





UNIVERSITÄT **HEIDELBERG** ZUKUNFT **SEIT 1386**

C++ Benchmarking

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Course Outline

Performance

- How to get good performance
- Why benchmarking is important

Micro-benchmarks

- Run a tiny snippet of code on real hardware
- Do it many times & see how long it takes

Profiling

- Run the entire code on real hardware, occasionally check what it's doing
- Non-deterministic statistical sampling of events, but on real hardware

Simulation

- Run entire code on a virtual machine & measure every single operation
- o Deterministic measurement of all events, but not real hardware



Performance



Motivation

- We use c++ because we need performance
 - Otherwise we could (should?) use a higher-level language, e.g. Python
- Writing fast code is simple
 - Need to use good algorithms for efficiency (do less work)
 - Need to use good data structures for speed (do work faster)
- Writing fast code is difficult
 - Often these two needs are in conflict with each other
 - So the answer to "what is the fastest way to do this?" is "it depends"
- Benchmarking can help
 - With the question "which method is fastest?"
 - With the question "where and why is my code slow?"



Algorithms

- Complexity
 - Count operations required to process N items
 - E.g. 2 N + 5 N log(N)
 - For large N, term with highest power of N dominates
 - "Big-O" algorithmic complexity is the highest power of N
 - E.g. O(N log(N))
- Memory
 - Required memory is also counted and expressed in Big-O notation
- Optimal algorithms
 - For many problems, can prove an algorithm is "optimal"
 - This means is has the same Big-O cost as the best possible algorithm



Algorithm goal: reduce complexity

- Generally we want the most efficient algorithm
 - I.e. the one with the smallest Big-O complexity
- Often there a multiple "optimal" algorithms
 - Different pros and cons depending on your data and your hardware
- Although sometimes a less efficient algorithm can be better
 - o if memory is a constraint, and there is a less efficient algorithm that requires less memory
 - o if the big-O coefficient is much smaller
 - E.g. unless N is very large, 1e9 * log(N) is larger than N
 - o if it allows for a much faster implementation on your hardware
 - E.g. more fast operations can be better than a few slow operations
 - o if it allows for parallelization over threads / cores / nodes
 - o etc

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Hardware

- CPU gets some data, operates on it:
- Get the data from
 - L1 cache (32kb): 1 ns
 - L2 cache (256kb): 3 ns
 - L3 cache (32mb): 20 ns
 - o RAM (32gb): 100 ns
 - SSD (1tb): 10000 ns
- Do the operation
 - CPU core cycle: 1 ns
 - SIMD: apply same instruction to multiple data in the same operation
 - Multi-core: a CPU typically has several cores
- CPU spends a lot of time idle, waiting for data

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Cache

Cache misses are very expensive

- L1 cache: 1 ns wait + 1 ns work
- O RAM: 100 ns wait + 1 ns work

Cache lines

- Cache data is organised in lines of 64 bytes
- CPU gets the whole cache line when it asks for an item of data

SIMD

- A single CPU instruction can operate on multiple items of data
- Eg AVX512 instructions can operate on all 64 bytes in a single operation

Pre-fetching

- CPU will also try to guess what data will be needed next
- If you are iterating linearly over contiguous data, will nearly always predict correctly



Hardware goal: avoid cache misses

- Spatial locality of data
 - Store data contiguously, and in the order in which it will be accessed
 - Also known as "Data-oriented design"
 - or old-fashioned C-style arrays of data
 - or SoA (struct-of-arrays) vs AoS (arrays-of-struct)
 - Why? So that it is more likely to already be in the cache
- Temporal locality of algorithms
 - When accessing the same data again, do it sooner rather than later
 - Why? So that it is more likely to still be in the cache



General Performance strategy

- For large problems
 - We care about the efficiency of our *algorithms*
 - We want to do less work
- For small problems
 - We care about cache friendly data structures
 - We want to do our work fast
- Sometimes a large problem can also be many small problems
 - Use an efficient algorithm to reduce a large problem to many small problems
 - Use a fast cache-friendly method to solve each small problem
- Algorithms and data structures are interconnected
 - But helpful to consider their impact separately
 - Helps to understand the tradeoffs you are making with your choices



