqueuerOrder(uint32_t enqueuer_rank)

1 return enqueuer_rank > Dequeuer_rank ? enqueuer_rank - 1 : enqueuer_rank

This procedure is rather straightforward: Each enqueuer is assigned an order in the range [0, size - 2], with size being the number of processes and the total ordering among the enqueuers based on their ranks is the same as the total ordering among the enqueuers based on their orders.

Enqueuer-local variables

Enqueuer_count: uint64_t

The number of enqueuers.

Self rank: uint32 t

The rank of the current enqueuer process.

Min_timestamp:

remote<timestamp_t>

Spsc: spsc_t

This SPSC is synchronized with the

dequeuer.

Dequeuer-local variables

Enqueuer_count: uint64_t

The number of enqueuers.

Spscs: **array** of spsc_t with Enqueuer_count entries.

The entry at index i corresponds to the Spsc at the enqueuer with an order of i.

Enqueuer initialization

Initialize Enqueuer_count, Self_rank and Dequeuer_rank.

Initialize Spsc to the initial state.

Initialize Min_timestamp to timestamp_t {MAX_TIMESTAMP, 0}.

Dequeuer initialization

Initialize Enqueuer_count, Self_rank and Dequeuer_rank.

Initialize Counter to 0.

Initialize Tree_size to
Enqueuer_count * 2.

Initialize Nodes to an array with Tree_size entries. Each entry is initialized to node_t {DUMMY_RANK}.

Initialize Spscs, synchronizing each entry with the corresponding enqueuer.

Initialize Timestamps, synchronizing each entry with the corresponding enqueuer.

We first present the tree-structure utility procedures that are shared by both the enqueuer and the dequeuer:

Procedure 7: uint32_t parent(uint32_t index)

2 **return** (index - 1) / 2

parent returns the index of the parent tree node given the node with index index. These indices are based on the shared Nodes array. Based on how we organize the Nodes array, the index of the parent tree node of index is (index - 1) / 2.



Procedure 8: vector<uint32_t> children(uint32_t index)

```
3 left_child = index * 2 + 1
4 right_child = left_child + 1
5 res = vector<uint32_t>()
6 if (left_child >= Tree_size)
7 | return res
8 res.push(left_child)
9 if (right_child >= Tree_size)
10 | return res
11 res.push(right_child)
12 return res
```

Similarly, children returns all indices of the child tree nodes given the node with index index. These indices are based on the shared Nodes array. Based on how we organize the Nodes array, these indices can be either index * 2 + 1 or index * 2 + 2.

```
Procedure 9: uint32_t leafNodeIndex(uint32_t enqueuer_rank)
```

```
13 return Tree_size + enqueuerOrder(enqueuer_rank)
```

leafNodeIndex returns the index of the leaf node that's logically attached to the enqueuer node with rank enqueuer_rank as in Image 1.

The followings are the enqueuer procedures.

Procedure 10: bool enqueue(data_t value)

```
timestamp
fetch_and_add_sync(Counter, 1)

spsc_enqueue(&Spsc, (value, timestamp))

propagate<sub>e</sub>()
```

To enqueue a value, enqueue first obtains a count by FAA the distributed counter Counter (line 14). Then, we enqueue the data tagged with the timestamp into the local SPSC (line 15). Finally, enqueue propagates the changes by invoking propagate_e() (line 16).



Procedure 11: void propagate_e()

```
if (!refreshTimestampe())

refreshTimestampe()

if (!refreshLeafe())

refreshLeafe()

current_node_index = leafNodeIndex(Self_rank)

repeat

current_node_index = parent(current_node_index)

if (!refreshe(current_node_index))

refreshe(current_node_index)

until current_node_index == 0
```

The propagate_e procedure is responsible for propagating SPSC updates up to the root node as a way to notify other processes of the newly enqueued item. It is split into 3 phases: Refreshing of Min_timestamp in the enqueuer node (line 18-19), refreshing of the enqueuer's leaf node (line 20-21), refreshing of internal nodes (line 23-27). On line 20-27, we refresh every tree node that lies between the enqueuer node and the root node.

