A Resizable C++ Container using Virtual Memory

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Abstract: Thread safety is required for shared data structures in shared-memory parallel approaches, but cannot be done efficiently for standard C++ containers with continuous memory storage, such as std::vector. Dynamically resizing such a container may cause pointer and reference invalidation and therefore cannot be done in parallel environments without exclusive access protection to the container. We present a thread-safe no-copy resizable C++ container class that can be used to store shared data among threads of a program on a shared-memory system. The container relies on the virtual memory controller to handle allocation as needed during execution. A block of memory of almost arbitrary size can be allocated, which is only mapped to physical memory during the first access, providing hardware-level thread blocking. All synchronization costs are already included in the operating system memory management, so using the container work as expected even after the container is resized. The implementation is not general however, and relies on the specifics of the operating system and computer architecture. Memory overhead can be high as the allocations are bound to the granularity of the virtual memory system.

1 INTRODUCTION

When initializing a data structure that is comprised of many elements of the same type, e.g., a dictionary, tree, graph, mesh, etc., it is often a case that the a) total size of the structure is not known in advance, b) the structure can be built iteratively with the elements being inserted at the end of the existing data structure's memory block. An example of such an initialization is the generation of scattered nodes for discretization of a computational domain (Slak and Kosec, 2019). The number of nodes that will be required to discretise the domain is not known in advance, and the generating procedure is iteratively adding them to a container structure and to a spatial indexing structure.

The generation procedure has good parallel potential but requires a container that:

- can be resized in a thread-safe manner,
- will not be modifying the elements after they are put in and
- will only be inserting the elements at the end, that is, the insertion will not require moving any existing elements.

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Most available resizable data structures with storage over a continuous block of memory offer only read-only element access in a thread-safe manner. During the insertion of a new element, the underlying block of memory can be filled up and the whole structure needs to be moved into a different block of memory. The move invalidates all references to existing elements and implementing it in a thread-safe manner can make it very inefficient. To make resizing such a container truly tread-safe, locking mechanisms such as mutexes (Raynal, 2012) must be employed across large portions of code, even those that do not deal with resizing directly. This can lead to poor scaling, especially when many insertions can occur at the same time.

In this paper we propose a container with threadsafe resizing that exploits virtual memory - a feature of modern operating systems that abstracts physical memory away from the programmer both for more efficient memory use and process isolation (Gorman, 2004; Bhattacharjee et al., 2017). It allows the programmer to offload the burden of memory allocations to the operating system, or more specifically to CPU's memory management unit (MMU). As a consequence, no locking mechanism is required in the program code, since thread safety is taken care of by

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the hardware. The use of virtual memory also represents the negative aspect of this approach, that is, it depends on virtual memory support being present in the operating system.

The rest of the paper is organised as follows. In section 2 we describe the standard approaches and virtual memory in more detail, together with the MMU. In section 3 we describe the implementation of the virtual memory container in C++. In section 4 we compare the performance of virtual memory container to some other possible approaches to the problem. In section 5 we discuss the portability of our implementation to other architectures and operating systems.

2 BACKGROUND

2.1 Standard Approach to Resizable Containers

Probably the most widely used container in standard C++ is the std::vector. It is an example of a dynamic array with the ability to resize itself automatically when one or more elements are inserted or deleted. Storage for the elements is being handled transparently by the container.

C++ standard (ISO, 2017) for example describes:

A vector is a sequence container that supports random access iterators. In addition, it supports (amortized) constant time insert and erase operations at the end; insert and erase in the middle take linear time. Storage management is handled automatically, though hints can be given to improve efficiency. A vector satisfies all of the requirements of a container and of a reversible container, of a sequence container, including most of the optional sequence container requirements, of an allocator-aware container, and, for an element type other than bool, of a contiguous container.

Since vector must be continuous, it can be used to interact with C code which expects continuous arrays of elements, it has minimal memory overhead, and the speed of its random access iterators is unparalleled among containers. This, along with its ability to grow automatically or per request, makes it a workhorse for the data structures in general.

