

Improving the Performance of Engineering Codes



Meet the POP CoE:
Federico Panichi, NAG Ltd



Who is NAG?

Renowned in Numerical Algorithms & HPC

Over the last 50 years, brands such as AMD, Arm, Intel, and Nvidia have become reliant on NAG IP



Global SME

With hubs in Asia, Europe and the US, NAG can support clients in their own time zone

NAG provides



Solutions

- NAG Library & custom algorithm development
- Fortran compiler
- Algorithmic differentiation



Consultancy

- Cloud HPC migration
- HPC technology evaluation & benchmarking
- Code porting & optimisation



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A Centre of Excellence in HPC

Performance Optimisation and Productivity

- Better Parallel Code
- Boost scalability of HPC Applications
 - Faster results, novel solutions, reduced expenditure
 - More efficient use of HPC infrastructures
 - In prep for Exascale computing
- A Team with
 - Excellence in performance tools and tuning
 - Excellence in programming models and practices
 - R & D background in real academic and industrial use cases



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What & who for



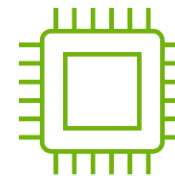
Methodology & Tools

- Quantitative picture of application behaviour
- Hierarchical set of Performance metrics
- Open-source tools that support high-scale analysis



Communities

- All domains of science & technology
- SMEs or large industries
- Academia & Research institutes
- All other HPC CoEs
- HPC Centres



Services

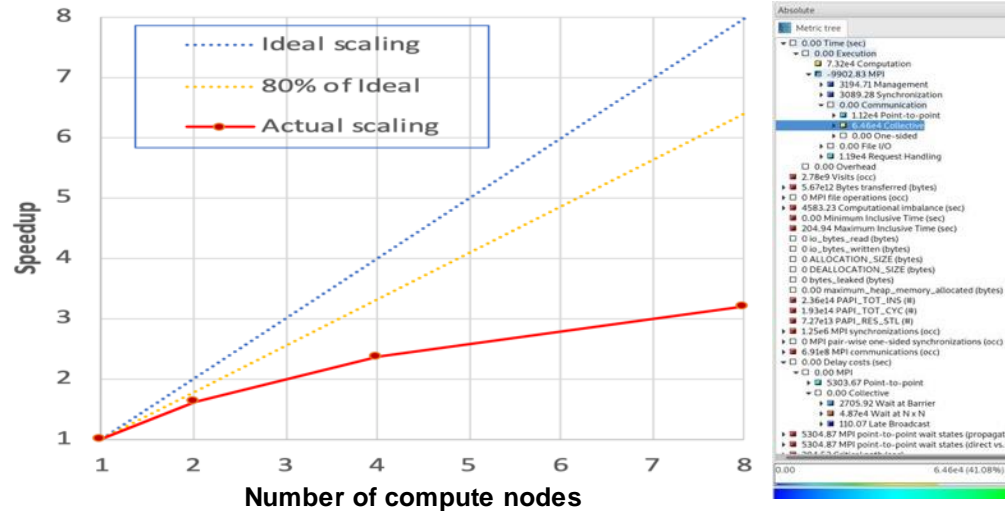
- Performance assessment
- Code Optimization
- Transversal across all programming models, architectures & scale
- Education & training



