

703308 VO High-Performance Computing WS2022/2023 Debugging Parallel Programs

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Overview

functional debugging

- generic guidelines
- serial debugging
- parallelism-specific debugging

performance debugging

- generic guidelines
- serial debugging
- parallelism-specific debugging

Motivation: First Ariane-5 Flight, 1996

Debugging Ariane-5's maiden flight

- booster nozzles adjusted for a 20% angle of attack, causing vehicle separation, triggering selfdestruction
- because a diagnostic bit pattern was sent to onboard computer as flight data
- because a data conversion from 64 bit float to 16 bit signed int overflowed
- because they used Ariane 4 inertial code for Ariane 5, which has larger possible value ranges (velocities)
 - and only protected 4 out of 7 critical variables against overflow and not this one because in Ariane 4 velocities were lower
 - also, this software component serves no purpose after lift-off (only used for inertial platform alignment before launch)



https://www.youtube.com/watch?v=gp D8r-2hwk

Motivation

- Why do we need debugging?
 - Because we make mistakes!
- Why do we need a lecture about this?
 - OpenMPI FAQ "Debugging applications in parallel", first question:
 - Q: "How do I debug OpenMPI processes in parallel?"
 - A: "This is a difficult question. [...] This FAQ section does not provide any definite solutions to debugging in parallel. [...]"

Functional Debugging

Functional debugging

- Dealing with everything that results in not getting the correct program output
 - program crashes
 - program not finishing (deadlocks, infinite loops)
 - incorrect output
- errors can be deterministic or non-deterministic

- ▶ all that applies to debugging serial programs is <u>crucial</u> for parallel ones
 - If you can't trust the serial implementation, why would you in a parallel context?

Coding guidelines

- write clean code that prevents bugs or facilitates their detection, e.g.
 - use meaningful identifiers
 - minimize vertical distance of variables
 - don't use OpenMP's private
 - follow the <u>Don't</u> <u>Repeat</u> <u>Yourself</u> (DRY) principle (single component per feature)
 - ...

- ▶ The toolchain you must use!
 - read & heed compiler warnings
 - write and regularly run unit and/or integration tests, especially aimed at (varying degrees of) parallelism
 - use code coverage tests
 - use continuous integration
 - use source version control

"Best of" real commit messages encountered in the past

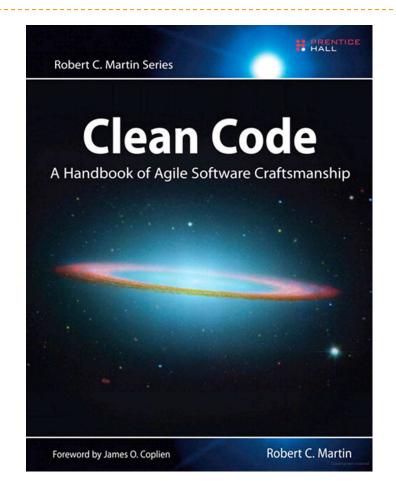
- stufF
- manager stuff
- more manager stuff
- Make things work
- ::w:qMerge branch 'master'
- dl;adlwa

- Added performance fix for DataItemManager::get() by caching fragment result in reference
- Removed debug print statement
- Fixed a linking issue of the unwrap_tuple function
- Redirected runtime system output to error stream
- fixing typos

Recommended reading/reference material

- "Clean Code" by Robert Martin, Prentice Hall 2008
 - ISBN 9780132350884
 - also available in German

naming, functions, commenting, formatting, data structures, error handling, unit tests, classes, concurrency, refinement & refactoring, ...



