

High performance tools to debug, profile, and analyze your applications

# Application Trapped Capacity and Energy Reporting

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**VP Product Management** 









# Application Trapped Capacity and Energy Reporting



**Today**: Application-centric performance tuning and energy metrics















# **allinea**The problem we solve...

From climate modeling to astrophysics, from financial modeling to engine design, the power of clusters and supercomputers advances the frontiers of knowledge and delivers results for industry. Writing and deploying software that exploits that computing power is a



How well do your applications exploit your hardware? Allinea Performance Reports are the most effective way to characterize and understand the performance of HPC application runs



Allinea Forge is the complete tool suite for software development - with everything needed to debug, profile, optimize, edit and build applications for high performance.

# Allinea provides high-performance software tools

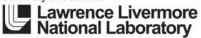








University of California

































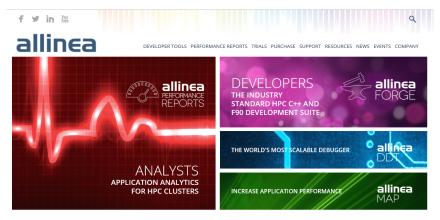












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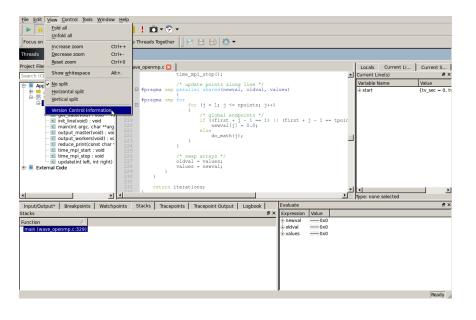


How well do your applications exploit your hardware?
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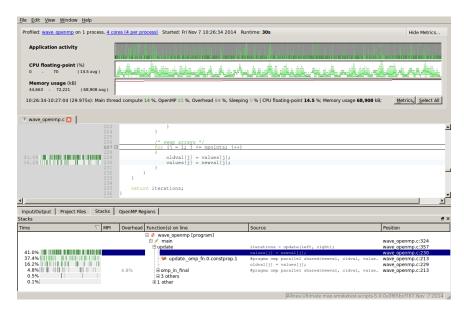


Allinea Forge is the complete tool suite for software development - with everything needed to debug, profile, optimize, edit and build applications for high nerformance.

# Allinea Forge Debug and profile codes











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# allinea The problem we solve...

software that exploits that computing power is a

From climate modeling to astrophysics, from financial modeling to engine design, the power of clusters and supercomputers advances the frontiers of knowledge



How well do your applications exploit your hardware? Allinea Performance Reports are the most effective way to characterize and understand the performance

Allinea Forge is the complete tool suite for software development - with everything needed to debug, profile, optimize, edit and build applications for high

# **Performance** Reports Monitor and tune applications

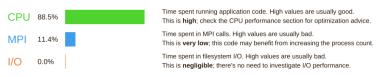


mpiexec -n 4 ./wave\_c 8000 4 processes, 1 node (4 physical, 8 logical cores per node) Fri Oct 17 17:00:27 2014 30 seconds (1 minute)



2.1 Ghz CPU frequency

## Summary: wave c is CPU-bound in this configuration



This application run was CPU-bound. A breakdown of this time and advice for investigating further is in the CPU section below. As very little time is spent in MPI calls, this code may also benefit from running at larger scales.

### CPU

### A breakdown of the 88.5% CPU time: Single-core code 100.0% Scalar numeric ops 22.4% Vector numeric ops 0.0% Memory accesses 77.6%

The per-core performance is memory-bound. Use a profiler to identify time-consuming loops and check their cache performance.

No time is spent in vectorized instructions. Check the compiler's vectorization advice to see why key loops could not be vectorized

A breakdown of the 0.0% I/O time: Time in reads Time in writes Effective process read rate 0.00 bytes/s Effective process write rate 0.00 bytes/s

No time is spent in I/O operations. There's nothing to optimize here!

### Memory

Per-process memory usage may also affect scaling: Mean process memory usage 49.7 MB

Peak process memory usage 53.6 MB Peak node memory usage

The peak node memory usage is very low. You may be able to reduce the amount of allocation time used by running with fewer MPI processes and more data on each process.