Map implementation performance

- Binary Tree: O(log(N)) to find an item
 - Good algorithmic efficiency
 - Terrible cache locality
 - Use case: large N
- Unsorted Vector: O(N) linear traversal to find an item
 - Terrible algorithmic efficiency
 - Good cache locality
 - Use case: small N
- Hash table: O(1)
 - Excellent algorithmic efficiency
 - Although not guaranteed need a good hash function
 - Cache locality depends on implementation
 - Use case: large N



Micro-benchmarks



Micro-benchmarks

- Run a tiny snippet of code on real hardware
- Do it many times & see how long it takes
- Good to get accurate results for specific functions
- May not be so close to "real world" scenarios
- Have to take care compiler doesn't optimize away the supposed work!
- Not deterministic: but can repeat many times & measure statistical significance
- Different possible metrics
 - Average time: most common and most relevant for many use cases
 - Median time: less noisy than the average, not affected by outliers
 - Fastest time: e.g. if investigating peak performance
 - Slowest time: e.g. if latency / worst-case scenario is important



Micro-benchmarking tools

Google benchmark

- Widely used and actively maintained
- See also <u>Quick-bench</u> (online version a bit like compiler explorer)

Celero

Actively maintained

Nonius

Older, less/not actively maintained

Catch2

- Some integration of nonius to add benchmarks to unit tests
- Do it yourself solution
 - E.g. using std::chrono::high_resolution_clock
 - Not recommended



Google benchmark minimal example

```
#include <benchmark/benchmark.h>
static void bench_map(benchmark::State &state) {
    Map map; // initialize data
    for (auto _ : state) {
        map.find(keys[0]); // code to benchmark
BENCHMARK(bench_map);
BENCHMARK_MAIN();
```



Live coding time

```
git clone --recursive https://github.com/ssciwr/cpp-benchmarking.git
cd cpp-benchmarking
mkdir build
cd build
cmake -DCMAKE_BUILD_TYPE=Release ..
cmake --build .
ctest
./bench/bench1
```



```
for (auto _ : state) {
   map.find(keys[0]);
}
```

```
bench_small_map/1 0.000 ns 0.000 ns 1000000000 bench_small_map/2 0.000 ns 0.000 ns 1000000000 bench_small_map/4 0.000 ns 0.000 ns 1000000000
```

- Looks bogus: Zero run-time, no N-dependence
- Why? Compiler optimizes away the entire statement



```
for (auto _ : state) {
   benchmark::DoNotOptimize(map.find(keys[0]));
}
```

```
bench_small_map/1 0.990 ns 0.990 ns 691986485
bench_small_map/2 1.00 ns 1.00 ns 705385115
bench_small_map/4 0.989 ns 0.989 ns 703257489
```

- Force compiler to put result into register or memory
- Non-zero runtime, but still no N-dependence
- Why? Always looking for the first key!



```
for (auto _ : state) {
   benchmark::DoNotOptimize(map.find(keys[n/2 + 1]));
}
```

```
bench_small_map/1 1.14 ns 1.14 ns 600298793
bench_small_map/2 1.38 ns 1.38 ns 507900144
bench_small_map/4 2.51 ns 2.51 ns 278833095
```

- Force compiler to put result into register or memory
- Ask for a key roughly in the middle of the vector
- Get reasonable-looking N-dependence



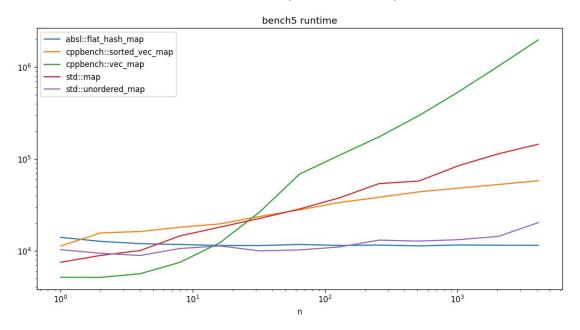
```
for (auto _ : state){
    for (auto key : keys) {
        benchmark::DoNotOptimize(map.find(key));
    }
}
```

```
bench_small_map/1 1.34 ns 1.34 ns 495155683
bench_small_map/2 3.21 ns 3.21 ns 218922642
bench_small_map/4 6.53 ns 6.53 ns 105903854
```

