



Seminar Report

Introduction to Performance Engineering in Rust

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Abstract

[Put your abstract here.] Note that we typically publish your report as PDF on our webpage. Let us know if you disagree.]

General structure of an abstract, write 1 to 2 sentences per section

- 1. A general statement introducing the broad research area of the particular topic being investigated.
- 2. An explanation of the specific problem (difficulty, obstacle, challenge) to be solved.
- 3. A review of existing or standard solutions to this problem and their limitations.
- 4. An outline of the proposed new solution.
- 5. A summary of how the solution was evaluated and what the outcomes of the evaluation were.

You may find the following resources useful:

- https://www.grammarly.com/blog/write-an-abstract/
- $\bullet \ \text{https://www.editage.com/insights/manuscript-structure-how-to-convey-your-most-implies} \\$
- More useful links: https://hps.vi4io.org/teaching/ressources/start

Statement on the usage of ChatGPT and similar tools in the context of examinations

In this work I have used ChatGPT or a similar AI-system as follows:

✓ Not at all
☐ In brainstorming
\Box In the creation of the outline
\square To create individual passages, altogether to the extent of 0% of the whole text
☐ For proofreading
□ Other, namely: -

I assure that I have stated all uses in full.

Missing or incorrect information will be considered as an attempt to cheat.

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List of Abbreviations

ADT Algebraic Data Types

HPC High-Performance Computing

IR Intermediate Representation

KDE Kernel Density Estimation

LLVM Low Level Virtual Machine

PBL Problem-Based Learning

PDF Probability density function

RAII Resource acquisition is initialization

1 Introduction

1.1 Motivation

From a programming language perspective, High-Performance Computing (HPC) is dominated by code written C, C++ or Fortran as these provide the low level control and optimization capabilities common in tightly-optimized code. However, as Rust was initially designed as a modern, memory-safe C++ replacement, it could be a valid choice for any kind of performance-critical code.

Instead of just providing yet another taxonomy of successful HPC projects in Rust, this report will rather provide an introduction to the topic on performance engineering in Rust, implicitly covering the current state of the ecosystem while providing an short explaination of each of the common concepts. Instead of just providing an enumeration of techniques, it rather uses Problem-Based Learning (PBL) to introduce the techniques just-in-time when they are relevant, providing a more coherent learning progression.

Problem-Based Learning can be difficult. The problem has to be

- Small enough to fit the scope of a report
- Complex enough to cover most of the concepts of real-life performance engineering
- Interesting enough to keep readers engaged in the topic

For this report, we decided to analyze matrix multiplications. More than just a toy-problem, the matrix multiplication is at the core of all deep learning frameworks. As the parameter count steadily increases into the trillions [1], fast matrix multiplications become evermore important for todays frameworks.

1.2 Rust

Rust [2] is a systems programming language initially released by Mozilla Research in 2015. It was designed as a memory safe alternative for C++ in Servo [3], which is the web rendering engine used in Firefox. Rusts main goal is to provide memory safety while having an on-par performance with other systems languages such as C or C++.

Having memory safety is paramount, as most security issues in traditional C/C++ codebases are a result of using a memory-unsafe language. To quote the overview by Alex Gaynor [4]

- Android [5]: "Our data shows that issues like use-after-free, double-free, and heap buffer overflows generally constitute more than 65% of High & Critical security bugs in Chrome and Android."
- Android's bluetooth and media components [6]: "Use-after-free (UAF), integer over-flows, and out of bounds (OOB) reads/writes comprise 90% of vulnerabilities with OOB being the most common."