Generic debugging guidelines

- create a <u>Minimal</u> <u>Working Example (MWE)</u>
 - minimize problem size
 - minimize software components/features involved
 - ensure/increase reproducibility (e.g. fix random seeds, CPU affinity, ...)
 - if parallel
 - minimize machine size (number of threads and/or ranks)
 - minimize complexity of parallel interaction (e.g. communication patterns, ...)
- minimizes debugging feedback cycles times, amount of memory to inspect, amount of code to consider, overall degree of complexity of component & parallel interaction
 - sounds simple, but don't underestimate this
 - every change along the way to an MWE gives you more information about the problem

Serial debuggers

gdb

- useful for inspecting memory contents and getting call stacks
- can work with multi-threaded programs and also MPI
 - ▶ mpiexec -n X gdb -ex 'run' -ex 'bt' -ex 'quit' ./a.out
- can be used to debug e.g. a single MPI process among many
 - ▶ mpiexec -n 1 gdb ./a.out : -n 7 ./a.out
- can be attached to already-running processes
 - ▶ gdb -pid 12345

valgrind

- mostly used for finding memory leaks (can also simulate cache or generate call graph)
- can work with multi-threaded programs (but only concurrent execution!)
- can yield some false positives e.g. for OpenMP related to thread-local storage

Sanitizers (still mostly serial)

- tools that instrument code at compile time to perform checks at runtime
 - often lower overhead compared to external tools such as valgrind
 - if in doubt, check same issue with multiple tools (e.g. address sanitizers of multiple compilers and valgrind)
- depending on compiler, several sanitizers available, e.g.
 - address: buffer overflows, use-after-free, stack corruption, etc.
 - undefined behavior: signed integer overflow, float division by zero, negative shift operands, etc.
 - thread: detects data races
 - leak: detects memory leaks

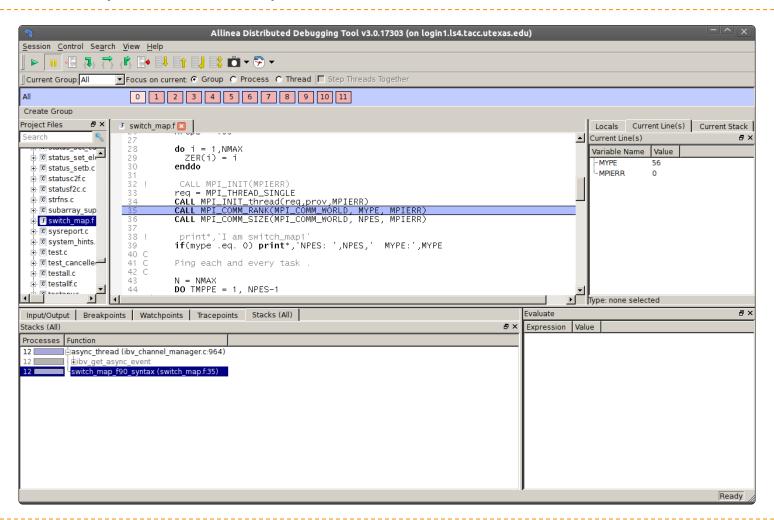
Call graph generators

- many tools available for generating call graphs
- static (at compile time)
 - doxygen, opt (Ilvm), cflow (gcc), etc.
- dynamic (at runtime)
 - gprof, callgrind, OpenPAT, pprof, CodeAnalyst, etc.
 - most performance analysis tools offer some form of call graph generation