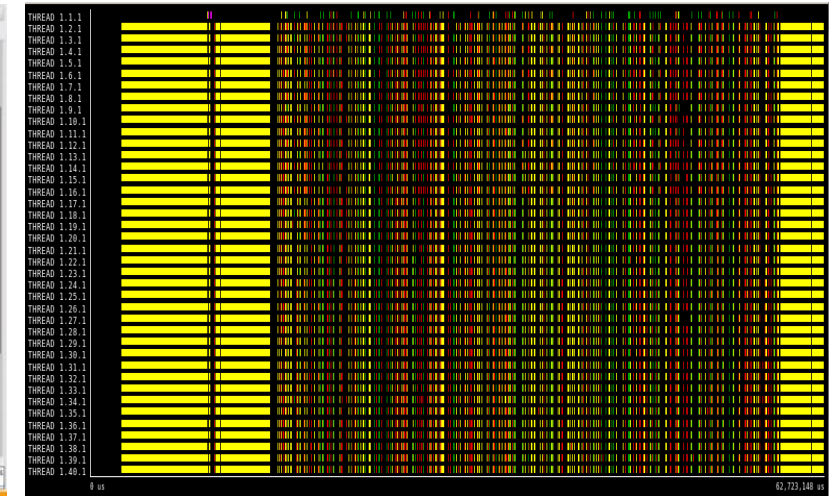
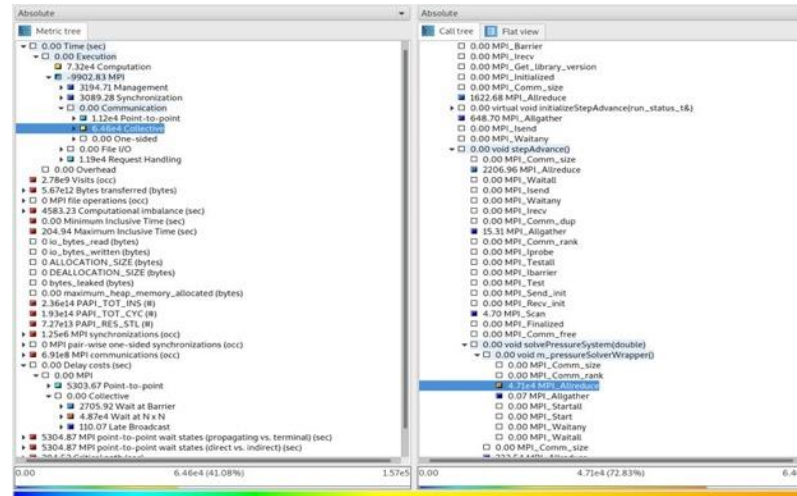
Parallel performance is hard to understand

How do we measure the performance of our parallel programs?

speed-up and efficiency plots?



collect data using performance monitoring tools?



*Cube, perf. metrics per routines/call stack,
data collected by Scalasca/Score-P*

*Paraver, timeline view of program
execution, data collected by Extrae*

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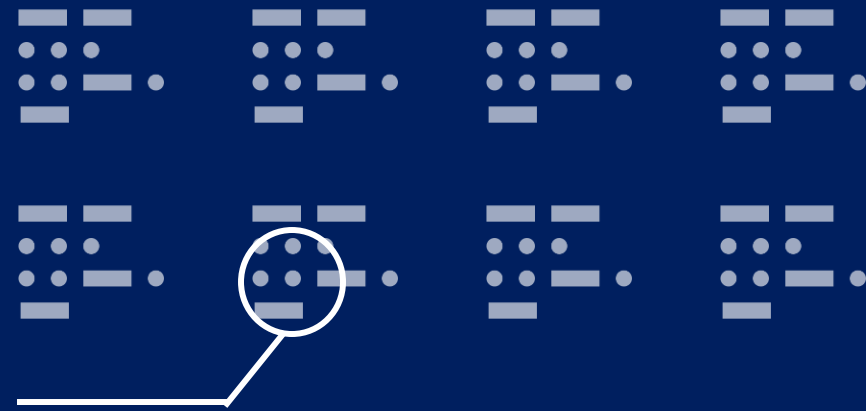




Tracing is powerful

But can generate overwhelming amount of data

Main Problem



Lack of quantitative understanding of the actual behaviour of a parallel application

Difficult to know where to start and what to look for



The POP metrics

A simple but powerful solution

Devise a simple set of performance metrics...

- using values easily obtained from the trace data
- where low values indicate specific causes of poor parallel performance