24.0%

### MPI

A breakdown of the 11.4% MPI time: Time in collective calls 3.1% Time in point-to-point calls Effective process collective rate 31.7 kB/s

Effective process point-to-point rate 269 kB/s

Most of the time is spent in point-to-point calls with a very low transfer rate. This suggests load imbalance is causing synchonization overhead; use an MPI profiler to investigate further.

### Threads

A breakdown of how multiple threads were used:

Computation 0.0% 0.0% Synchronization Physical core utilization 100.0% Involuntary context switches per second

No measurable time is spent in multithreaded code

A breakdown of how the total 588 J energy was spent:

Accelerators 0.0% Peak power 23.00 W Mean power 19.80 W

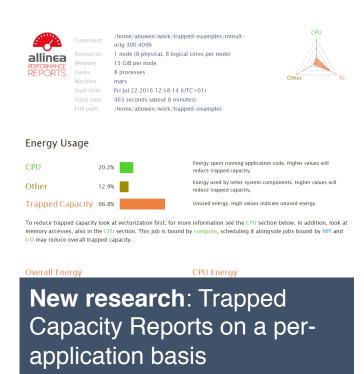
The CPU is responsible for all measured energy usage. Check the CPU breakdown section to see if it is being well-used.

Note: system-level measurements were not available on this run.

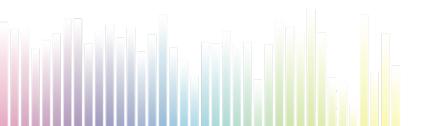
# Application Trapped Capacity and Energy Reporting



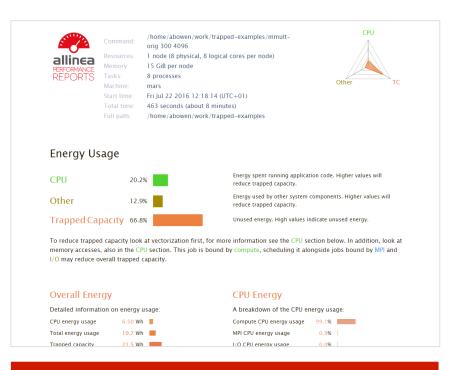
**Today**: Application-centric performance tuning and energy metrics



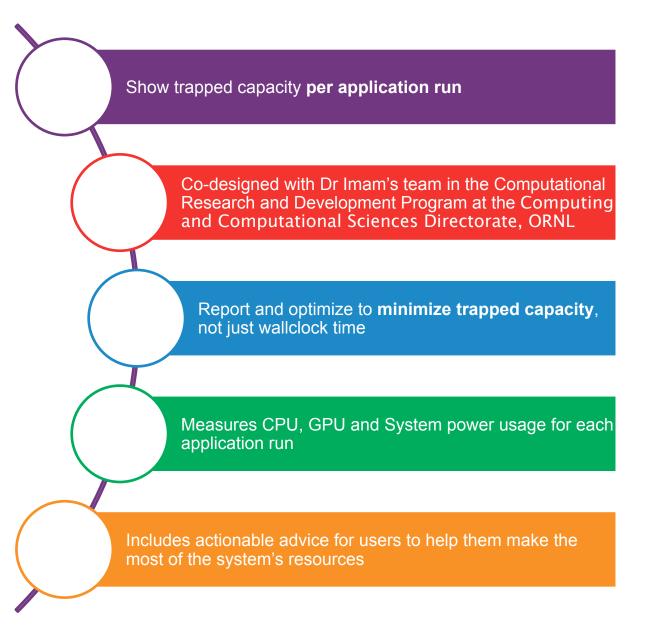








# Application-centric Trapped Capacity Reports





/home/abowen/work/trapped-examples/mmult-

orig 300 4096

Resources: 1 node (8 physical, 8 logical cores per node)

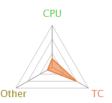
Memory: 15 GiB per node

Tasks: 8 processes

Machine: mars

Start time: Fri Jul 22 2016 12:18:14 (UTC+01)
Total time: 463 seconds (about 8 minutes)

Full path: /home/abowen/work/trapped-examples



# **Energy Usage**



To reduce trapped capacity look at vectorization first, for more information see the CPU section below. In addition, look at memory accesses, also in the CPU section. This job is bound by compute, scheduling it alongside jobs bound by MPI and I/O may reduce overall trapped capacity.