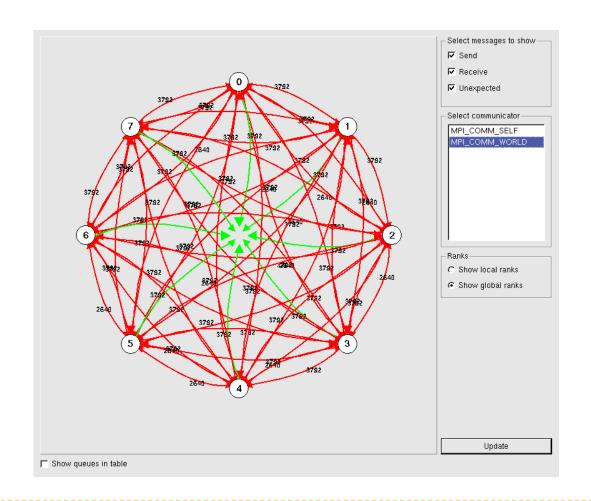
Parallel debuggers

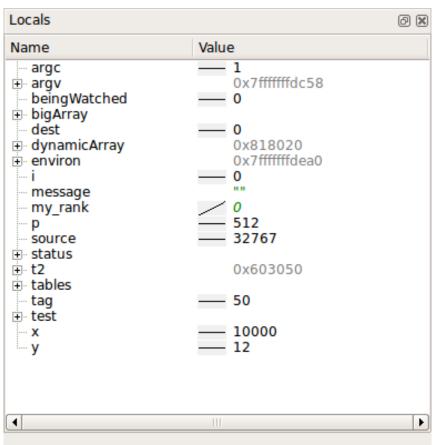
- very little free software
- two commercial top dogs: DDT (ARM) and TotalView (Rogue Wave Software)
- support OpenMP, MPI, CUDA, etc.
- several features centered around parallelism
 - examine variables per rank/thread, examine send/receive queues of MPI libraries, etc.
 - still, limited usefulness

DDT screenshot (overview)



DDT screenshots (communication patterns, data across ranks)





Automatic race condition debugging

difficult to do automatically and exactly

- statically detecting race conditions is NP-hard
- dynamically detecting race conditions incurs large runtime overhead (every memory access and synchronization action must be logged and checked)

most solutions resort to heuristics

- several experimental tools available in research
- many issues: limited scope, only apply to a subset of programming language, etc.
- few "mature" tools, e.g. Intel Inspector

Intel Inspector

features

- free
- Linux & Windows version
- automatically finds bugs in multithreaded programs
 - deadlocks
 - memory corruption
 - race conditions
 - vulnerabilities
- supports OpenMP, TBB, Pthreads, Windows threads

limitations & issues

- slowdown by 1-2 orders of magnitude!
- explicit support only for Intel OpenMP runtime
- error detection only at runtime, only in executed control flow branches
- false positives and negatives possible

OpenMP data race example 1

```
int counter = 0;

#pragma omp parallel for
for(int i = 0; i < 10; ++i) {
   counter++;
}</pre>
```

Description 🔺	Source	Function	Module
Read	ConsoleApplication1.cpp:9	main	consoleapplication1.exe
7 8	#pragma omp parallel for (int $i = 0$; $i < 10$		
9	counter++;		
10	}		
11			
Write	ConsoleApplication1.cpp:9	main	consoleapplication1.exe
7	#pragma omp parallel f	for	
8	for (int $i = 0$; $i < 10$); ++i) {	
9	counter++;		
10	}		
11			

OpenMP data race example 2

```
int sum = 0;
#pragma omp parallel for
for(int i = 0; i < 10; i++) {
 int tmp = sum;
 tmp = tmp + 1;
  sum = tmp;
```

Description 🔺	Source	Function	Module
Read	ConsoleApplication1.cpp:17	main	consoleapplication1.exe
15 16	#pragma omp parallel for (int $i = 0$; $i < 10$		
17 18	<pre>int tmp = sum; tmp = tmp + 1;</pre>		
19	sum = tmp;		
Write	ConsoleApplication1.cpp:19	main	consoleapplication 1.exe
17 18	int tmp = sum; tmp = tmp + 1;		
19	sum = tmp;		
20 21	}		

OpenMP data race example 2: wrong fix

```
int sum = 0;
#pragma omp parallel for
for(int i = 0; i < 10; i++) {
  int tmp;
  #pragma omp critical
  tmp = sum;
  tmp = tmp + 1;
  #pragma omp critical
  sum = tmp;
```

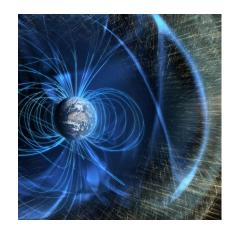


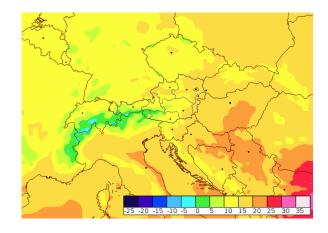
(not detected by Intel Inspector 2020)

Domain-specific debugging

- Visualize the output using appropriate tools
 - gnuplot
 - matplotlib
 - ParaView
 - ...
- note that this usually prohibits automatic checking
 - whenever feasible, unit and integration tests are preferred







Best approach to debugging parallel programs

- know your algorithm and implementation
 - e.g. "a structured grid problem with 2 ghost cells in every direction"
- know your programming models and languages, and their semantics
 - "reading the send buffer during an MPI_Isend(...) is illegal"
 - "this C++ object's destructor will be called at the end of the full-expression"
- Don't trust (seemingly) automatic analysis tools too much, read and understand the source code when available!