- Force compiler to put result into register or memory
- Loop over all keys: now doing N finds per benchmark iteration
- Get average performance over all inputs

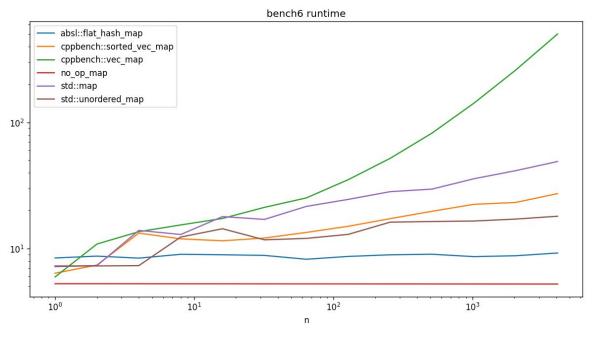


- Compare with std::map, std::unordered_map implementations
- Do the same number of find operations per benchmark iteration



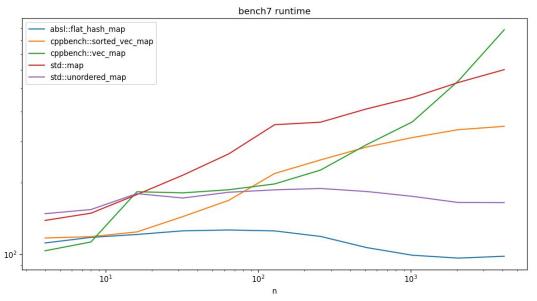


- Pick a random key at each iteration (more realistic, but adds overhead)
- Add a "no-op" Map to measure this overhead





- Try to simulate a "cold cache scenario"
 - Generate millions of identical maps (a few gig of RAM)
 - Pick a random map (probably not already in cache) for each benchmark iteration





Micro-benchmarking tips

- Compile in release mode
- Reduce the noise
 - Disable CPU turbo-boost / frequency scaling
 - Avoid running anything else at the same time
 - Repeat benchmarks to estimate noise, statistical significance of differences
- Try to be sure you're measuring what you intend to
 - Check the inputs and outputs are correct
 - Check the compiler is not optimizing away the work
 - Try to use or generate representative realistic input data
 - Compare timings with a no-op version of your operation / function
 - Look at the scaling as you vary the size of the input data
 - Try varying the order of the input data
 - Cold cache or hot cache scenario



Profiling



Profiling

- Run real production code on real hardware
- Occasionally check what it's doing
 - Not too often to avoid interfering with execution / slowing it down!
- Good to get an overview of real-world bottlenecks
- Not deterministic
- Only sample a tiny fraction of events
 - But likely to find hotspots this way



Profiling Tools

- Perf
 - No special instrumentation or re-compilation of code required
 - Relatively simple to use, pre-installed on most linux distributions
 - But linux only, and need sudo permissions:
 https://www.kernel.org/doc/html/latest/admin-guide/perf-security.html
- Gperftools
- Intel VTune
- Likwid
- Tracy
- and many more...



Compiling for profiling

- Generally want to compile in Release mode
 - Goal is to run your code exactly as in real-world setting
- Additional flags to improve profile output at minimal cost:
 - o -g: include debugging symbols (e.g. function names)
 - CMAKE_BUILD_TYPE = RelWithDebInfo
 - o -fno-omit-frame-pointer : don't omit frame pointers (i.e. better stack traces)
- Any additional requirements for your profiler
 - Link to their library?
 - o Include their header?
 - Annotate your code?
 - Set environment variables?



