

Rethinking RAM: Testing alternative models of computation

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15 May 2013

Abstract

This is where the abstract will briefly summarize how great our project is. After all, we discovered that the VAT/RAM model is superior to the RAM/VAT model of computation.

Keywords RAM Model, computational models, benchmarks, algorithms

1 Introduction

The Random Access Machine (RAM) model of computation is widely used for analyzing the performance of algorithms. Some, such as Jurkiewicz and Mehlhorn question whether this model accurately represents the complexities of modern hardware. They propose the Virtual Address Translation (VAT) model of computation to account for added complexities, such as the memory address translation needed to reference ever larger amounts of system memory in modern computers.

This paper analyzes a set of algorithms using both models, VAT and RAM, and presents the results of benchmarking these algorithms on real hardware. The goal of this is to then use the results of the analysis and benchmarks to determine which model more accurately reflects the real hardware.

The novel contribution includes not only the benchmarking of these procedures, but also the comparative analysis of the two models. This work could be very significant if VAT is shown more accurate than RAM. We

would be the first to independently verify this and we would be contributing some of the first analysis of common, reference algorithms. (We’re still working on results.)

This paper is organized as follows: Section 2 presents the paper’s motivation. Section 3 addresses previous work. Section 4 details how the benchmarks and analyses were done. Section 5 presents the results. Section 6 discusses the results, selecting the more accurate model. Section 7 and 8 address the potential for future work and conclude the paper.

2 Motivation

Analysis of algorithms on abstract machines serves as a crucial element of computer science, which allows incredible increases in the speed and utility of computer programs. Until now, computer scientists have used the RAM and the EM models to perform algorithmic analysis accurately. But Jurkiewicz and Mehlhorn observe discrepancies with some experimental findings and attempt to push forward the VAT model to account for these discrepancies.

Before algorithm analysts shift to this new model, they must carefully verify its claims and correctness, it’s utility. This is where Jurkiewicz and Mehlhorn’s contribution falls short. Their paper does mention that experimental findings tally with the theoretical predictions of the new VAT model, but it does not clearly represent these findings. Additionally the number of test cases is limited. So before scientists readily accept this new model for theoretically analyzing the running time of algorithms the test set must be broadened to verify and compare the experimental results with the model’s predictions.

If it can be verified by experimental comparison that the new VAT model is more accurate, then this will serve as an incentive to researchers to use the proposed model for more accurate estimations of running time of algorithms. More accurate algorithm analysis will help maintain the pace of innovation enjoyed by computer science for a generation.

3 Previous Work

It is evident that most papers on algorithms tend to ignore the cost of virtual address translation, though researchers recognize these costs. This is because of the model that researchers follow, and that is the classic RAM model without virtual memory. Papers like “The cost of virtualization” by

Ulrich Drepper and other materials like “AMD64 Architecture Programmer’s Manual Vol 2” describe the implementation of virtual memory and its associated costs during translation; but no study has tried to develop a model that considers these costs for algorithmic analysis prior to the recently proposed VAT model.

Keeping this trend in mind, it is not surprising that there has been little related work that carefully verifies the experimental results and strengthens the new model that accounts for these costs. Our contribution, experimentally examining and comparing the proposed model to the older model, will fill this gap.

4 Methods

Three components are necessary to determine which model of computation yields a more accurate analysis of the performance of algorithms running on real hardware. First, there must be implementations of algorithms running on real hardware. Running time benchmarks establish a ground truth for doing comparisons. Second, the algorithms must be analyzed according to the new VAT model. And third, the algorithms must be analyzed using the RAM model. The authors selected five well-studied algorithms which have been previously analyzed with the RAM model.

4.1 Benchmarking

To generate a standard for comparison, the authors selected a set of five algorithms: Binary Search, Heapsort, Insertionsort, Quicksort, and Permute. These five are well understood, have a variety of running times, and should be straightforward enough to analyze on the new model.

The algorithms were coded in C++ and compiled with GCC and Make. They were run on a VirtualBox virtual machine, with Debian GNU/Linux for 32 bit x86 processors. The virtual machine was hosted by 64 bit Windows 7 on a 2.0 GHz Core i7 processor. The physical machine has 8 GiB of memory and four processor cores, of which 2.5 GiB and 2 cores were allocated to the virtual machine. Find a link to the source code in the appendix.

The benchmark program ran each procedure 10 times on each input size. The smallest input to each procedure was an integer array of length 1. The input doubled in size after every tenth run. The maximum size input array to Binary Search was 536870912 32 bit integers, the 2 GiB size reaching the limits of the machine’s memory capacity. The maximum size input array to Insertionsort was 1048576 32 bit integers, the execution time on any larger

input becoming excessive and overflowing the capacity of the timer. The other three procedures had maximum size input arrays of 268435456 32 bit integers, each array taking up 1 GiB of system memory.