# Overall Energy

Detailed information on energy usage:

CPU energy usage	6.50 Wh	
Total energy usage	10.7 Wh	
Trapped capacity	21.5 Wh	
Total energy budget	32.1 Wh	
Mean node power budget	250 W	
Mean node power	82.8 W	
Peak node power	98.0 W	

The peak power draw is significantly higher than the mean, use a profiler to identify times when the power draw is low.

## **CPU Energy**

A breakdown of the CPU energy usage:

Compute CPU energy usage 99.1%

MPI CPU energy usage 0.9% |
I/O CPU energy usage 0.0% |
Mean power during compute 50.5 W

Mean power during MPI 54.0 W

Mean power during I/O 0.00 W |

Note: Power and energy values during MPI, I/O and compute are estimated from available information and may not be accurate.

### **CPU**

A breakdown of the 99.2% CPU time:

Scalar numeric ops 50.8%

Vector numeric ops 0.0% |

Memory accesses 49.2%

No time is spent in vectorized instructions. Vectorized instructions generally require more power than scalar computation. To reduce trapped capacity check the compiler's vectorization advice to see if key loops can be vectorized.

Significant time is spent on memory accesses. Memory operations generally require less power than computation. To reduce trapped capacity use a profiler to identify time-consuming loops and check their cache performance.

### Threads

A breakdown of how multiple threads were used:

Computation 0.0% |
Synchronization 0.0% |
Physical core utilization 100.0% 
System load 100.6%

### MPI

A breakdown of the 0.8% MPI time:

Time in collective calls

2.9% |

Time in point-to-point calls

97.1%

Effective process collective rate

0.00 bytes/s |

Effective process point-to-point rate

33.4 MB/s

Optimising MPI communication will have little impact on the trapped capacity.

### I/O

A breakdown of the 0.0% I/O time:

Time in reads 0.0% |

Time in writes 0.0% |

Effective process read rate 0.00 bytes/s |

Effective process write rate 0.00 bytes/s |

No time is spent in I/O, there's nothing to optimise here!

### Memory

Per-process memory usage may also affect scaling:

Mean process memory usage 210 MiB

Peak process memory usage 406 MiB

Peak node memory usage 14.0%



mars

orig 300 4096

15 GiR ner node

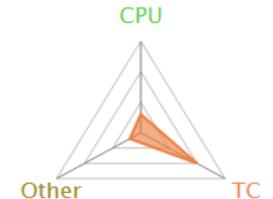
8 processes Fri Jul 22 2016 12:18:14 (UTC+01) 463 seconds (about 8 minutes) /home/abowen/work/trapped-examples

1 node (8 physical, 8 logical cores per node)

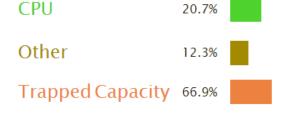
/home/abowen/work/trapped-examples/mmult-



# Trapped Capacity at a glance



# Actionable advice



To reduce trapped capacity look at vectoriz memory accesses, also in the CPU section. I/O may reduce overall trapped capacity.

## Detailed breakdown

# Overall Energy

Detailed information on energy usage:

CPU energy usage	6.50 Wh	
Total energy usage	10.7 Wh	
Trapped capacity	21.5 Wh	
Total energy budget	32.1 Wh	
Mean node power budget	250 W	
Mean node power	82.8 W	
Peak node power	98.0 W	

The peak power draw is significantly higher than the mean, profiler to identify times when the power draw is low.

# **Energy Usage**



Energy spent running application code. Higher values will reduce trapped capacity.

Energy used by other system components, Higher values will reduce trapped capacity.

Unused energy. High values indicate unused energy

To reduce trapped capacity look at vectorization first, for more information see the CPU section below. In addition, look at memory accesses, also in the CPU section. This job is bound by compute, scheduling it alongside jobs bound by MPI and I/O may reduce overall trapped capacity.

### Overall Energy



The peak power draw is significantly higher than the mean, use a profiler to identify times when the power draw is low

### CPU

A breakdown of the 99,2% CPU time Scalar numeric ops 50.8% Vector numeric ops 0.0% Memory accesses 49.2%

No time is spent in vectorized instructions. Vectorized instructions generally require more power than scalar computation. To reduce trapped capacity check the compiler's vectorization advice to see if key loops can be vectorized.