Perf stat

- perf stat ./app
 - default performance counters
- perf list
 - show all available counters
- perf stat -e L1-dcache-load-misses,L1-dcache-load ./app
 - o how often do we have a L1 data cache miss
- perf stat -e branches,branch-misses ./app
 - how often do we have a branch prediction miss



Perf stat example

```
perf stat -e L1-dcache-load-misses,L1-dcache-load \
   ./bench/bench6 \
   --benchmark_filter=".*:vec_map.*/16\$" \
   --benchmark_min_time=2
```

- First line: Measure cache misses using perf stat
- Others: run n=16 vec_map benchmark from bench6 for at least 2 secs

```
627,014 L1-dcache-load-misses # 0.01% of all L1-dcache accesses 8,377,940,732 L1-dcache-load
```



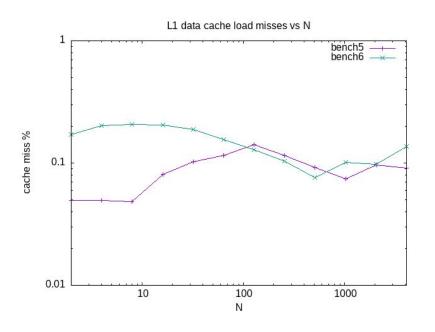
Branch prediction

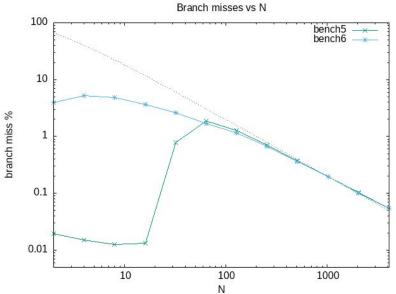
- CPU constructs a pipeline of instructions
 - Often "out of order" i.e. not evaluated in the order you wrote them
 - E.g. part of iteration n+1 in a loop may be calculated in iteration n
 - You need to know what instructions are coming next to be able to do this
- At every branch (e.g. if statement) in the code
 - CPU doesn't know where execution will jump to next until condition is evaluated
 - This stalls the pipeline
- So it tries to guess the most likely outcome (branch prediction)
 - Constructs the pipeline based on guessing where the branch will jump to
- Sometimes it guesses wrong
 - This is called a "branch miss" (similar to a cache miss)
 - CPU has to throw away / undo a lot of potentially incorrect operations
 - Then find and execute the correct instructions from the correct branch
 - This is expensive (also similar to a cache miss)



Perf branch prediction

• Branch predictor is outsmarting our benchmark in bench5





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Perf record

- perf record ./app && perf report
 - see annotated assembly with % of time for each line
 - h: show shortcuts
 - H: go to hotspot
 - Enter: go to function / follow jump

```
51.47 1d0: cmp %ecx,(%rbx,%rsi,4)

0.16 ↑ je 1a0

2.94 inc %rsi

0.01 cmp %rsi,%rax

41.29 ↑ jne 1d0

↑ jmp 1a7
```

./bench/bench5 --benchmark_filter=".*:vec_map.*/1024\\$" --benchmark_min_time=2



Perf visualization

- Raw assembly can be overwhelming
- Can be easier to interpret a call graph, flame graph, etc
- Options (all inconvenient in different ways)
 - https://github.com/KDAB/hotspot
 - Works directly on perf.data, but output seems incomplete / wrong
 - https://github.com/brendangregg/FlameGraph
 - Use supplied perl script to convert trace data first
 - https://github.com/google/pprof
 - But also need https://github.com/google/perf_data_converter
 - https://www.eclipse.org/tracecompass
 - But you need to re-compile perf with CTF support



Simulation



Simulation

- Run real production code on a virtual machine
- Record every cpu operation, every cache read, etc
 - Doesn't affect (simulated) speed of operations
- Deterministic
 - No statistical variation, not affected by other running processes
- Samples all events
- But: not real hardware
- Also: typically vastly slower than real-time (e.g. 50x slower)