- iOS and macOS [7]: "Across the entirety of iOS 12 Apple has fixed 261 CVEs, 173 of which were memory unsafety. That's 66.3% of all vulnerabilities." and "Across the entirety of Mojave Apple has fixed 298 CVEs, 213 of which were memory unsafety. That's 71.5% of all vulnerabilities."
- Chrome [8]: "The Chromium project finds that around 70% of our serious security bugs are memory safety problems."
- Microsoft [9]: " $\sim 70\%$ of the vulnerabilities Microsoft assigns a CVE each year continue to be memory safety issues"
- Firefox's CSS subsystem [10]: "If we'd had a time machine and could have written this component in Rust from the start, 51 (73.9%) of these bugs would not have been possible."
- Ubuntu's Linux kernel [11]: "65% of CVEs behind the last six months of Ubuntu security updates to the Linux kernel have been memory unsafety."

Furthermore, it is now adapted by many big tech firms such as Amazon [12], Google [13], Meta [14], and Microsoft [15]. Lastly, in december 2022, it became the first language other than C and Assembly supported for Linux kernel development [16].

1.2.1 Why Rust is a good fit for HPC

Basically, one can think of Rust as a modern dialect of C++ enforced by the compiler. It uses Resource acquisition is initialization (RAII) internally to ensure memory safety, while references are roughly equivalent to std::unique_ptr.

Especially relevant is the great interoperability with other languages. It supports easy integration with C++ using bindgen [17], which is developed by the Rust core team. Rust also allows for easy embedding into Python code using PyO3 [18], allowing for high-performant native extensions.

Furthermore, it allows for very low level control, even to the extend of bare metal deployment support. Due to Rusts aforementioned RAII-alike memory management model, the runtime has no need for a garbage collector. One can even bring their own memory allocator and do raw pointer arithmetic if required. Lastly, Rust supports architecture based conditional compilation which makes it possible to write fast programs leveraging modern CPU instructions while providing portable alternatives. To support bare metal, OS-less development, Rust's standard library is split into 3 tiers:

- core: The core library provides essential types and functionality that do not require heap memory allocation.
- alloc: The alloc library builds upon the core library but expects heap-allocations, thus supporting things such as dynamically sized vectors.
- std: The std library is the highest-level tier, requiring not only a memory allocator but also several OS capabilities such as I/O management.

Although Rust itself is a relatively new language, its compiler supports most modern compiler optimizations. This is possible through Low Level Virtual Machine (LLVM). Instead of producing native assembly for all architectures, the compiler just provides a LLVM frontend generating LLVM Intermediate Representation (IR) which then gets translated to native code by LLVM.

Lastly, it supports many modern functional concepts such as immutability by default, flat traits instead of deep inheritance, exhaustive pattern matching with Algebraic Data Typess (ADTs) sum types as well as providing alternatives to nullability, which is commonly known as the billion dollar mistake [19]. Its language design is in fact so popular that according to the yearly StackOverflow surveys it was voted as the most loved language for the 7th year in the row [20].

1.3 Quadratic Matrix Multiplication

Let $A, B \in \mathbb{R}^{n \times n}, n \in \mathbb{N}$. Then $C \in \mathbb{R}^{n \times n}$ is defined as

$$C_{ij} := \sum_{k=1}^{n} A_{ik} \cdot B_{kj}.$$

One can think of C_{ij} as the dot product of the *i*-th row of A and the *j*-th column of B.

1.4 Structure

This report is structured as follows: Section 2 will explore a simplified version of the matrix multiplication problem where the dimension is fixed. Here, the focus will be set on microbenchmarking, full application benchmarking, and assembly analysis. Section 3 will then explore the full matrix multiplication, exploring the topics of profiling, compiler optimizations, cache oblivious as well as how to benchmark in noisy environments. Section 4 will provide a short introduction of parallelism. Lastly, section 5 concludes this report by providing an overview of all shown tools as well as further ressources.