The benchmark program timed each run of each procedure on each size of input. The raw results are available at the link provided in the appendix. The next section of the paper provides a summary presentation.

4.2 VAT Model Analysis

As mentioned in the previous sections, the VAT model introduces the concept of virtual memory to the existing RAM machines. The motivation behind using the VAT model is multiprocessing where several programs are executed concurrently on the same machine. In this scenario the VAT model provides each of the concurrently running programs with a linear address space with non negative indices. But these addresses are virtual and are simulated with one physical memory. This implies that to get the actual physical memory location, some translation from the virtual address space to the physical address space is required which in turn adds some costs to the algorithmic complexity. Costs are also associated with page faults and TLB misses.

The main data structure that is used for the translation is a tree with outdegree K and the translation process is a walk in this tree. The tree is also referred as the page table (consisting of entries that map virtual to physical addresses). The leaves of the tree store indices of physical pages and the offset determines the cell in the physical address.

The translation process is done by a Translation Cache (TC) residing in the RAM which stores some nodes of the translation tree. The TC is changed by insertions and evictions and follows efficient replacement strategies[5]. To translate a virtual address to a physical address we start from the root node of our tree and continue traversing the nodes as mentioned in the virtual address and stop when we reach a leaf. Therefore translating a virtual address requires access to the nodes of the translation path in the TC in the correct order. The length of the translation is the number of insertions performed during translation and the cost of the translation is r times the length. An elaborate explanation of the translation tree can be found in the appendix section of our reference paper.[5]

For simplicity, we consider the virtual memory of a single program. Now for analyzing the translation cost of algorithms as a function of problem size n and memory m , we consider $m = \Theta(n)$ and we assume.[5]

- $rd \leq P$ the cost of moving a single translation path to the TC is no

more than the size of a page, i.e., if at least one instruction is performed for each cell in a page, the cost of translating the index of the page can be amortized.

- $K \geq 2$ the fanout of the translation tree is at least two.
- $\frac{m}{P} \leq Kd \leq \frac{2m}{P}$ the translation tree suffices to translate all addresses but is not large. As a consequence $\log(\frac{m}{P}) \leq d \log(K) = Kd \leq 1 + \log(\frac{m}{P})$ and hence $\log(k)(\frac{m}{P}) \leq d \leq \frac{1}{k}(1 + \log(\frac{m}{P}))$.
- $d \leq W$ the translation cache can hold at least one translation path.

We have used the above concepts to calculate the algorithmic complexities of the algorithms that we have benchmarked. Also, the graphical representations in the following sections provide an intuitive summary of our findings.

5 Results

Here the results are presented. First come predictions of running times generated by the RAM model. Then come the analyses of the algorithms under the VAT model. Last are the results of benchmarking the algorithms.

5.1 RAM

This paper selected five reference algorithms, with running times well studied under the RAM model. The following table presents those running times. These will be used later to make comparisons.

Algorithm	Average Case	Worst Case
BinarySearch	$O(\log(n))$	$O(\log(n))$
Heapsort	$O(n \log(n))$	$O(n \log(n))$
Insertionsort	$O(n^2)$	$O(n^2)$
Permute	$O(n)$	$O(n)$
Quicksort	$O(n \log(n))$	$O(n^2)$

It is left as an excersize for the reader to look these up in any algorithms reference text.

5.2 VAT

The VAT model presents more of a challenge than checking a reference text. Jurkiewicz and Mehlorn provided analyses[5] for two of the algorithms. Those results are here:

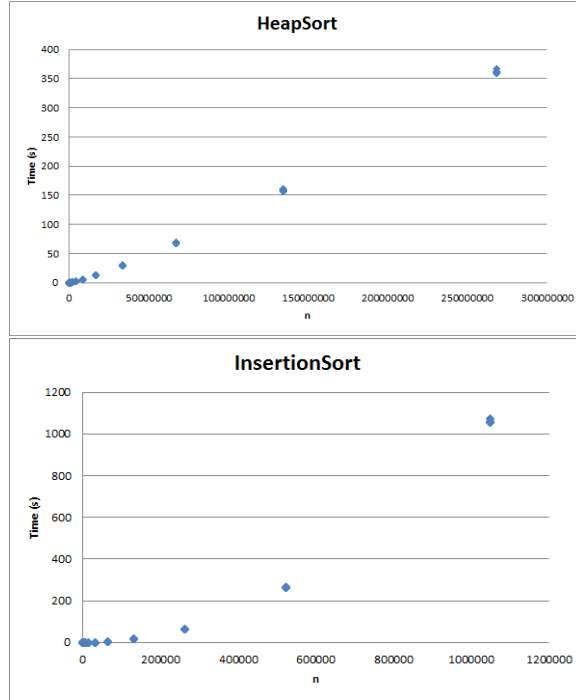
Algorithm	VAT Complexity
BinarySearch	$O\left(\frac{r}{k} \log\left(\frac{2n}{P}\right) \log\left(\frac{2dn}{W}\right)\right)$
Heapsort	$O\left(r\left(d + \frac{pn}{P}\right)\right)$

Where r is the unit cost of translation (length/number of insertions in the TC). d is the length of translation. W represents the nodes of the translation tree that are in the TC. P is the page size. And, n is the input size.