Procedure 12: bool refreshTimestamp_e()

The refreshTimestamp_e procedure is responsible for updating the Min_timestamp of the enqueuer node. It simply looks at the front of the local SPSC (line 310 and CAS Min_timestamp accordingly (line 33-36).



Procedure 13: bool refreshNode_e(uint32_t current_node_index)

```
37 current_node = node_t {}
38 aread_sync(Nodes, current_node_index, &current_node)
39 {old-rank, old-version} = current_node.rank
40 min_rank = DUMMY_RANK
41 min_timestamp = MAX_TIMESTAMP
42 for child_node_index in children(current_node)
    child_node = node_t {}
43
44
     aread_sync(Nodes, child_node_index, &child_node)
     {child_rank, child_version} = child_node
45
    if (child_rank == DUMMY_RANK) continue
46
    child_timestamp = timestamp_t {}
47
    aread_sync(Timestamps[enqueuerOrder(child_rank)], &child_timestamp)
48
49
    if (child_timestamp < min_timestamp)</pre>
       min_timestamp = child_timestamp
50
      min rank = child rank
51
  return compare_and_swap_sync(Nodes, current_node_index,
52 node_t {rank_t {old_rank, old_version}},
  node_t {rank_t {min_rank, old_version + 1}})
```

The refreshNode_e procedure is responsible for updating the ranks of the internal nodes affected by the enqueue. It loops over the children of the current internal nodes (line 42). For each child node, we read the rank stored in it (line 44), if the rank is not DUMMY_RANK, we proceed to read the value of Min_timestamp of the enqueuer node with the corresponding rank (line 48). At the end of the loop, we obtain the rank stored inside one of the child nodes that has the minimum timestamp stored in its enqueuer node (line 50-51). We then try to CAS the rank inside the current internal node to this rank.

Procedure 14: bool refreshLeaf_e()

```
1 leaf_node_index = leafNodeIndex(Self_rank)
1 leaf_node = node_t {}
1 saread_sync(Nodes, leaf_node_index, &leaf_node)
1 saread_sync(Nodes, leaf_node.rank)
2 saread_sync(Min_version) = leaf_node.rank
3 saread_sync(Min_timestamp_t {}
3 saread_sync(Min_timestamp, &min_timestamp)
4 saread_sync(Min_timestamp.timestamp)
5 saread_sync(Min_timestamp.timestamp)
5 saread_sync(Min_timestamp, &min_timestamp)
6 saread_sync(Min_timestamp.timestamp)
6 saread_sync(Min_timestamp, &min_timestamp)
7 saread_sync(Min_timestamp, &min_timestamp)
8 saread_sync(Min_timestam
```



The refreshLeafe procedure is responsible for updating the rank of the leaf node affected by the enqueue. It simply reads the value of Min_timestamp of the enqueuer node it's logically attached to (line 58) and CAS the leaf node's rank accordingly (line 60).

The followings are the dequeuer procedures.

Procedure 15: bool dequeue(data_t* output)

```
61 root_node = node_t {}
62 aread_sync(Nodes, 0, &root_node)
63 {rank, version} = root_node.rank
64 if (rank == DUMMY_RANK) return false
65 output_with_timestamp = (data_t {}, timestamp_t {})
  {f if} (!spsc_dequeue(&Spscs[enqueuerOrder(rank)]),
       &output_with_timestamp))
   return false
68 *output = output_with_timestamp.data
69 propagated (rank)
70 return true
```

To dequeue a value, dequeue reads the rank stored inside the root node (line 62). If the rank is DUMMY_RANK, the MPSC is treated as empty and failure is signaled (line 64). Otherwise, we invoke spsc_dequeue on the SPSC of the enqueuer with the obtained rank (line 66). We then extract out the real data and set it to output (line 68). We finally propagate the dequeue from the enqueuer node that corresponds to the obtained rank (line 69) and signal success (line 70).