It is, however, limited by its implementation of dynamic resizing, which may catch an unprepared programmer off guard with its linear time performance, invalidation of iterators, pointers and references to elements, and the resulting lack of thread-safety even in cases when elements are only added to the vector and never modified. The usual implementation of the algorithm to grow vector in size is to allocate a larger block of memory, copy or move the contents from the old to the new block, and deallocate the old block of memory. When done automatically, the growth ratio is usually exponential, e.g. doubling the memory block size on each growth. Only when done manually, memory block of an exact required size may be obtained for holding the vector's elements.

When vector is grown, all the newly available elements, that is, the elements with indices *i*, where old_size $\leq i < \text{new_size}$ are also default initialized. The creation of vector of fixed size itself includes the growth operation and the overhead associated with it.

For parallel programming, vector in raw form is ill suited, since it is not thread-safe in the slightest. Full read-only concurrent access is allowed, but read-only access to one element while another element is being modified is only allowed if vector does not get resized by he modification, i.e. function push_back should not be used on a vetor when it is being concurrently accessed by other threads. When dynamic resizing is required, all element access has to be protected, at least with a readers-writer lock(Raynal, 2012).

2.2 Virtual Memory and MMU

Virtual memory is an abstraction of the physical memory that hides the arrangement of physical memory and other resources from the programmer and separates different processes' address spaces (Bhattacharjee et al., 2017). The mapping between virtual and physical memory addresses is carried out by the operating system kernel and the memory management unit (MMU), a dedicated address translation hardware, built into the CPU (Gorman, 2004).

On 64-bit version of Linux each process can access a total of 256 TiB of virtual memory. The upper 128 TiB are reserved for the kernel while the lower 128 TiB can be used by the program, as shown in Figure 1. A part of this user memory region is reserved by the stack and executable code, but the vast majority is freely allocatable. Not only physical memory can be mapped into virtual memory, but also swap partition, normal files on disk, I/O devices, etc.

While up to 128 TiB of virtual memory can be allocated, in practice only as much of it can be used as there is physical memory in the system. For a fixed size, densely populated data structures virtual memory presents no real improvement over direct access



Figure 1: Structure of the virtual memory address space belonging to a process in the 64-bit Linux environment. Not to scale, a large part of address space not in use is omitted.

in terms of performance. Much of the improvement is achieved when using either large sparse structures or structures that must increase in size frequently. We can allocate a large sequential chunk of virtual memory and only write to a small portion. The kernel then maps only the accessed memory locations into physical memory.

Address mapping is done on the granularity of a page. Pages are fixed-size blocks of memory. Their size is determined by the operating system and the CPU architecture and can range from 1 KiB to 1 GiB, the most common being 4 KiB (Gorman, 2004). When a page in virtual memory is first written to, the kernel uses the MMU to map it to a page in physical memory.

The MMU contains a complex hardware component that translates memory addresses between process' virtual address space and physical address space. The details of its operation surpass the scope of this work, we can think of it as a black box that performs address mapping in a thread-safe manner, regardless of its implementation.

3 IMPLEMENTATION

The virtual memory container is implemented using the provided memory management functions. POSIX-compliant operating systems such as Linux expose a memory management header containing such functions (IEEE, 2018), sys/mman.h. To implement the container we used functions mmap, mprotect, and munmap, as shown in Figure 2.

During the construction of the container the required amount of virtual memory is allocated using mmap. Initially all memory is flagged as inaccessible, to prevent an out-of-memory error on systems with /proc/sys/vm/overcommit_memory set to 0. The memory is then marked as readable and writable in 1 GiB chunks¹ using the mprotect function.

Note that the implementation is deliberately kept minimal and includes only the constructors, destructor, element accessor functions (operator[]) and size function. This is a consequence of our current requirements for the container; the implementation could be extended into random-access container such as vector. Comparison to containers of the C++ standard library is therefore limited. We do believe, though, that the presented approach could be extended into a standard library compatible allocator class, similar to mmap_allocator², to facilitate the use of virtual memory by the containers of the standard library.