2 Fixed Size Matrix Multiplication

To start off with a simplified problem, this section focuses on a fixed quadratic matrix size of n = 3. Using the mathematical definition, this can be trivially implemented:

```
fn matmul(a: Vec<Vec<f32>>, b: Vec<Vec<f32>>) -> Vec<Vec<f32>> {
        let mut result = vec![vec![0.0; 3]; 3];
2
        for i in 0..3 {
3
            for j in 0..3 {
                for k in 0..3 {
5
                     result[i][j] += a[i][k] * b[k][j];
6
                }
            }
8
        }
        result
10
   }
11
   fn driver_code(a: Vec<Vec<f32>>, b: Vec<Vec<f32>>>, c: Vec<Vec<f32>>>)
12
        -> Vec<Vec<f32>> {
13
        matmul(matmul(a, b), c) // D := A * B * C
14
   }
15
```

Listing 1: Naive implementation of a 3×3 matrix multiplication.

The Vec arguments are currently passed as a call-by-value, which means that the whole vector gets copied onto the function's stack. Intuitively, this could be improved by using call-by-reference semantics, which just copies the pointer instead of the underlying data. Theoretically, this should result in a performance improvement. In reality, it is very hard to predict actual performance. Thus, some benchmarking is required. In order to measure the performance, either microbenchmarking or full application benchmarking can be used.

2.1 Microbenchmarking

Microbenchmarking is the performance evaluation of small isolated functions. In the Rust ecosystem, there are two obvious solutions for microbenchmarking: Rust's native cargo bench as well as criterion.rs, which is the modern canonical benchmark library.

Native Benchmarking Cargo, Rust's package manager, supports benchmarking natively through the cargo bench [21] subcommand. Unfortunately, this is still experimental, thus only part of the unstable nightly Rust versions. Furthermore, no clear roadmap to stability exists [22].

cargo bench is a very lightweight microbenchmarking solution. It provides no integrated regression testing nor any kind of visualization or plotting. The 3rd-party cargo-benchcmp [23] utility can be used to compare different benchmarks.

Criterion The other solution is criterion.rs [24], which is also available in stable Rust. It uses basic statistical outlier detection to measure regressions and their significance. Furthermore, it blocks constant folding using the criterion::black_box, which is described as a "function that is opaque to the optimizer, used to prevent the compiler from optimizing away computations in a benchmark" [25]. It automatically generates HTML reports with plots using gnuplot. For benchmark comparisons, the cargo-critcmp [26]

program can be used.

As there is currently no active development in cargo bench, criterion should always be the prefered solution for microbenchmarking.

2.2 Full Application Benchmarking

There are several solutions for benchmarking whole applications, especially as they are usually agnostic to the application's programming language. But to stick to the modern Rust ecosystem, this report will focus on Hyperfine [27], a very actively developed command-line benchmarking tool written in Rust.

From a simplified perspective, full application benchmarking is quite trivial. First, take a timestamp of the current time. Then, run the command to be benchmarked. Afterwards, take a new timestamp. The time delta is the benchmark time. But beyond this core functionality, Hyperfine supports many important and fundamental features for proper benchmarking and analysis.

```
▶ hyperfine --warmup 3 'fd -e jpg -uu' 'find -iname "*.jpg"
Benchmark #1: fd -e jpg -uu
                                                   [User: 1.019 s, System: 1.433 s]
  Time (mean \pm \sigma):
                         329.5 \text{ ms} \pm
  Range (min ... max):
                         326.6 ms ... 333.6 m
                                                   10 runs
Benchmark #2: find -iname "*.jpg"
                                                   [User: 461.2 ms, System: 777.0 ms]
  Time (mean \pm \sigma):
                          1.253 s ±
  Range (min ... max):
                          1.233 s ...
                                                   10 runs
Summary
  'fd -e jpg -uu' ran
    3.80 ± 0.05 times faster than 'find -iname "*.jpg"'
```

Figure 1: An example picture of Hyperfines output comparing fd and find [27]

Firstly, Hyperfine supports out of the box statistical analysis and outlier detection. Since it can assume that the program run times are approximately equal, benchmark times are normal distributed. Thus, by fitting a normal distribution over all runs and computing its confidence interval, it can detect any outliers. Secondly, it allows for warmup runs and cache-clearing commands ¹ between each run. Warmup runs are useful to fill caches such as the page cache for disk I/O. Lastly, it supports further analysis by providing an export to various formats, such as CSV, JSON, Markdown or AsciiDoc, which can then be analyzed programmatically. Hyperfines repository contains several python scripts for basic visualization [29], which can be used as a starting point for further analysis.