Significant time is spent on memory accesses. Memory operations generally require less power than computation. To reduce trapped capacity use a profiler to identify time-consuming loops and

### MPI

A breakdown of the 0.8% MPI time:

Time in collective calls	2.9%	i i
Time in point-to-point calls	97.1%	
Effective process collective rate	0.00 bytes/s	1
Effective process point-to-point rate	33.4 MB/s	

Optimising MPI communication will have little impact on the trapped capacity.

### CPU Energy



Note: Power and energy values during MPI, I/O and compute are estimated from available information and may not be accurate.

### Threads

A breakdown of how multiple threads were used: Computation 0.0% 0.0% Synchronization Physical core utilization 100.0% 100.6% System load

### I/O

A breakdown of the 0.0% I/O time:

Time in reads Effective process read rate 0.00 bytes/s Effective process write rate 0.00 bytes/s

No time is spent in I/O, there's nothing to optimise here!

# Track CPU and GPU energy use

# **CPU Energy**

A breakdown of the CPU energy usage:

Compute CPU energy usage	99.1%	
MPI CPU energy usage	0.9%	
I/O CPU energy usage	0.0%	
Mean power during compute	50.5 W	
Mean power during MPI	54.0 W	
Mean power during I/O	0.00 W	

Note: Power and energy values during MPI, I/O and estimated from available information and may not

# Identify inefficient applications

# **CPU**

A breakdown of the 99.2% CPU time:

Scalar numeric ops 50.8% Vector numeric ops 0.0% Memory accesses

No time is spent in vectorized instructions instructions generally require more power computation. To reduce trapped capacity vectorization advice to see if key loops ca Significant time is spent on memory acces

# I/O and MPI breakdowns

# I/O

Time in reads

A breakdown of the 0.0% I/O time:

Time in reads	0.070	1
Time in writes	0.0%	
Effective process read rate	0.00 bytes/s	
Effective process write rate	0.00 bytes/s	ı

0.0%

No time is spent in I/O, there's nothing to op

# Example 1: application binary not optimized for CPU architecture



Command: /home/abowen/work/trapped-examples/mmult-

orig 300 4096

Resources: 1 node (8 physical, 8 logical cores per node)

Memory: 15 GiB per node

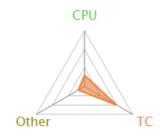
Tasks: 8 processes

Machine: mars

Start time: Fri Jul 22 2016 12:18:14 (UTC+01)

Total time: 463 seconds (about 8 minutes)

Full path: /home/abowen/work/trapped-examples



# **Energy Usage**

CPU	20.2%		ergy spent running application code. Higher values will uce trapped capacity.
Other	12.9%		ergy used by other system components. Higher values will uce trapped capacity.
Trapped Capacity	66.8%	Uni	used energy. High values indicate unused energy.

To reduce trapped capacity look at vectorization first, for more information see the CPU section below. In addition, look at memory accesses, also in the CPU section. This job is bound by compute, scheduling it alongside jobs bound by MPI and I/O may reduce overall trapped capacity.

# Overall Energy

Detailed information on energy usage:

CPU energy usage	6.50 Wh	
Total energy usage	10.7 Wh	
Trapped capacity	21.5 Wh	
Total energy budget	32.1 Wh	
Mean node power budget	250 W	
Mean node power	82.8 W	
Peak node power	98.0 W	

# **CPU**

A breakdown of the 99.2% CPU time:

Scalar numeric ops 50.8% Vector numeric ops 0.0% |

Memory accesses 49.2%

No time is spent in vectorized instructions. Vectorized instructions generally require more power than scalar computation. To reduce trapped capacity check the compiler's vectorization advice to see if key loops can be vectorized.

Significant time is spent on memory accesses. Memory operations generally require less power than computation. To reduce trapped capacity use a profiler to identify time-consuming loops and check their cache performance.

# Example 1: recompiled with architecture-specific optimizations



Command: /home/abowen/work/trapped-examples/mmult-vec

300 256

Resources: 1 node (8 physical, 8 logical cores per node)

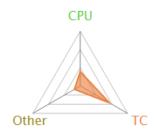
Memory: 15 GiB per node

Tasks: 8 processes

Machine: mars

Start time: Fri Jul 22 2016 15:19:11 (UTC+01)
Total time: 301 seconds (about 5 minutes)

Full path: /home/abowen/work/trapped-examples



# **Overall Energy**

Detailed information on energy usage:

CPU energy usage	5.43 Wh	
Total energy usage	7.97 Wh	
Trapped capacity	12.9 Wh	
Total energy budget	20.8 Wh	
Mean node power budget	250 W	
Mean node power	95.6 W	
Peak node power	98.0 W	

# Energy Usage

CPU 26.1% Energy spent running application code. Higher values will reduce trapped capacity.