Indexing into the container is done directly as indexing into an array pointer. If the page of virtual memory inside which the target element is located is not yet mapped into the physical memory, the MMU creates this mapping on the first access. During the mapping, several virtual memory pages may be remapped, moved to or from swap and a target physical memory page is allocated and cleared. Therefore, mapping is a complex procedure that can potentially take a long time to complete. During this time, all the threads that wish to access the new page are blocked in their memory access instruction.

During the destruction of the container, the allocated virtual memory is deallocated using munmap. Care must be taken to unmap all allocated memory. For this reason the size of the container is recorded during construction. Deletion of all mappings from virtual to physical memory is taken care of by the kernel by the system function call.

¹This can be avoided if overcommit_memory is set to 1 within the operating system.

²https://github.com/johannesthoma/mmap_allocator

```
#include <sys/mman.h>
#include <errno.h>
#include <string>
#include <stdexcept>
#include <iostream>
template <typename ElementType>
class VirtualMemoryContainer {
private:
 size_t num_elements;
  ElementType* container;
public:
  VirtualMemoryContainer(size_t max_elements) {
    num_elements = max_elements;
    container = (ElementType*) mmap(nullptr, sizeof(ElementType) * max_elements, PROT_NONE,
    MAP_PRIVATE | MAP_ANONYMOUS, -1, 0);
    if (container == (ElementType*) -1) throw vmc_error(errno);
    const int64_t block_size = 1048576;
    int64_t remaining = sizeof(ElementType) * max_elements;
    uint8_t* head = (uint8_t *) container;
    while (remaining > 0) {
      int ret = mprotect(head, std::min(block_size, remaining), PROT_READ | PROT_WRITE);
      if (ret != 0) throw vmc_error(errno);
      head += block_size;
      remaining -= block_size;
 }
 VirtualMemoryContainer(VirtualMemoryContainer& other)
                                                         = delete;
 VirtualMemoryContainer(VirtualMemoryContainer&& other)
   num_elements = other.num_elements;
    other.num_elements = 0;
    container = other.container;
    other.container = nullptr;
  }
  ~VirtualMemoryContainer() {
    if (container != nullptr) {
      int ret = munmap(container, sizeof(ElementType) * num_elements);
      if (ret != 0) throw vmc_error(errno);
    }
  }
  inline ElementType& operator[](size_t idx) {
    return container[idx];
  }
  inline const ElementType& operator[](size_t idx) const {
    return container[idx];
  }
  inline size_t size() const {
    return num_elements;
  }
```

};

Figure 2: The implementation of the virtual memory container. Note that supporting classes such as vmc_error are not listed since their implementation is not relevant.

4 PERFORMANCE COMPARISON

In this section we demonstrate the performance advantage of the proposed container over a standard implementation of vector. Our use case for which we developed the container is holding the elements of dynamically built *k*-d tree, which performs spatial indexing for numerical domain discretization (Slak and Kosec, 2019). This use case makes it difficult to quantitatively compare the container implementations, since it involves complex computing that masks the performance gains of the container itself.

Furthermore, there seem to be no standard benchmark for perfomance measurement of thread-safe containers, as other approaches present their own benchmark problems (Dechev et al., 2006). To make the demonstration simple, we consider a solver for Collatz conjecture (Lagarias, 1985) on a predefined range of numbers, which involves only light computing and is memory bound.

4.1 Collatz Conjecture

Consider the following function on positive integers:

$$f(n) = \begin{cases} \frac{n}{2}, & \text{if } n \text{ is even} \\ 3n+1, & \text{if } n \text{ is odd} \end{cases}$$
(1)

Now repeatedly apply this function to some positive integer a_0 , so that $a_{n+1} = f(a_n)$. Does the sequence a_0, a_1, a_2, \ldots eventually reach 1? The Collatz conjecture states that it will, no matter which positive integer is chosen as a_0 .

The conjecture has not been proven or disproven, although numerical tests have shown that it holds for all starting values up to 10^{20} (Bařina, 2021). Unpredictable behaviour of a generated sequence means that although the starting integer may be small, the processing can involve very large numbers before the sequence reaches 1. One integer that exhibits such behaviour is 27: it peaks at 9232, after 77 successive applications of f, a number more than 300 times larger than the starting number.