Procedure 16: void propagate_d(uint32_t enqueuer_rank)

```
71 if (!refreshTimestamp<sub>d</sub>(enqueuer_rank))
72 | refreshTimestampd(enqueuer_rank)
73 if (!refreshLeaf<sub>d</sub>(enqueuer_rank))
74 ∣ refreshLeafd(enqueuer_rank)
75 current_node_index = leafNodeIndex(enqueuer_rank)
76 repeat
     current_node_index = parent(current_node_index)
77
78
     if (!refresh<sub>d</sub>(current_node_index))
     | refresh<sub>d</sub>(current_node_index)
80 until current_node_index == 0
```

The propagated procedure is similar to propagate, with appropriate changes to accommodate the dequeuer.



Procedure 17: bool refreshTimestamp_d(uint32_t enqueuer_rank)

```
81 enqueuer_order = enqueuerOrder(enqueuer_rank)
82 min_timestamp = timestamp_t {}
83 aread_sync(Timestamps, enqueuer_order, &min_timestamp)
84 {old-timestamp, old-version} = min_timestamp
85 front = (data_t {}, timestamp_t {})
86 is_empty = spsc_readFront(&Spscs[enqueuer_order], &front)
87 if (is_empty)
    return compare_and_swap_sync(Timestamps, enqueuer_order,
88
    timestamp_t {old-timestamp, old-version},
    timestamp_t {MAX_TIMESTAMP, old-version + 1})
89 else
    return compare_and_swap_sync(Timestamps, enqueuer_order,
90
    timestamp_t {old-timestamp, old-version},
   timestamp_t {front.timestamp, old-version + 1})
```

The refreshTimestamp_d procedure is similar to refreshTimestamp_e, with appropriate changes to accommodate the dequeuer.

Procedure 18: bool refreshNoded(uint32_t current_node_index)

```
91 current_node = node_t {}
92 aread_sync(Nodes, current_node_index, &current_node)
93 {old-rank, old-version} = current_node.rank
94 min_rank = DUMMY_RANK
95 min_timestamp = MAX_TIMESTAMP
96 for child_node_index in children(current_node)
     child_node = node_t {}
97
98
     aread_sync(Nodes, child_node_index, &child_node)
     {child_rank, child_version} = child_node
99
100
     if (child_rank == DUMMY_RANK) continue
     child_timestamp = timestamp_t {}
101
102
     aread_sync(Timestamps[enqueuerOrder(child_rank)], &child_timestamp)
103
     if (child_timestamp < min_timestamp)</pre>
104
        min_timestamp = child_timestamp
105
       min_rank = child_rank
   return compare_and_swap_sync(Nodes, current_node_index,
106 node_t {rank_t {old_rank, old_version}},
   node_t {rank_t {min_rank, old_version + 1}})
```

The refreshNode_d procedure is similar to refreshNode_e, with appropriate changes to accommodate the dequeuer.

Procedure 19: bool refreshLeaf_d(uint32_t enqueuer_rank)

```
107 leaf_node_index = leafNodeIndex(enqueuer_rank)
108 leaf_node = node_t {}
109 aread_sync(Nodes, leaf_node_index, &leaf_node)
110 {old_rank, old_version} = leaf_node.rank
111 min_timestamp = timestamp_t {}
112 aread_sync(Timestamps, enqueuerOrder(enqueuer_rank), &min_timestamp)
113 timestamp = min_timestamp.timestamp
    return compare_and_swap_sync(Nodes, leaf_node_index,
114 node_t {rank_t {old-rank, old-version}},
115 node_t {timestamp == MAX ? DUMMY_RANK : Self_rank, old_version + 1})
```

The refreshLeaf_d procedure is similar to refreshLeaf_e, with appropriate changes to accommodate the dequeuer.

4.4 LTQueueV2 - Optimized LTQueue for distributed context

4.4.1 Motivation

Even though the straightforward LTQueue algorithm we have ported in Section 4.3 pretty much preserve the original algorithm's characteristics, that is wait-freedom and time complexity of $\Theta(\log n)$ for both enqueue and dequeue operations (which we will prove in Chapter V), we have to be aware that this is $\Theta(\log n)$ remote operations, which is potentially expensive and a bottleneck in the algorithm.