Performance Debugging

Performance debugging

- also sometimes known as "non-functional" debugging (not related to functional output)
 - short execution time not necessarily but most often the only goal
 - much more tricky than functional debugging
 - ▶ How do you know the performance bug was fixed? Because it's "faster" now?
- most aspects of functional debugging or sequential programs still apply
 - coding guidelines & best practice
 - + reproducibility (e.g. fix random seeds, scheduling affinities, ...)
 - + if required, performance unit tests, performance regression checks
 - + performance tools (the ones for sequential programs can also be useful)
 - + a lot more knowledge about hardware required

(h)top

- Don't underestimate the power of top or htop!
- Get a high-level overview of the workload on the system (and it's components) and compare to what you expected!
 - What's the ratio between user time and system time?
 - ▶ high system time could be caused by inefficient I/O, high amount of context switching, etc.
 - Which CPU cores am I really using?
 - the only way to verify affinity policies
 - What is the actual memory footprint vs. what it should be?
 - detect e.g. memory leaks or inefficient memory management without any additional analysis tools

htop & affinity

- > 2x Intel E5-2699 v3 (18 cores per CPU) in a single node
- ▶ htop shows cores 1-18 and 37-54 busy, hence 36 cores total right?



Recap: perf

```
[c703429@login.lcc2 ~]$ perf stat ./heat stencil 1D seq
                       # 2.471 GHz
28,826,239,136 cycles:u
35,220,856,783 instructions:u # 1.22 insn per cycle
6,711,849,029 branches:u # 575.356 M/sec
    1,295,209 branch-misses:u # 0.02% of all branches
        1,044 LLC-load-misses:u
           26 LLC-store-misses:u
   15,312,122 L1-dcache-load-misses:u
  476,440,489 L1-dcache-store-misses:u
```

Terminology

instrumentation

- add source/machine code that will measure something when executed
- can happen manually, automatically, during compilation, linking, runtime, ...
- do not confuse with "measurement"
- inclusive/exclusive measurements
 - do measurements include data for nested code regions (e.g. functions)?

```
int outside() {
  for(int i = 0; i < N; ++i) {
      // work
  }
  inside();
  for(int j = 0; j < M; ++j) {
      // more work
  }
}</pre>
```

More terminology: sample- vs. trace-based profiling

Sampling

- gives aggregated information of how much time spent where in the code
- based on statistics: does not provide information on the order of events, their time interval or exact numbers
- easy to accomplish, comparatively low overhead, no code changes required
 - stop program periodically and read program counter register of the CPU core
 - build histogram at the end

Tracing

- produces a detailed log of which event happened at what point in time
- allows to establish order of events, even across processes/nodes if clocks are in sync
- requires code changes/instrumentation

```
e.g. wrap every function call with
get_timestamp();
func_call();
get_timestamp();
```

gprof

- sample-based profiler
 - also limited code instrumenter for call graph generation and call counts
 - very simplistic, not always accurate
- available with every GCC installation
- very simple in its use
 - compile with debug symbols (-g) and gprof support (-pg)
 - run binary as usual
 - run gprof binary gmon.out to view results
 - use --line to get more detailed, line-based results

gprof example

```
int foo() {
 long long counter = 0;
 #pragma omp parallel for
  for(int i = 0; i < N; ++i) {
   #pragma omp critical
    counter++;
  return counter;
```

```
int bar() {
  long long partSum[MAX_NUM_THREADS][8];
  long long counter = 0;
  #pragma omp parallel
    int tid = omp get thread num();
    partSum[tid][0] = 0;
    #pragma omp for
    for(int i=0; i<N; ++i) partSum[tid][0]++;</pre>
    #pragma omp critical
    counter += partSum[tid][0];
  return counter;
```