Other 12.2% Energy used by other system components. Higher values will reduce trapped capacity.

Trapped Capacity 61.7% Unused energy. High values indicate unused energy.

To reduce trapped capacity look at memory accesses first, for more information see the CPU section below. In addition, look at vectorization, also in the CPU section. This job is bound by compute, scheduling it alongside jobs bound by MPI and I/O may reduce overall trapped capacity.

# **CPU**

A breakdown of the 97.2% CPU time:

Scalar numeric ops 17.3%

Vector numeric ops 37.8%

Memory accesses 44.9%

Significant time is spent on memory accesses. Memory operations generally require less power than computation. To reduce trapped capacity use a profiler to identify time-consuming loops and check their cache performance.

# Example 2: CPU idle due to inefficient OpenMP scheduling



Command: /home/abowen/work/trapped-examples/umult-

omp-static 300 1024

Resources: 1 node (8 physical, 8 logical cores per node)

Memory: 15 GiB per node

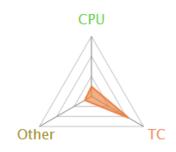
Tasks: 1 process, OMP\_NUM\_THREADS was 8

Machine: node-5

Start time: Fri Jul 22 2016 14:39:01 (UTC+01)

Total time: 301 seconds (about 5 minutes)

Full path: /home/abowen/work/trapped-examples



# Energy Usage



To reduce trapped capacity look at OpenMP overhead first, for more information see the OpenMP section below. In addition, look at vectorization, see the CPU section below. This job is bound by compute, scheduling it alongside jobs bound by MPI and I/O may reduce overall trapped capacity.

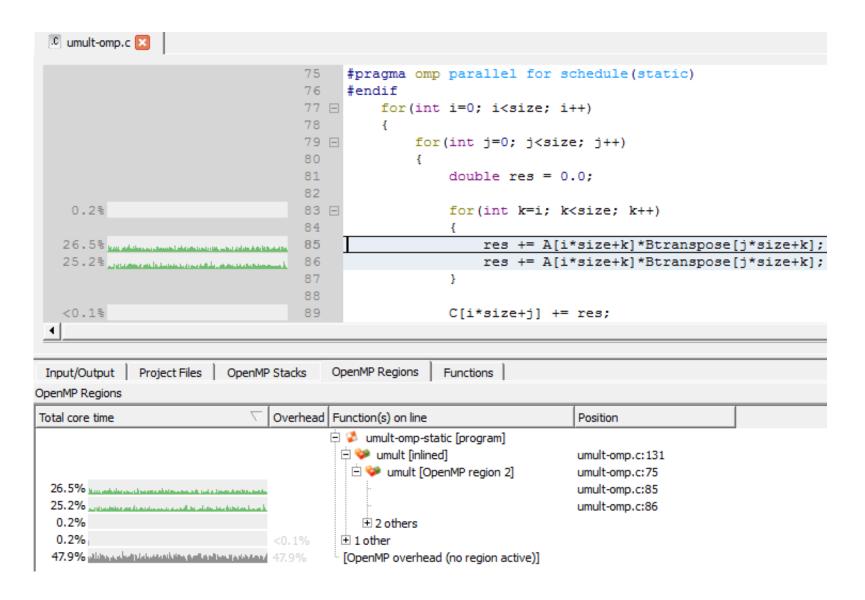
# OpenMP

A breakdown of the 100.0% time in OpenMP regions:

Computation	52.0%	
Synchronization	48.0%	
Physical core utilization	100.0%	
System load	55.4%	

Significant time is spent synchronizing threads. Sleeping CPU cores require minimal power. Check which locks cause the most overhead with a profiler and improve workload balance to reduce trapped capacity.