5.3 Benchmarks

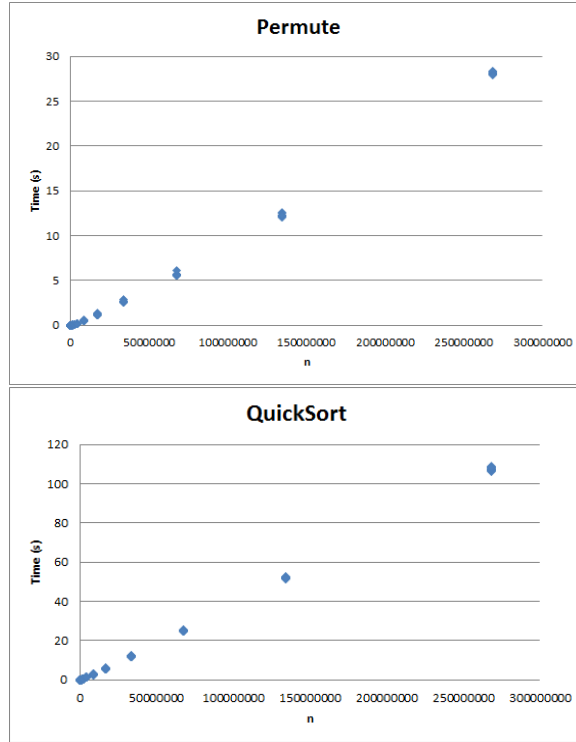
Benchmarking the algorithms on a modern computer is a key step in determining whether VAT or RAM more accurately models a computer for the purposes of algorithm analysis.

BinarySearch of an integer array was the first algorithm measured. Reviewing the results of timing, one would see all search times as 0.000000 seconds. The BinarySearch procedure executed too quickly, even on a 2 GiB array, to produce a result within the precision of the program's measurement.



The sorting algorithms were executed on randomly generated integer arrays, and the permute procedure was executed on a sorted array to yield

a randomized array of integers. The accompanying scatter plots show the execution times of each procedure at each size input.



6 Discussion

This is where we discuss results.

6.1 RAM vs Reality

Here is where we will compare the benchmark results with the standard runtimes. This will need a table to compare the two algorithm-by-algorithm.

6.2 VAT vs Reality

Here is where we will compare the benchmark results with the new model's runtimes. This will need a table to compare the two algorithm-by-algorithm.

6.3 VAT vs RAM

The big finale, old vs new, here is where we will declare a winner between RAM and VAT. This will likewise need a table. Really all three tables could be combined into one good chart for the whole section.

7 Future Work

This paper leaves open some possibilities for future work. In the first place, mathematical models are always up for refinement. No mathematical construct has been found that can perfectly simulate the physical universe. Thus, there is always room for improvements to the model.

With respect to the analysis and conclusions given in this paper, one might see that the analysis was conducted on consumer grade, personal computer hardware. This analysis may not necessarily apply to situations involving big data, servers, cloud computing, or new technology. Perhaps there is not a single mathematical model to precisely describe every arrangement. These areas are worth further study.

8 Conclusions

Our paper has a focused aim of experimentally verifying the VAT model of algorithmic complexity. Jurkiewicz and Melhorn has come up with an innovative concept of incorporating the costs of virtual address translation[5] in algorithmic complexities, but did not provide enough evidence to support the use of this new model. Thus, an experimental verification was required and we have managed to do that with our findings. Our straightforward verification and benchmarking process will help computer scientists decide whether to shift to this new VAT model or not. Though we experienced some difficulties in the mathematical deduction and we plan to refine our calculations in future, our experimental results can be still be used for decision making. As we continuously analyze and benchmark more algorithms, we will be able to present our personal view on using the VAT model, in our final paper. Should the project determine that the VAT model is indeed superior to other existing models of computation, the field of algorithm analysis will be changed and its predictions will more accurately reflect the real state of the world. This will enable better algorithm design and help to allow computer scientists to continue advancing information technology to meet the needs of the society.

9 References

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Appendix

All source code, benchmarks, and report drafts can be found at:

<https://github.com/dslachut/adv-algo-project>

Item	Address
Benchmarking Source Code	adv-algo-project/tree/master/prof++/
Benchmarking Executable	adv-algo-project/blob/master/prof++/Debug/prof++
Benchmark Results	adv-algo-project/tree/master/benchmarks
Final Report	adv-algo-project/tree/master/Final

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