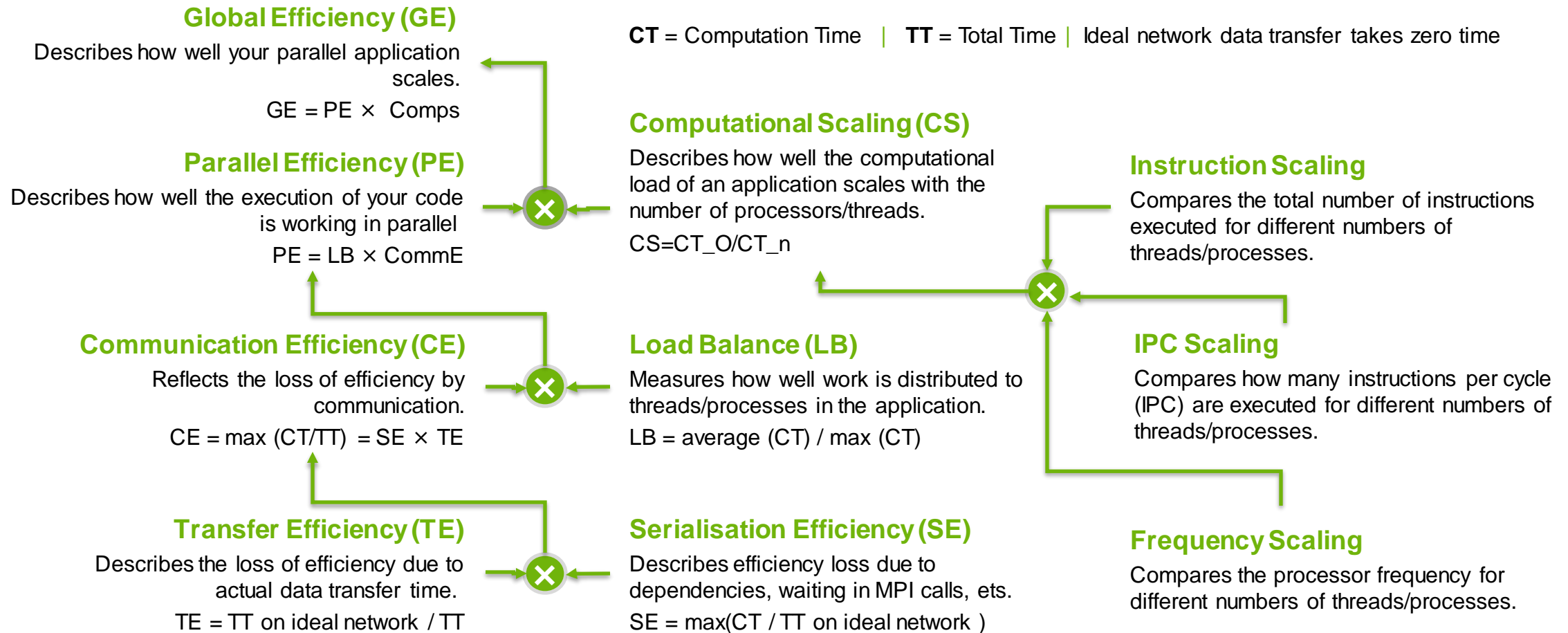
These metrics then are used to understand...

- what are the causes of poor performance
- what to look for in the trace data





POP MPI Parallel Efficiency Metrics



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POP performance monitoring tools

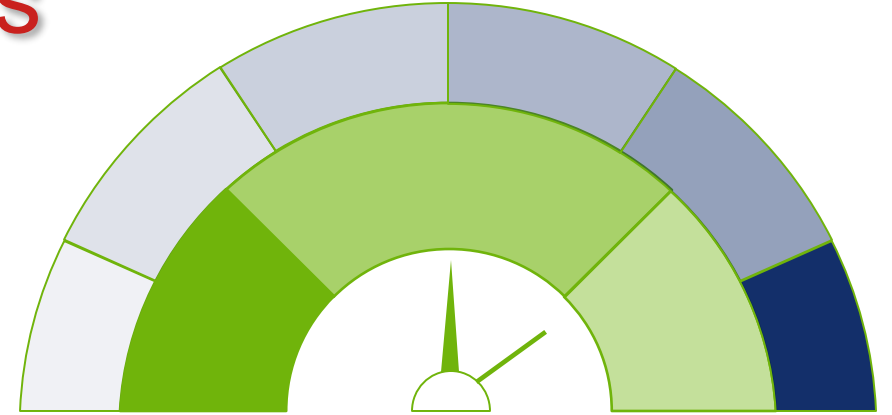
- Developing open-source tools

- Extrae (tracing), Paraver (visualisation) & Dimemas
<https://tools.bsc.es>
- Score-P (<https://score-p.org/> profiling and tracing), Scalasca (<https://scalasca.org/> Post Processing) & Cube (visualisation) <https://tools.bsc.es/>
- MAQAO: synthetic reports and hints with a focus on core performance <http://www.maqao.org>
- PyPOP: automated generation of POP metrics from Extrae traces
<https://github.com/numericalalgorithmsgroup/pypop>

For more help on how to use these tools and calculate the POP metrics see the POP website learning material & online training