# Example 2: Allinea Forge identifies the inefficient loop





# Example 2: Dynamic scheduling reduces trapped capacity



/home/abowen/work/trapped-examples/umult-

omp-dynamic 300 1024

Resources: 1 node (8 physical, 8 logical cores per node)

Memory: 15 GiB per node

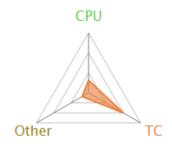
Tasks: 1 process, OMP\_NUM\_THREADS was 8

Machine: node-5

Start time: Fri Jul 22 2016 14:44:34 (UTC+01)

Total time: 301 seconds (about 5 minutes)

Full path: /home/abowen/work/trapped-examples



# OpenMP

A breakdown of the 100.0% time in OpenMP regions:

Computation 99.9%

Synchronization <0.1% |

Physical core utilization 100.0%

System load 100.1%

# **Energy Usage**

CPU 20.7% Energy spent running application code. Higher values will reduce trapped capacity.

Other 12.3% Energy used by other system components. Higher values will reduce trapped capacity.

Trapped Capacity 66.9% Unused energy. High values indicate unused energy.

To reduce trapped capacity look at vectorization first, for more information see the CPU section below. In addition, look at memory accesses, also in the CPU section. This job is bound by compute, scheduling it alongside jobs bound by MPI and I/O may reduce overall trapped capacity.

# **CPU**

A breakdown of the 100.0% CPU time:

Single-core code 0.0% |
OpenMP regions 100.0% |
Scalar numeric ops 85.1% |
Vector numeric ops 0.0% |
Memory accesses 14.9% |

No time is spent in vectorized instructions. Vectorized instructions generally require more power than scalar computation. To reduce trapped capacity check the compiler's vectorization advice to see if key loops can be vectorized.

# Example 3: GPU accelerators unused by application



/home/abowen/work/trapped-examples/gpu-

waiting-nogpu 300 12

1 node (12 physical, 24 logical cores per node)

2 GPUs per node available

Memory: 15 GiB per node, 11 GiB per GPU

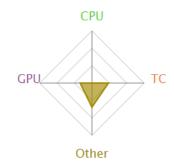
Tasks: 1 process

Machine: rhel-7-amd64

Start time: Wed Jul 20 2016 11:38:37 (UTC+01)

Total time: 307 seconds (about 5 minutes)

Full path: /home/abowen/work/trapped-examples



# **Energy Usage**

CPU	0.0%	Energy spent running application code. Higher values will reduce trapped capacity.
GPU	22.4%	Energy spent running accelerator code. Higher values will reduce trapped capacity.
Other	46.1%	Energy used by other system components. Higher values will reduce trapped capacity.
Trapped Capacity	31.5%	Unused energy. High values indicate unused energy.

CPU metrics are not supported (no intel\_rapl module)

To reduce trapped capacity look at accelerator usage first, for more information see the Accelerators section below. In addition, look at vectorization, see the CPU section below. This job is bound by compute, scheduling it alongside jobs bound by MPI, accelerator usage, and I/O may reduce overall trapped capacity.

# Accelerators

A breakdown of how accelerators were used:

GPU utilization 0.0% |
Global memory accesses 0.0% |
Mean GPU memory usage 0.2% |
Peak GPU memory usage 0.2% |

Accelerators are available but are not used. It may be more efficient to make use of accelerators or run on nodes without them.

# Example 3: Developer uses CUDA library. Are we done?



Command: /home/abowen/work/trapped-examples/gpu-

waiting 300 12

1 node (12 physical, 24 logical cores per node)

2 GPUs per node available

Memory: 15 GiB per node, 11 GiB per GPU

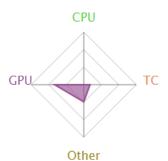
Tasks: 1 process

Machine: rhel-7-amd64

Start time: Wed Jul 20 2016 11:27:05 (UTC+01)

Total time: 301 seconds (about 5 minutes)

Full path: /home/abowen/work/trapped-examples



## **CPU**

A breakdown of the 100.0% CPU time:

Scalar numeric ops 0.0% |

Vector numeric ops 0.0% |

Memory accesses 0.0% |

Waiting for accelerators 99.9%

Most of the time is spent waiting for accelerators. Sleeping CPU cores require minimal power. To reduce trapped capacity use asynchronous calls to overlap CPU and accelerator workloads.

Significant time is spent on memory accesses. Cache-inefficient memory operations generally require less power than computation. To reduce trapped capacity use a profiler to identify time-consuming loops and check their cache performance.

# Energy Usage

CPU	0.0%	Energy spent running application code. Higher values will reduce trapped capacity.
GPU	55.5%	Energy spent running accelerator code. Higher values will reduce trapped capacity.
Other	32.6%	Energy used by other system components. Higher values will reduce trapped capacity.
Trapped Capacity	11.9%	Unused energy. High values indicate unused energy.