2.3 Performance Optimization

For the benchmarking, the aforementioned criterion.rs benchmarking framework is used. The benchmark was done on a Dell Latitude 7420 with an Intel i5-1145G7 and 16GB of LPDDR4 RAM, compiled with rustc 1.72.0. The CPU idle was around 1%.

¹such as echo 1 > /proc/sys/vm/drop_caches to free the page cache [28].

2.3.1 Call By Reference

Although the data part of Vec<> is stored on the heap² the stack-part is still very complex, since it has to keep track of several things such as the current size and its current maximal capacity. More importantly, the Vec<> struct has a bigger memory footprint than a pointer. Thus, using Call-By-Reference should improve the performance by requiring less memory copies! In Rust, this can be archived using the & operator:

```
fn matmul(a: &Vec<Vec<f32>>, b: &Vec<Vec<f32>>) -> Vec<Vec<f32>> {
    /* Only the signature changes... */
}
fn driver_code(a: Vec<Vec<f32>>, b: Vec<Vec<f32>>, c: Vec<Vec<f32>>)
    -> Vec<Vec<f32>> {
    matmul(matmul(a, b), c) // D := A * B * C
}
```

Listing 2: Changing the signature to Call-By-Reference semantics with references.

Using the default criterion settings ³, the following results were benchmarked:

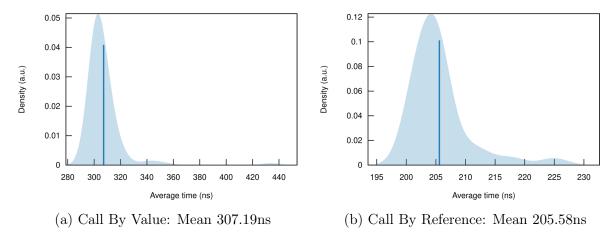


Figure 2: Comparsions of the PDFs computed using KDE for Call-By-Value and Call-By-Reference.

According to the benchmarks, this change alone resulted in a 49.426% mean increase! Note that the full metric table is included in the appendix.

There is one more obvious possible improvement to try: In this code, Rust's Vec<>, a dynamically sized, heap allocated vector is used. This could be replaced with a normal C-type array.

²Since Vec<> are dynamically resizable, the absolute size can't be known at compile time.

³Default Rust Release build

2.3.2 Primitive Stack Arrays

Compared to static arrays, Vec<> has much overhead. Firstly, since its size is not known at compile time, it performs several run-time bounds checks ⁴. Next, it has to be heap-allocated, which can be way more expensive and results in worse memory locality. Lastly, Vec<> is a complex struct with many functions and features, which in turn results in more computation required.

This is the code using primitive stack arrays instead of the more sophisticated, dynamically allocated vectors:

```
fn matmul(a: &[[f32; 3]; 3], b: &[[f32; 3]; 3], result: &mut [ [f32; 3]; 3]) {
1
        for i in 0..3 {
2
            for j in 0..3 {
3
                for k in 0..3 {
                    result[i][j] += a[i][k] * b[k][j];
5
                }
6
            }
        }
   }
9
10
   fn driver_code(a: &[[f32; 3]; 3], b: &[[f32; 3]; 3], c: &[[f32; 3]; 3],
11
                   res_buf: &mut [[f32; 3]; 3]) {
12
        let mut temp = [[0.0; 3]; 3];
13
        matmul3(a, b, &mut temp);
        matmul3(&temp, c, res_buf);
15
   }
16
```

Listing 3: Changing the signature to Call-By-Reference semantics with references.