<https://pop-coe.eu/further-information/learning-material> and <https://pop-coe.eu/further-information/online-training>

Other tools can also be used



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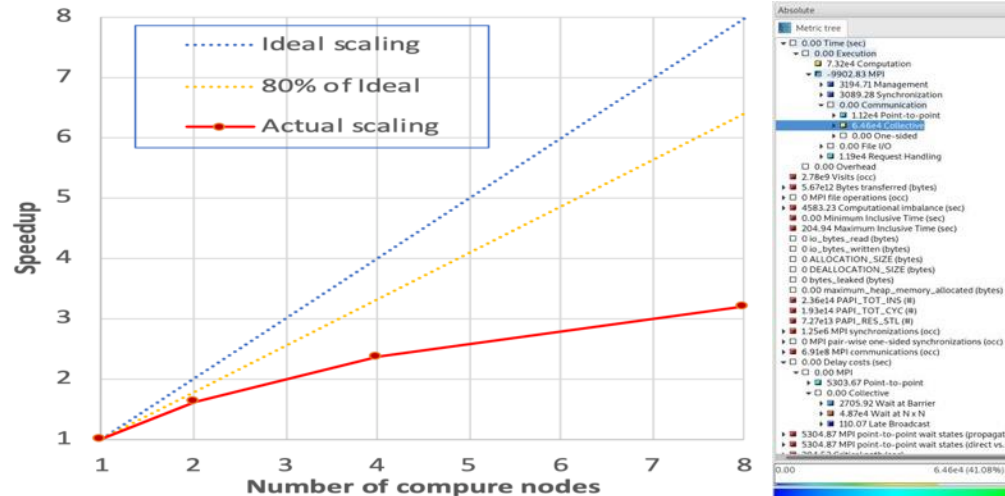


Parallel performance is hard to understand

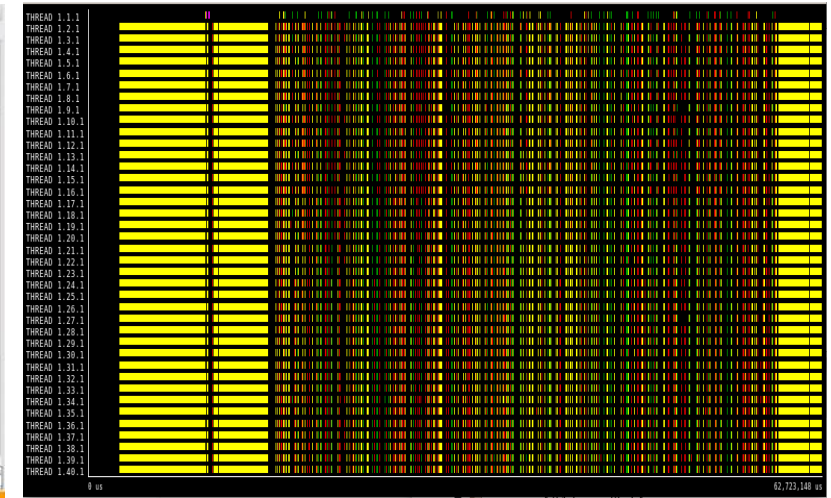
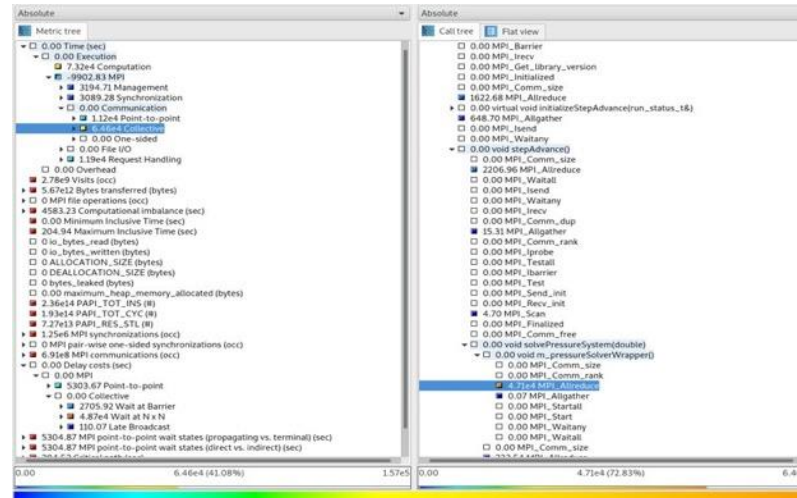
How do we measure the performance of our parallel programs?



speed-up and efficiency plots?



collect data using performance monitoring tools?