4.2 **Problem Statement**

To estimate how fast the sequence converges to 1 for some starting positive integer k, we want to determine the number of applications of f before the sequence reaches 1. Let us name this number the *orbit length* o_k of k. We want to generate a list of orbit lengths for all integers between 1 and n, where n is given in advance. To speed up the computation we make the following observation: If a number k has an orbit length of o_k , then its successor f(n) has an orbit length of $o_k - 1$. Therefore we can compute orbit lengths recursively, stopping at k = 1. Since numbers larger than n might appear more than once in the generated sequences, we want to store their orbit lengths as well. We, however do not know the largest number for which orbit length will have to be stored and cannot preallocate the required memory to store all the numbers. The largest encountered number will be known only after the algorithm completes.

For efficiency, we do not want to preallocate all of the available system memory and would like to grow the container for orbit lengths as necessary. Growing as necessary is the main strength of the proposed container, since it can be done with minimal overhead. For standard library containers, though, this is not the case. We opt to use vector as comparison reference, since it is a random access container with minimal memory overhead and can be resized at will. The standard vector is different however in the way resizing is implemented (large computational overhead) and in its initialization of all the new elements upon each size change. Note that deque from standard library could be used instead, since its implementation usually uses a different trade-off approach to slightly slower random element access for a faster resize operation.

The proposed container does not perform element initialization and relies on the fact that the system clears all the allocated pages with zeros when mapping them. Initialization with zeros is sufficient both in the presented case and in our use-case, thus no additional compliance with the standard containers is attempted. Additional initialization would require the concept of size to be implemented in addition to the concept of allocated space, which is not necessary in our use case. Nevertheless, it is important to notice that the comparison is not entirely fair since the proposed container has one less function to perform in all experiments.

4.3 Implementation

To measure the performance of virtual memory container against the vector we implemented the described problem as a computational experiment in C++. We want to gain insight into which parts of computation take the longest, so we measured setup time, calculation time and resize time separately, where calculation time includes time spent reading and writing to the container.

We compare the virtual memory container with



Figure 3: Total execution times of the described problem for the virtual memory container and different configurations of the vector on one CPU thread.

the vector in three configurations: Where vector is completely empty and must be resized, where we construct the vector with of size *n*, which we shall refer to as *target size*, and where we construct the vector with the maximum allowed number of elements, which we shall refer to as *maximum size*.

We decided to impose an upper limit to the size of containers so we don't run into problems with a lack of physical memory. For each value of n we repeated the experiment many times and then averaged measured times to minimize any potential random variance from external factors, such as the varying CPU clock and process scheduler.

A single execution of the experiment proceeds as following:

- For each integer *i* between 1 and *n* noninclusive the orbit length is calculated recursively in ascending order.
- First f(i), the successor of *i*, is calculated.
- If f(i) is larger than or equal to the maximum size, then f is applied to it iteratively until the calculated value is smaller than the maximum size. The total number of iterations is recorded.
- If iteration was not necessary the number of iterations is set to 1.
- The size of the container is checked. If the size is smaller that the successor, the container is resized so that it contains the successor. (This only happens when vector is used).
- If the successor is 1, then the orbit length is also set to 1.
- If the successor's orbit length was not yet calculated then the program recursively calculates it.



Figure 4: Total execution times of the described problem for the virtual memory container and different configurations of the vector on eight CPU threads.

The final orbit lenght is the orbit length of the successor plus the number of iterations.

4.4 Performance

We executed the benchmark problem on an 8-core, 16-thread computer (2x Intel Xeon E5520) with 12 GB of RAM. In the experiments we limited the maximum allowed size of the containers and we performed several experiments with various limits. We present the results only for the limit being exactly 1 GiB of physical memory in Figures 3 and 4. We further performed the experiments also on the limits of 1 GiB, 2 GiB, 4 GiB, and 8 GiB, but discovered that they follow the same pattern.