CPU metrics are not supported (no intel\_rapl module)

To reduce trapped capacity look at time spent waiting for accelerators first, for more information see the CPU section below. In addition, look at accelerator usage, see the Accelerators section below.

# **Accelerators**

A breakdown of how accelerators were used:

GPU utilization 48.6% Global memory accesses 13.7% Mean GPU memory usage 0.9% Peak GPU memory usage 0.9%

The accelerator usage is low. Increasing the accelerator usage could reduce the trapped capacity.

# Example 3: Expert helps user overlap CPU and GPU work



/home/abowen/work/trapped-examples/gpu-

waiting-async 300 12

1 node (12 physical, 24 logical cores per node)

2 GPUs per node available

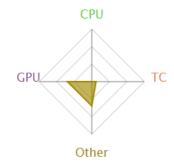
Memory: 15 GiB per node, 11 GiB per GPU

Tasks: 1 process

Machine: rhel-7-amd64
Start time: Wed Jul 20 2016 11:32:57 (UTC+01)

Total time: 304 seconds (about 5 minutes)

Full path: /home/abowen/work/trapped-examples



# **Energy Usage**

CPU	0.0%	Energy spent running application code. Higher values will reduce trapped capacity.
GPU	45.9%	Energy spent running accelerator code. Higher values will reduce trapped capacity.
Other	46.1%	Energy used by other system components. Higher values will reduce trapped capacity.
Trapped Capacity	8.0%	Unused energy. High values indicate unused energy.

CPU metrics are not supported (no intel\_rapl module)

To reduce trapped capacity look at accelerator usage first, for more information see the Accelerators section below. In addition, look at vectorization, see the CPU section below. This job is bound by compute, scheduling it alongside jobs bound by MPI, accelerator usage, and I/O may reduce overall trapped capacity.

# **CPU**

A breakdown of the 100.0% CPU time:

Scalar numeric ops 48.5%

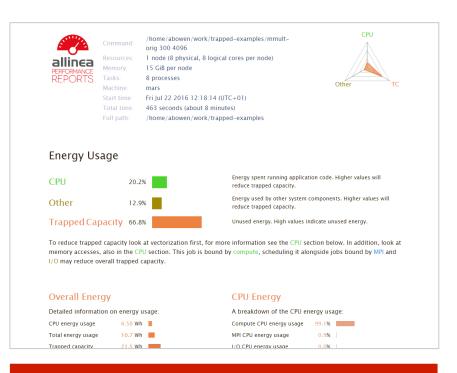
Vector numeric ops 0.0% |

Memory accesses 51.2%

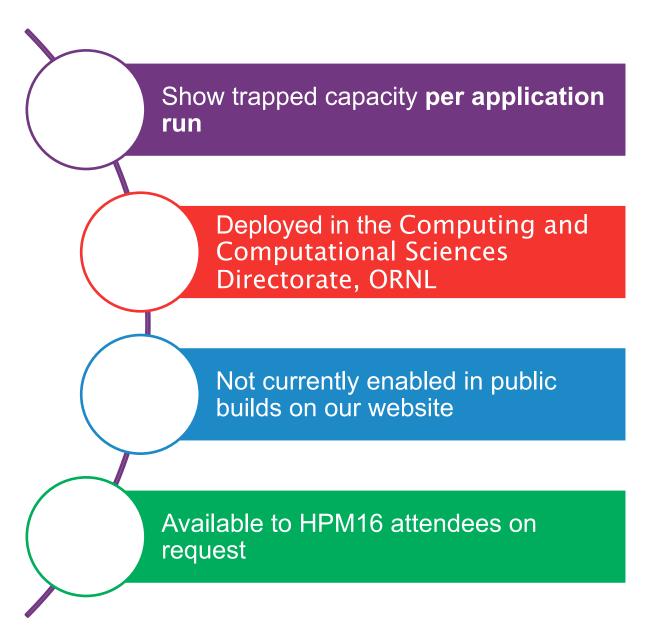
Waiting for accelerators 0.4% |

Significant time is spent on memory accesses. Cache-inefficient memory operations generally require less power than computation. To reduce trapped capacity use a profiler to identify time-consuming loops and check their cache performance.

No time is spent in vectorized instructions. Vectorized instructions generally require more power than scalar computation. To reduce trapped capacity check the compiler's vectorization advice to see if key loops can be vectorized.