*Cube, perf. metrics per routines/call stack,
data collected by Scalasca/Score-P*

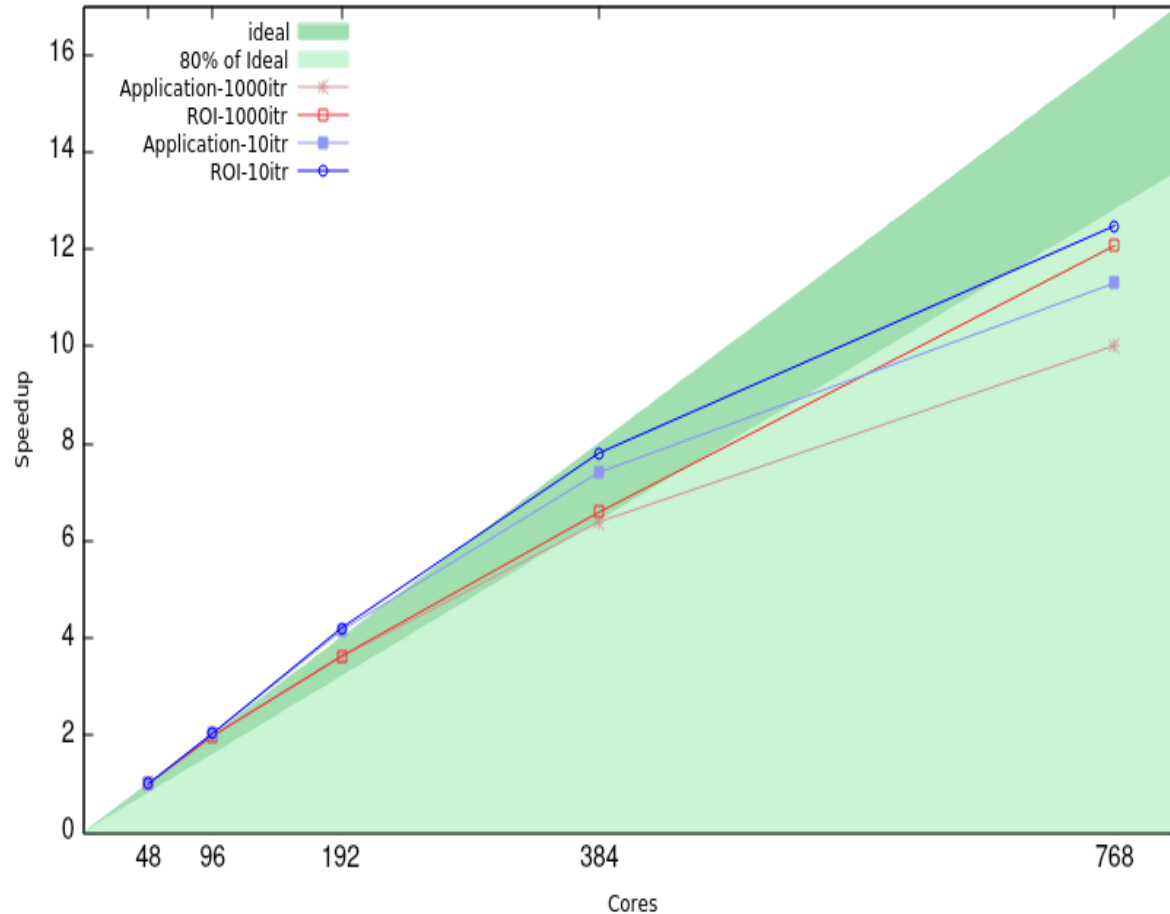
*Paraver, timeline view of program
execution, data collected by Extrae*

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Example 1



A Computational Fluid Dynamics Code

- Code: C++, MPI
- Platform: MareNostrum-IV
 - Dual Intel Xeon Platinum 8160 Skylake 48-core nodes
- Performance data collection:
 - *Score-P/Scalasca* using compiler instrumentation filter and hardware counters
- Scale: 48-768 cores (1-16 nodes)