gprof example cont'd

```
Flat profile:
Each sample counts as 0.01 seconds.
     cumulative self
                      self
                                      total
       seconds seconds calls Ts/call Ts/call name
time
                                            foo._omp_fn.0 (test.c:13 @ 400a3d)
100.71
           0.02
                  0.02
                                       0.00 bar (test.c:19 @ 40092c)
 0.00
           0.02
                  0.00 	 1 	 0.00
                                        0.00 foo (test.c:8 @ 4008e6)
           0.02
                  0.00
 0.00
                                0.00
```

perf record/report

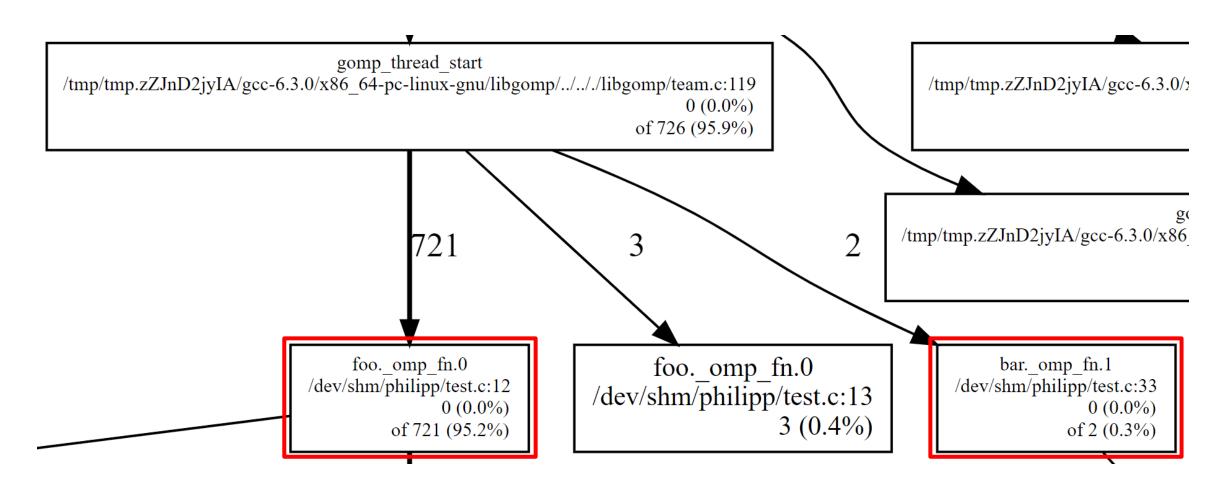
- Perf also supports profiling
 - perf record ./application
 - perf report
- Supports time as well as performance counters
- Also records a lot of platform information
 - perf report --header-only
 - ▶ CPU type, OS version, date & time of record, etc.
 - perf report --header-only -I
 - includes NUMA topology, cache info, etc.

```
$ perf record -e cache-misses ./02_matrix_mul
Samples: 262K of event 'cache-misses:u', Event
count (approx.): 1072111977
Overhead Symbol
                Source:Line
         [.] main 02_matrix_mul.c:68
         [.] main 02_matrix_mul.c:69
         [.] vfprintf vfprintf+15272
  0.01%
         [.] vfprintf vfprintf+15262
  0.01%
  0.00%
         [.] vfprintf vfprintf+15292
         [.] vfprintf vfprintf+15255
  0.00%
```

gperftools

- sample-based profiler
 - formerly Google Performance Tools
- actually a collection of performance analysis tools and high-performance multithreaded memory allocators
- very simple in its use
 - install gperftools library
 - ▶ link with -lprofiler
 - run with environment variable CPUPROFILE=prof.out
 - run pprof binary prof.out to view results (--gv for graphical visualization)