Within the experiment with 8 GiB limit though, we were reminded of one often neglected negative aspect of vector. In order to resize a vector, temporarily the amount of physical memory needed equals the new size plus the old size, which in limit goes up to twice the requested size. This happens because during the resize process, first new memory of at least the requested size is allocated, then elements are copied and only then the old memory is freed. Our experimental machine had enough physical memory to host an 8 GiB structure, but not enough to support the resize, causing it to use swap file on each resize after the container reached about 6 GiB in size. The performance of vector was severly degraded and thus the results were nonrepresentative.

The execution times for 1 CPU thread are shown on the figure 3. The fact that vector copies all values during the resizes causes it to fall far behind the virtual memory container. The trend continues up to about $n = 10^5$, where the difference in compute times





Figure 5: Execution time percentages of the described problem for the virtual memory container and different configurations of the vector. Blue: container setup, orange: calculation of values, green: container resizing.

starts to shrink. At this point the percentage of filled elements in the container starts becoming significant, so the virtual memory container must allocate a lot more pages. Regardless, the virtual memory container remains the fastest, challenged only by the *maximum* size vector for *n* above 10^8 .

The execution times for 8 CPU threads are shown on the figure 4. The difference between the virtual memory container and vectors widens even further, with execution over different configurations of the vector being slowed down significantly. The feature of the virtual memory container, which allows it to be resized without any locking mechanisms allows it to be used without changes, while vector has to be protected using a readers-writer lock (using the standard C++ shared_lock and unique_lock on shared_mutex). While vector configurations were slower on the order of a magnitude because of the use of mutexes and full locks during resizes, the virtual memory only slowed down for some n while gaining speed for other.

5 PORTABILITY

The presented implementation of the virtual memory container may not be used on all operating systems nor on all the hardware. As was discussed in section 2, for virtual memory mapping to work, a MMU is required. Such hardware is not present in all computer architectures, such as those in embedded and IOT environments for example.

Furthermore, one of the requirements of the container is the availability of memory management functions. While such functions may be available, they might not behave in the same way as those defined by the POSIX standard. One such case is the Microsoft Windows operating system. While it contains the required facilities needed to implement the container, a different set of functions or a translation layer such as mman-win32³ must be used.

6 CONCLUSION

In this paper we presented an often neglected approach to containers in C++, the virtual memory based container. The container relies on the virtual memory controller to handle allocation when its size increases beyond the current value. All synchronization is included already in the operating system memory management, so there is no additional cost to us-

³https://github.com/klauspost/mman-win32

ing the container in parallel environment. As a bonus, references and pointers to elements of the container are allowed and work as expected even when the container is resized. There are negative aspects though, for example, not all hardware architectures allow for the implementation and currently only Linux implementation is provided. The amount of the allocated memory can also not be as fine grained as with the classic C++ containers, and there is some memory overhead attached.

Although we present a very basic, array-like (or std::vector-like) container, the approach could be extended into an allocator, to support the allocations for containers of the standard library. Our main motivation for developing this container lies in its inherent thread safety for a special scenario, which is often found in real life. That scenario is one in which the elements are continuously appended to the container, and the total number of elements is not known beforehand. While the existing elements of the container are also frequently read from, they are never modified. If this scenario is required in a parallel environment, where either the elements are read in parallel or appended in parallel, or both, then the existing containers require extensive protection of shared data structures and thus offer very low performance. In addition, we found that the container also has several positive aspects, compared to for example vector, even when used in completely sequential code, which then makes it an even more compelling option to consider in everyday development.

Use cases where the presented container presents the largest performance gain are those which are memory bound, where data processing cannot be localized and requires many memory accesses. Furthermore the standard approach with container locking during the resize operation scales poorly when many CPU threads are used compared to the presented lockfree design. Through experiments on an artificial example - a solver for orbit lengths for the Collatz conjecture - we demonstrate where the presented container shines the most, compared to the containers offered by the standard library. We make a short scan of the input parameters to show the performance both on sequential and parallel access to the container. In the sequential program, we find that the performance of the presented container is only matched by vector with constant size equal to the maximum allowed size. Since thread safety is required for the shared data structures in shared-memory parallel approaches, but cannot be done efficiently for C++ containers with continuous memory storage and dynamic size, such as std::vector, we find that in parallel programs, the presented container perfomance is unmatched.

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