Example 1

Number of cores	48	96	192	384	768
Global Efficiency	0.93	0.94	0.93	0.84	0.76
Parallel Efficiency	0.93	0.91	0.87	0.77	0.68
Load balance	0.99	0.98	0.98	0.97	0.95
Communication Efficiency	0.94	0.92	0.89	0.79	0.72
Serialisation	0.95	0.94	0.92	0.85	0.81
Transfer efficiency	0.99	0.99	0.97	0.94	0.89
Computational Scaling	1.00	1.03	1.07	1.09	1.12
Instruction Scaling	1.00	0.99	0.97	0.95	0.92
IPC Scaling	1.00	1.05	1.10	1.18	1.27
Frequency Scaling	1.00	1.00	1.00	0.98	0.96

POP Metrics

We immediately see that Serialisation is the main factor that limits the scalability



Example 1

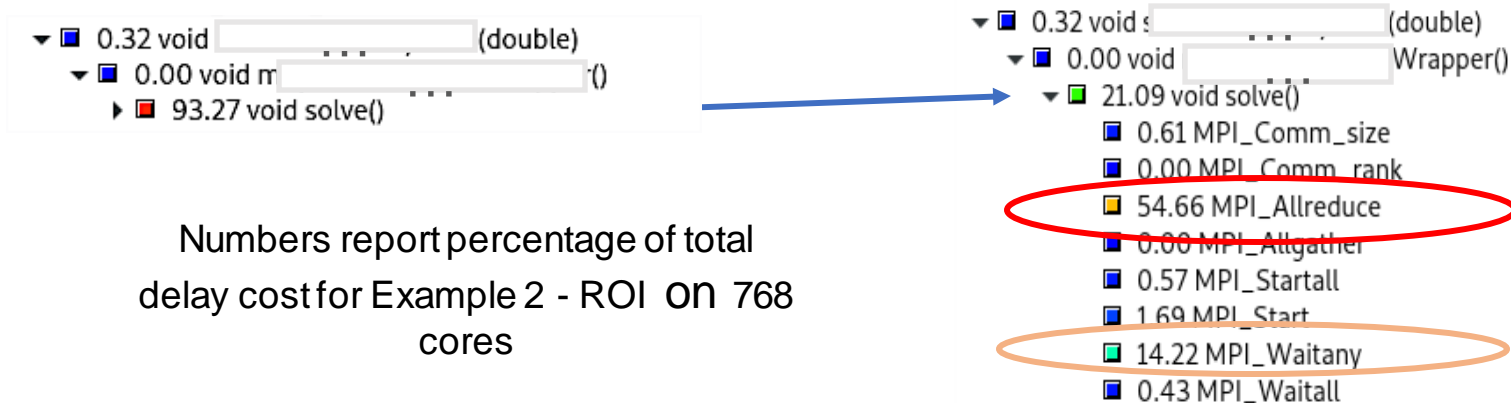
Cause of Low Serialisation Efficiency

Serialisation

- typically happens due to at least one process arriving early/late at synchronization point

Scalasca calculates a delay cost metric

- This metric highlights the root causes of serialization



The MPI collective calls and imbalanced computation regions within a Library call were the main causes of the serialisation on 768 cores

Inclusive values (►)
Exclusive values (▼)