gperftools example



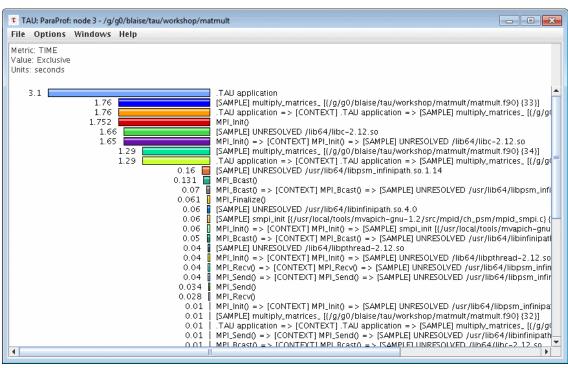
Performance analysis tools for parallel programs

profiling and analysis software

- Intel Pin: dynamic binary instrumentation
- Intel VTune: performance analysis for multi-threaded programs
- Intel Advisor: dependency, vectorization and cache analysis tool
- AMD CodeXL / NVIDIA Nsight: profiler and debugger for GPUs
- ▶ TAU: profiling and tracing suite
- ▶ HPCToolkit: profiling and tracing suite, includes GUI tools and tools for mapping binary code to source code
- ▶ PAPI: library for access to hardware performance counters
- OProfile: sampling-based profiler with hardware performance counter support
- ▶ also, some software built into your IDE, e.g. MS Visual Studio
- analysis and visualization/reporting tools
 - Scalasca, Vampir, Paraver, JumpShot, paraprof, CUBE, etc.
- These lists are by far not complete!

TAU & ParaProf





Tau instrumentation files

- allows to control the scope of instrumentation
 - reduces measurement overhead
 - reduces collected data to relevant content

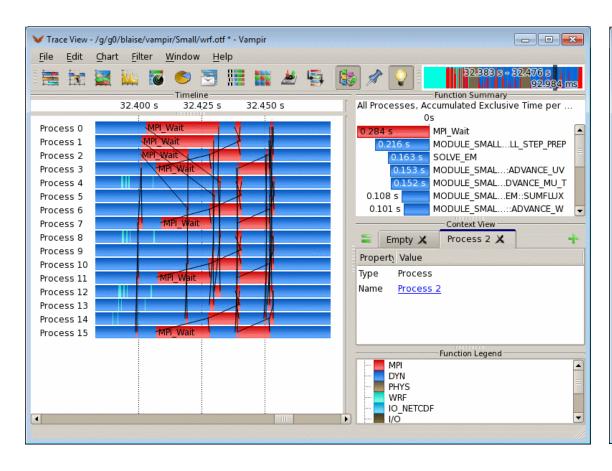
allows to select function patterns, loops, code lines, etc.

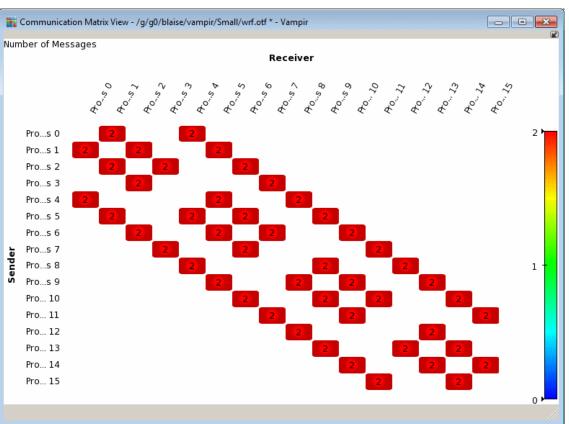
```
BEGIN_INSTRUMENT_SECTION

loops file="foo.cpp" routine="int bar(int*, double)"

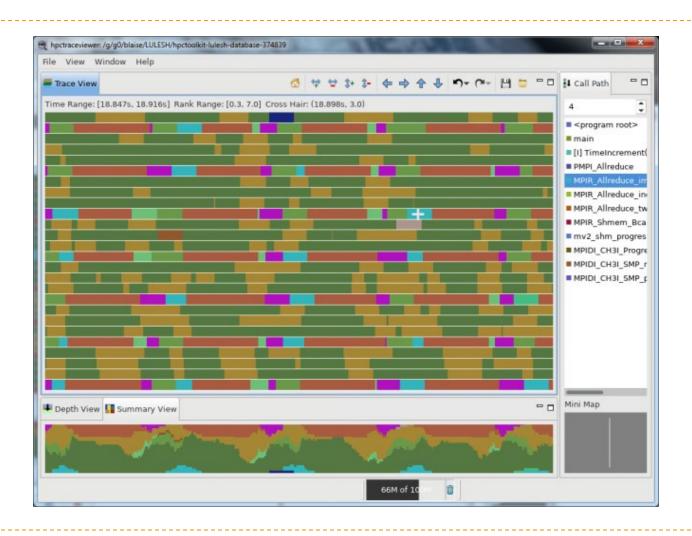
END_INSTRUMENT_SECTION
```