Therefore, to be more suitable for distributed context, we propose a new algorithm that's inspired by LTQueue, in which both enqueue and dequeue only perform a constant number of remote operations, at the cost of dequeue having to perform $\Theta(n)$ local operations, where n is the number of enqueuers. Because remote operations are much more expensive, this might be a worthy tradeoff.

4.4.2 Structure

The structure of LTQueueV2 is shown as in Figure 4.

Each enqueuer hosts a distributed SPSC as in LTQueueV1 (Section 4.3). The enqueuer when enqueues a value to its local SPSC will timestamp the value using a distributed counter hosted at the dequeuer.

Additionally, the dequeuer hosts an array whose entries each corresponds with an enqueuer. Each entry stores the minimum timestamp of the local SPSC of the corresponding enqueuer.



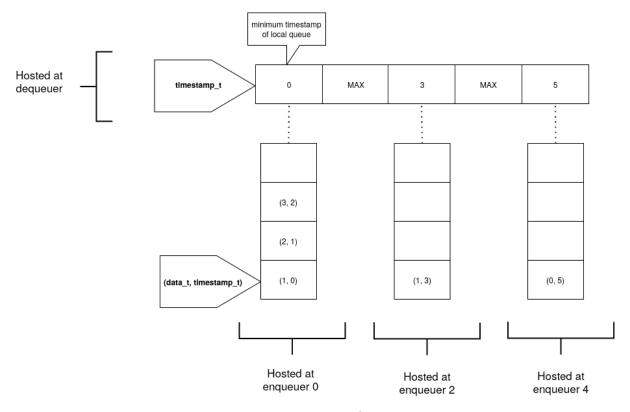


Figure 4: Basic structure of LTQueueV2

4.4.3 Pseudocode

We first introduce the types and shared variables utilized in LTQueueV2.

Types

data_t = The type of data stored $timestamp_t = uint64_t$ spsc_t = The type of the SPSC each enqueuer uses, this is assumed to be the distributed SPSC in Section 4.2

Shared variables

Slots: remote<timestamp_t*>

An array of timestamp_t with the number of entries equal to the number of enqueuers.

Hosted at the dequeuer.

Counter: remote<uint64_t>

A distributed counter.

Hosted at the dequeuer.

Dequeuer_rank: uint32_t

The rank of the dequeuer process. This is read-only.



Similar to the idea of assigning an order to each enqueuer in LTQueueV1, the following procedure computes an enqueuer's order based on its rank:

```
Procedure 20: uint64_t enqueuerOrder(uint64_t enqueuer_rank)
```

```
1 return enqueuer_rank > Dequeuer_rank ? enqueuer_rank - 1 : enqueuer_rank
```

Again, each enqueuer is assigned an order in the range [0, size - 2], with size being the number of processes and the total ordering among the enqueuers based on their ranks is the same as the total ordering among the enqueuers based on their orders.

Reversely, enqueuerRank computes an enqueuer's rank given its order.

```
Procedure 21: uint64_t enqueuerRank(uint64_t enqueuer_order)
```

```
{\color{red} return enqueuer\_order}
                                 Dequeuer_rank ?
                                                       enqueuer_order +
                                                                              1
engueuer order
```

Enqueuer-local variables

Dequeuer_rank: uint64_t Enqueuer_count: uint64_t The number of enqueuers. Self_rank: uint32_t The rank of the current enqueuer process. Spsc: spsc_t This SPSC is synchronized with the dequeuer.

Dequeuer-local variables

order of i.

```
Dequeuer_rank: uint64_t
Enqueuer_count: uint64_t
The number of enqueuers.
                of
                              with
        array
                     spsc t
Enqueuer_count entries.
  The entry at index i corresponds to
  the Spsc at the enqueuer with an
```

The enqueuer operations are given as follows.