# Application-centric Trapped Capacity Reports

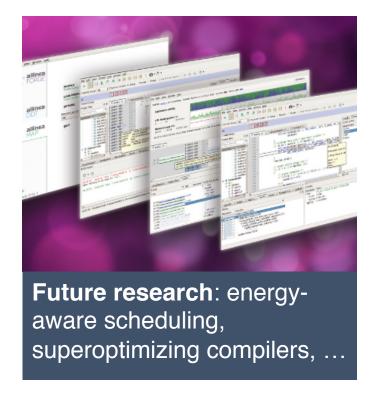


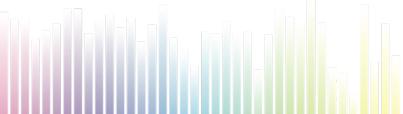
# Application Trapped Capacity and Energy Reporting



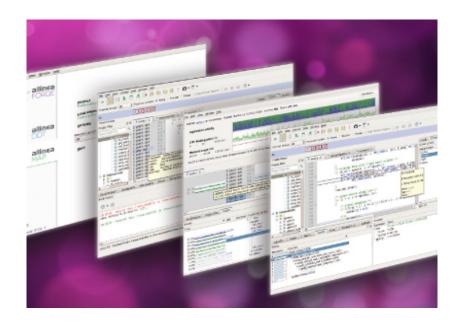
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# Active research

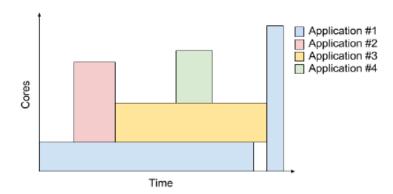
# Ongoing research:

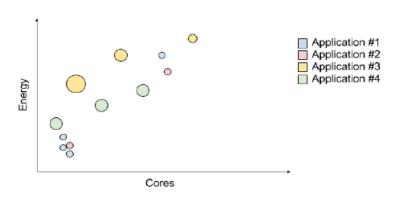
- Application Energy Dashboard
- Provide energy cost model for superoptimizing compiler
- Feed machine learning algorithm that chooses optimal compilation flags
- Data- and energy-aware scheduling via SLURM integration

# Total Software Energy Reporting and Optimization

# Application Energy Dashboard

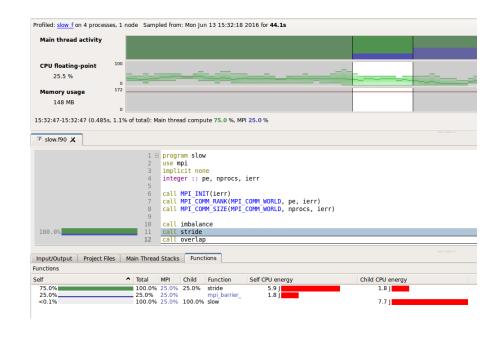
- Collect perapplication energy metrics
- Visualize application runs and energy usage over time
- Dive into particular applications for a Performance Report

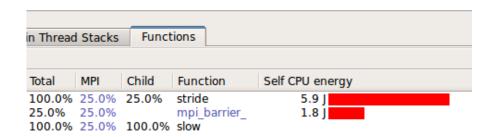






# Total Software Energy Reporting and Optimization





Superoptimizing compiler

- MAP provides function-level performance data to the compiler
- EMBECOSM's stochastic superoptimizer targets most energy-costly functions
- Reruns under MAP to measure impact of changes

Machine Guided Energy Efficient Compilation

- Building on existing MAGEEC project for embedded systems
- Uses machine learning to predict compiler option combinations that boost energy efficiency
- Extending to take additional data from MAP in the input vector

# Horizon 2020: Next Generation I/O for Exascale

Work
package
overview

Target: 50% reduction in energy to solution when compared to today's systems

Target: 20x improved I/O performance for real applications

**Performance Reports** will feed energy and I/O measurements into the SLURM workload scheduler

# Scheduler integration

Preload data onto nodes before job starts or schedule work close to existing data location

Schedule compute-intensive jobs on cooler parts of the system

Limit clock speed on jobs known to be memory-bound

# Application Trapped Capacity and Energy Reporting



**Today**: Application-centric performance tuning and energy metrics









High performance tools to debug, profile, and analyze your applications

# Technology wish list: open cross-vendor energy APIs (that do *not* require root access)



VP Product Management