Vampir

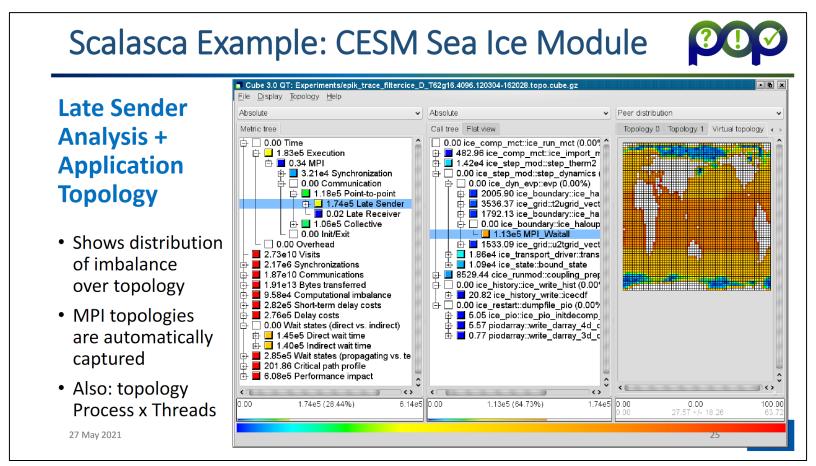




HPCToolkit



Cube (from Scalasca suite)



Source: Bernd Mohr, https://pop-coe.eu/sites/default/files/pop-files/pop-webinar-scalasca.pdf

General hints when working with debuggers

- –g when compiling if source code locations are required
- careful with optimization flags, especially -0#
 - function inlining, loop fusion/fission, ...
 - likely to obfuscate source code locations
 - ▶ if required and feasible, work in -00 or temporarily disable conflicting flags
 - ▶ careful: -00 performance != -01/2/3/... performance
- check whether child processes are included in analysis/reports
- check whether threads are included in analysis/reports
- ▶ if tracing or otherwise large-overhead instrumentation required, restrict to code regions of interest

Angles of attack in order of benefit

1/0

 file formats, buffering, distributed I/O, ram disks, data structures

Network

• comm. patterns, non-blocking & one-sided comm., topology mappings, load balancing

Memory

 data structures, NUMA & affinity, cache optimizations (e.g. tiling, alignment, padding)

Computation vectorization, data types, intrinsics, load balancing, hardware changes

Summary

functional debugging

- adhere to coding guidelines and best practice of software engineering
- especially relevant for parallelism: know your programming models and semantics, don't trust automatic tools blindly

performance debugging

- don't underestimate the power of simple tools
- many more advanced tools out there, but not straight-forward to use
- know your hardware and your program hotspots

Image sources

- Yoda: https://www.deviantart.com/biggiepoppa/art/Master-Yoda-Star-Wars-395511111
- DDT: https://portal.tacc.utexas.edu/software/ddt, https://developer.arm.com/docs/101136/latest/ddt/viewing-variables-and-data
- Domain-specific debugging: https://twitter.com/maven2mars/status/984440044659159040, https://twitter.com/maven2mars/status/984440044659159040, https://twitter.com/maven2mars/status/984440044659159040, https://twitter.com/maven2mars/status/984440044659159040, https://twit
- ► TAU & ParaProf: https://hpc.llnl.gov/software/development-environment-software/tau-tuning-and-analysis-utilities
- Vampir: https://hpc.llnl.gov/software/development-environment-software/vampir-vampir-server
- ▶ HPCToolkit: https://hpc.llnl.gov/software/development-environment-software/hpc-toolkit