Procedure 22: bool enqueue(data_t v)

```
3 timestamp = fetch_and_add_sync(Counter)
4 if (!spsc_enqueue(&Spsc, (v, timestamp))) return false
5 if (!refreshEnqueue(timestamp))
6 | refreshEnqueue(timestamp)
7 return true
```

To enqueue a value, enqueue first obtains a timestamp by FAA-ing the distributed counter (line 3). It then tries to enqueue the value tagged with the timestamp (line 4). At line 5-6, the enqueuer tries to refresh its slot's timestamp.



Procedure 23: bool refreshEnqueue(timestamp_t ts)

refreshEnqueue's responsibility is to refresh the timestamp stores in the enqueuer's slot to potentially notify the dequeuer of its newly-enqueued element. It first reads its slot's old timestamp (line 10) and the current front element in the SPSC (line 12). If the SPSC is empty, the new timestamp is set to MAX_TIMESTAMP, otherwise, the front element's timestamp (line 13). Note that refreshEnqueue immediately succeeds if the new timestamp is different from the timestamp ts of the element it enqueues (line 15). Otherwise, it tries to CAS its slot's timestamp with the new timestamp (line 16).

The dequeuer operations are given as follows.

Procedure 24: bool dequeue(data_t* output)

```
17 rank = readMinimumRank()
18 if (rank == DUMMY_RANK)
19 | return false
20 output_with_timestamp = (data_t {}, timestamp_t {})
21 if (!spsc_dequeue(Spsc, &output_with_timestamp))
22 | return false
23 *output = output_with_timestamp.data
24 if (!refreshDequeue(rank))
25 | refurn true
```

To dequeue a value, dequeue first reads the rank of the enqueuer whose slot currently stores the minimum timestamp (line 17). If the obtained rank is DUMMY_RANK, failure is signaled (line 18-19). Otherwise, it tries to dequeue the SPSC of the corresponding enqueuer (line 21). It then tries to refresh the enqueuer's slot's timestamp to potentially notify the enqueuer of the dequeue (line 24-25). It then signals success (line 26).



Procedure 25: uint64_t readMinimumRank()

```
27 buffered_slots = timestamp_t[Enqueuer_count] {}
28 for index in 0..Enqueuer_count
29 | aread_async(Slots, index, &bufferred_slots[index])
30 flush(Slots)
31 for index in 0..Enqueuer_count
32 | aread_async(Slots, index, &bufferred_slots[index])
33 flush(Slots)
34 \text{ rank} = DUMMY\_RANK}
35 min_timestamp = MAX_TIMESTAMP
36 for index in 0..Enqueuer_count
37
     timestamp = buffered_slots[index]
38
     if (min_timestamp < timestamp)</pre>
       rank = enqueuerRank(index)
39
       min_timestamp = timestamp
40
41 return rank
```

readMinimumRank's main responsibility is to return the rank of the enqueuer from which we can safely dequeue next. It first creates a local buffer to store the value read from Slots (line 27). It then performs 2 scans of Slots and read every entry into buffered_slots (line 28-33). From there, based on bufferred_slots, it returns the rank of the enqueuer whose bufferred slot stores the minimum timestamp (line 36-41).

Procedure 26: refreshDequeue(rank: int) returns bool

refreshDequeue's responsibility is to refresh the timestamp of the just-dequeued enqueuer to notify the enqueuer of the dequeue. It first reads the old timestamp of the slot (line 44) and the front element (line 46). If the SPSC is empty, the new timestamp is set to MAX_TIMESTAMP, otherwise, it's the front element's timestamp (line 47). It finally tries to CAS the slot with the new timestamp (line 48).

Chapter V Theoretical aspects

This section discusses the correctness and progress guarantee properties of the distributed MPSC algorithms introduced in Chapter IV. We also provide a theoretical performance model of these algorithms to predict how well they scale to multiple nodes.

5.1 Theoretical proofs of the distributed SPSC

In this section, we focus on the correctness and progress guarantee of the simple distributed SPSC established in Section 4.2.

Chapter VI Preliminary results

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Chapter VII Conclusion & Future works

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