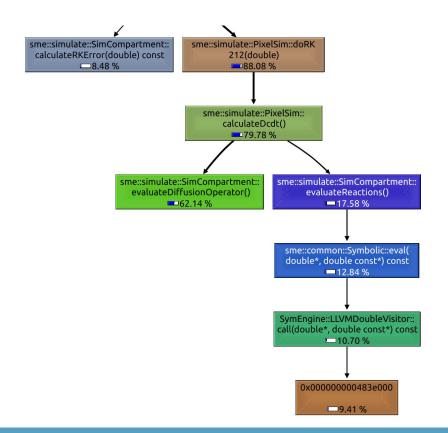


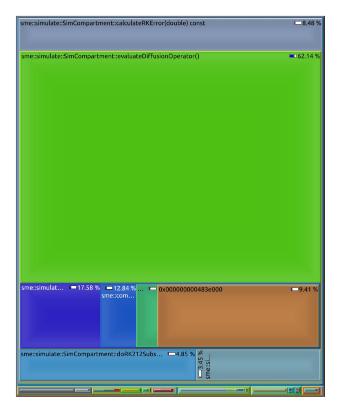
Callgrind/cachegrind

- Part of valgrind set of tools
 - Default tool is memcheck, but there are others!
- valgrind –tool=cachegrind ./app
 - Command line output of simulated cache hits/misses
- valgrind –tool=callgrind --simulate-cache=yes ./app
 - Generates call-graph data and cache hits/misses
- kcachegrind
 - GUI to visualize above data



Example call graph







Perf report for same code

```
spatial-cli spatial-cli
                                              sme::simulate::SimCompartment::evaluateDiffusionOperator
       spatial-cli spatial-cli
                                             sme::simulate::SimCompartment::calculateRKError
      spatial-cli spatial-cli
                                              sme::simulate::SimCompartment::evaluateReactions
      spatial-cli spatial-cli
                                              sme::simulate::SimCompartment::doRK212Substep2
       spatial-cli spatial-cli
                                              sme::simulate::SimCompartment::doRK212Substep1
      spatial-cli
                    [JIT] tid 263317
                                              0x00007fe92ef69035
      spatial-cli
                    [JIT] tid 263317
                                              0x00007fe92ef69004
       spatial-cli
                   spatial-cli
                                              SymEngine::LLVMDoubleVisitor::call
      spatial-cli spatial-cli
                                              sme::common::Symbolic::eval
      spatial-cli
                    [JIT] tid 263317
                                              0x00007fe92ef6904b
      spatial-cli
                         tid 263317
                                              0x00007fe92ef6901b
      spatial-cli
                         tid 263317
                                              0x00007fe92ef6900c
       spatial-cli
                         tid 263317
                                              0x00007fe92ef69031
      spatial-cli
                    [JIT] tid 263317
                                              0x00007fe92ef69047
       spatial-cli
                    [JIT] tid 263317
                                              0x00007fe92ef6901f
0.29%
       spatial-cli
                    libc.so.6
                                                      strtod l internal
      spatial-cli
                    [JIT] tid 263317
                                              0x00007fe92ef69039
                                              0x00007fe92ef69023
      spatial-cli
                    [JIT] tid 263317
                                              sme::simulate::SimCompartment::undoRKStep
      spatial-cli spatial-cli
```



Summary



Summary

- Benchmarking is vital but also difficult to do reliably
- Important not to blindly take one metric or benchmark as the "truth"
- We covered three complementary approaches
 - Micro-benchmarking
 - Profiling
 - Simulation
- Each have their own strengths and weaknesses
- Together they allow you to understand the performance of your code
- Also where and how to improve the performance of your code



Recommended resources

Benchmarking

- Chandler Carruth talks on benchmarking/performance
 - o Tuning C++: Benchmarks, and CPUs, and Compilers! Oh My!
 - <u>Efficiency with Algorithms, Performance with Data Structures</u>
 - Going Nowhere Faster
- Fedor Pikus talk on branchless programming
 - Branchless Programming in C++



Recommended resources

Performance

- Mike Acton talk on data oriented design
 - <u>Data-Oriented Design and C++</u>
- Ulrich Drepper paper on memory and cache
 - What Every Programmer Should Know About Memory
 - Not for "every programmer" but excellent info on RAM / cache / performance
 - From 2007 but still very relevant
- A curated list of c++ performance-related resources
 - AwesomePerfCpp



More recommended resources

Hash maps

- Malte Skarupke talk on hash map implementations and performance
 - You Can Do Better than std::unordered_map
- Matt Kulukundis talks on Google's hash map
 - Designing a Fast, Efficient, Cache-friendly Hash Table, Step by Step
 - Abseil's Open Source Hashtables: 2 Years In