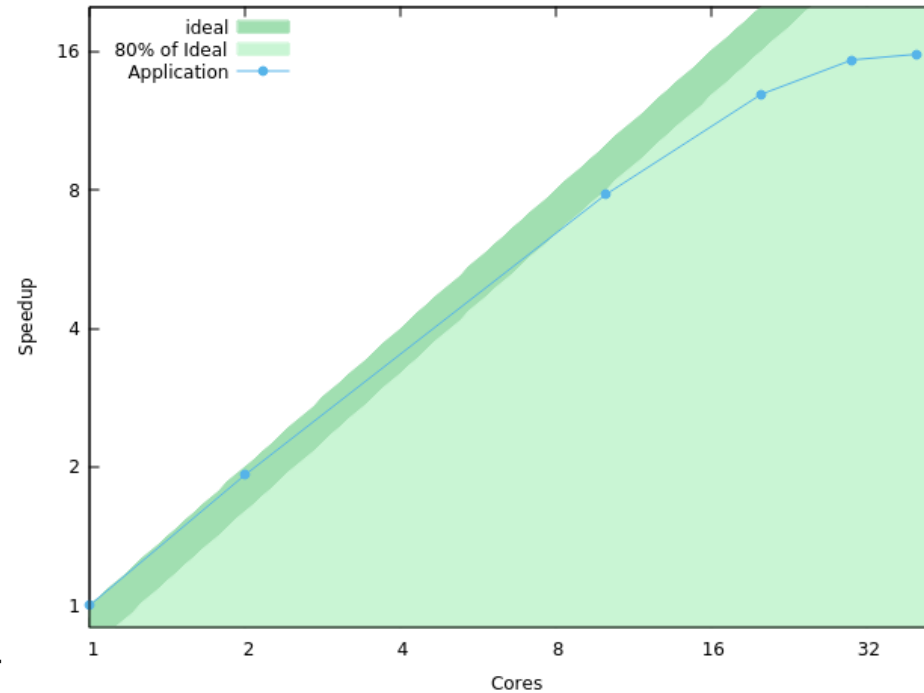
Example 2

Methodology & Tools

- Code: C++, Fortran, MPI
- No access to the source code
- **Platform:**
- Dual Intel Xeon Gold 6248 CPU @ 2.50GHz – 40 cores
- Intel Fortran and C++ compiler with MKL and MPI Library (2019 version)