Here are the results, compared to the initial version:

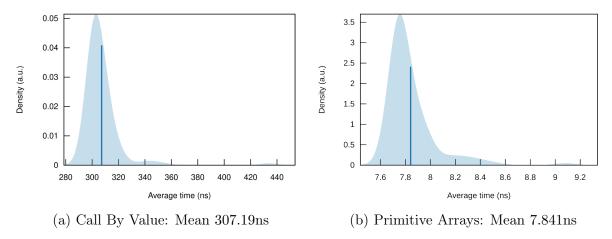


Figure 3: Comparsions of the PDFs computed using KDE for the initial version and the one using primitive arrays.

⁴This can be partially avoided. For more information, see the bounds checks cookbook [30]

This results in a staggering 3831.27% mean increase! Once again, note that the full metric table is included in the appendix.

There are several possible explainations for those results. It could be that the bounds check ruin the instruction pipelining. But the main reason ist most likely that the prior version requires an expensive heap allocation for the return value while the static arrays are already preallocated on the stack ⁵. Now that we did all high-level optimizations, the next step would be to optimize on the assembly level. The next section will show how to use proper tooling for assembly level optimization in Rust.

2.4 Assembly Optimizations

Modern compilers, such as the LLVM based rustc, do a lot of optimization for performance. This results in vastly different assembly when comparing unoptimized code (-00) to their optimized counter part (-03). Thus, it often makes sense to look at the assembly for hot code paths⁶. Naively, one could just look at the whole binary and decode the bytes into their instructions. This is neither useful nor reasonable for large programs to analyze. Instead, in this section, two different ways to analyze assembly will be analyzed: The well known Compiler Explorer [31] as well as the cargo-show-asm crate [32].

Compiler Explorer Compiler Explorer [31] is an online development environment initially developed by Matt Godbolt, primarily used for analysis of C and C++ applications. It was started in 2023 for optimizing financial quantitative analysis algorithms. Compiler Explorer supports over 30 different languages; from typical high performance languages such as C, C++ and Fortran, bytecode languages such as Python and Java to more niche languages such as Haskell and Solidity. Additionally, one can compare different compilers (for example gcc, clang and msvc for C applications) and manually specify different compiler arguments such as -Osize instead of -O3. Since it can also be hosted on-premise, propriatary and other custom compilers can be added if needed.

Its main feature is the color coding; Compiler Explorer assigns each function line to a specific color. The same color will then be used in the assembly window, providing an intuitive mapping and an good overall user experience. Unfortunately, it does not support multiple files and hasn't any dependency management ⁷. To summarize, Compiler Explorer is the best fit for small, single file programs. For larger applications, the next tool can be used.

cargo-show-asm cargo-show-asm [32] is a more minimalist, less polished console application for analyzing assembly. It works with any rust code base, no matter the size or amount of dependencies. Instead of showing the assembler for all functions, one can query single functions as an CLI parameter. Lastly, instead of the architecture specific assembler code it can also return the LLVM IR instead.

⁵Further analysis could be done through statistical profiling, which will be explained later. But since it doesn't help explaining performance engineering concepts, it is left as an exercise to the reader.

⁶Code paths are 'hot' when they are executed very frequently, therefore crucial for the overall performance.

⁷Although one could work around this restriction by installing all dependencies globally on a self hosted instance.

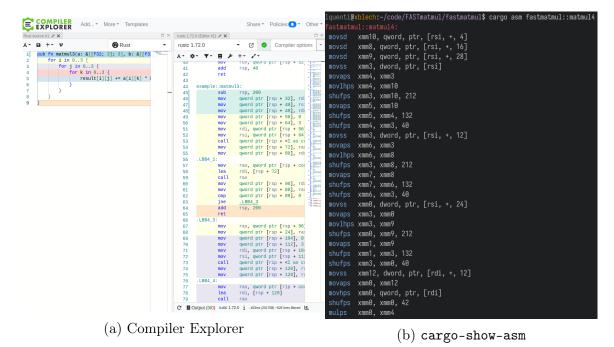


Figure 4: Two applications to analyze Rust assembly: Compiler explorer and cargo-show-asm.

Next, two common compiler optimizations are covered: Loop Unrolling and Function inlining.

2.4.1 Loop Unrolling

On the surface, loop unrolling is a simple concept: If the number of loop iterations are known at compile time, replicate the inner code that amount of time. This has multiple benefits. Firstly, it reduces the number of comparisons and jumps made in every iteration. Secondly, it allows for easier pipelining since there is no need for path prediction anymore! The main drawback is that code duplication obviously increases binary size.

Looking at the assembly, loop unrolling was already applied in our case. But if the compiler did not unroll the function, there are two main ways to force it manually:

- Unroll [33] is a Rust macro to unroll the applications at compile time by replacing the unrolled rust code in preprocessing. Currently, it can just detect loops with integer literal bounds.
- For more sophisticated loops, LLVM can be configured to more agressively apply loop unrolling. This is controlled by the -unroll-threshold parameter⁸.

As an important remark, do not manually apply loop unrolling without benchmarking, as it can worsen performance by consuming too much of the L1 CPU instruction cache.

2.4.2 Function Inlining

The next, very common compiler optimization is function inlining. Instead of jumping into a subroutine, executing it, and then returning to the previous instructions it places

⁸So to apply it with rustc, use -C llvm-args=-unroll-threshold=N where N is an integer.

the assembly of the subroutine into the outer function. On the upside, this eliminates the calling overhead (such as moving arguments into registers) specified by the calling convention ⁹. As already mentioned above, function inlining can also worsen performance though increased binary size and consequently less efficient cache utilization.

In our case, it was not applied. But it can be applied manually with the #[inline(/always/never)] attribute [34].

After covering benchmarking and assembly analysis, we can now approach the harder problem of variadic size matrix multiplication!

3 Variadic Size Matrix Multplication

Now, after having a simplified toy problem, lets say you get the following task:

"Our department has built a deep learning framework in Rust that inferences too slow. Please try to optimize the overall performance of this project"

Before being able to optimize the code, one has to first ask the following question: Why is it so slow? In order to answer this question, profiling is needed.

3.1 Profiling

Profiling is used to find out which parts of the program are executed frequently enough to effect runtime performance¹⁰. Since Rust produces normal binaries, most traditional profilers just work, including common ones such as Linux perf [35] and cachegrind [36]. See the profiling section in thr Rust Performance Book for an more exhaustive list [37].

Since rust supports polymorphism Both ad-hoc polymorphism through traits and parametric polymorphism based on generics. name mangling occurs. As assembly labels have to be unique, name mangling is a technique used to map all polymorphic instances of a function to an unique assmbler label. If the profiler doesn't support unmangling natively, rustfilt can be used manually [38].

In this chapter, we use cargo-flamegraph [39] and later iai [40] for profiling.

⁹Interestingly, Rust does not have a default non-FFI calling convention for performance benefits. Instad, it requires all dependencies to be compiled with the same version, which is the canonical way when the package manager cargo is used.

¹⁰So called **hot paths**.

- 3.2 Cargo Flamegraph
- 3.3 Applying Previous Knowledge
- 3.4 Compiler Optimizations
- 3.5 Cache-oblivious Algorithms
- 3.6 Iai

4 A glimpse of (Inter-Node) Parallelism

- 4.1 SIMD
- 4.2 Multithreading
- 4.2.1 Rayon

5 Conclusion and Further Ressources

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A Work sharing

If you worked in a group, describe here how you distributed the work and the actual contributions of each peer.

A.1 Hans

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A.2 Peter

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B Code samples

This is part of the appendix...