Performance data collection: Extrae

- Scale: 2- 40 cores



Timeline of the program execution on 40 cores



Example 2

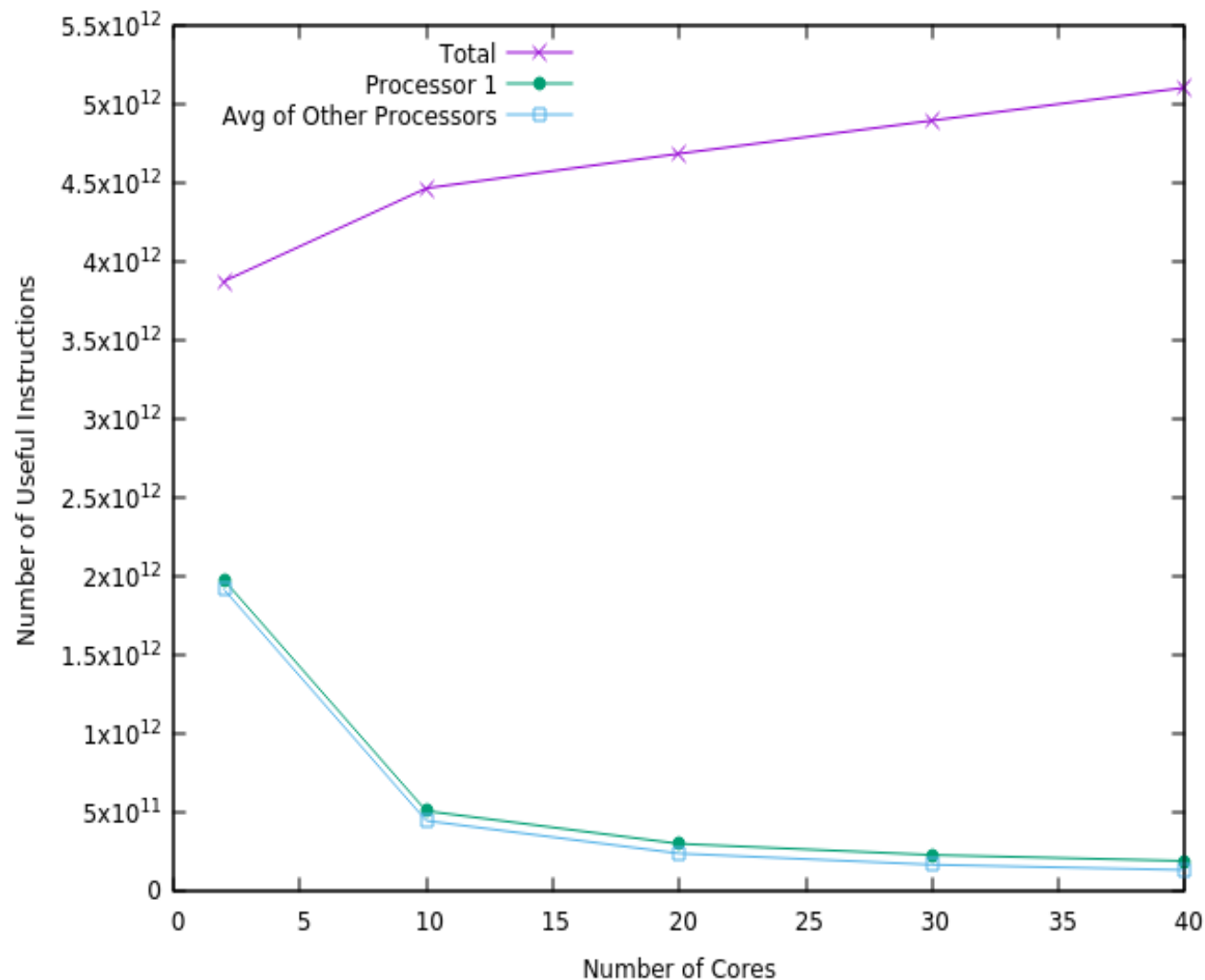
POP Metrics

Number of cores	2	10	20	30	40
Global Efficiency	0.95	0.73	0.60	0.47	0.36
Parallel Efficiency	0.95	0.89	0.81	0.75	0.68
Load balance	0.95	0.92	0.85	0.81	0.80
Communication Efficiency	0.99	0.97	0.95	0.92	0.85
Serialisation	1.00	0.99	0.99	0.98	0.94
Transfer efficiency	0.99	0.98	0.96	0.94	0.91
Computational Scaling	1.00	0.82	0.74	0.63	0.53
Instruction Scaling	1.00	0.87	0.83	0.79	0.76
IPC Scaling	1.00	0.99	0.95	0.90	0.83
Frequency Scaling	1.00	0.95	0.94	0.88	0.84

Load imbalance & increasing instruction count are major factors that limit the scalability



Example 2



Useful Instructions

Total number of useful instructions increases with increasing number of processes

- Low Instruction scaling

Process 1 always executes more instructions compared with other processes

- Load imbalance

With 40 processes, Processor 1 executes 46% more instructions with respect to average number of instruction per process

- Amdahl's law



Example 3

A Computational Fluid Dynamics Code

POP metrics from the <i>Performance Assessment</i>				
# threads	1	10	30	45
Global Efficiency	1.00	0.80	0.36	0.26
↳ Parallel Efficiency	1.00	0.86	0.60	0.55
↳ OpenMP Region Efficiency	1.00	0.95	0.74	0.70
↳ Serial Region Efficiency	1.00	0.91	0.86	0.85
↳ Computational Scaling	1.00	0.94	0.60	0.48
↳ Instruction Scaling	1.00	1.01	1.00	1.00
↳ IPC Scaling	1.00	0.92	0.61	0.50
↳ Frequency Scaling	1.00	1.00	0.98	0.95

Code: Fortran, OpenMP

Platform: MareNostrum-IV

Dual Intel Xeon Platinum 8160 Skylake 48-core nodes

Scale: 1-45 threads

Tools: Extrae & Paraver, Vtune, MAQAO

Poor scalability of the code is due to multiple factors:

- OpenMP Region Efficiency and reducing IPC are major limiting factors,
- Resulting in, respectively, poor Parallel Efficiency and poor Computational scaling



Example 3

Improving the Performance

	Original code for <i>Proof of Concept</i>						Modified code					
# threads	1	2	10	18	30	45	1	2	10	18	30	45
Global Efficiency	1.00	0.86	0.65	0.41	0.31	0.15	1.00	0.86	0.72	0.62	0.51	0.37
↳ Parallel Efficiency	1.00	0.97	0.80	0.69	0.62	0.59	1.00	0.97	0.90	0.83	0.78	0.75
↳ OpenMP Region Efficiency	1.00	0.97	0.81	0.69	0.63	0.60	1.00	0.97	0.91	0.85	0.80	0.78
↳ Serial Region Efficiency	1.00	1.00	0.99	0.99	0.99	0.99	1.00	1.00	0.99	0.98	0.98	0.98
↳ Computational Scaling	1.00	0.89	0.81	0.60	0.49	0.26	1.00	0.88	0.81	0.75	0.65	0.49
↳ Instruction Scaling	1.00	1.00	1.00	1.00	0.99	0.97	1.00	1.00	1.00	0.99	0.99	0.98
↳ IPC Scaling	1.00	0.87	0.80	0.60	0.51	0.36	1.00	0.89	0.82	0.77	0.67	0.56
↳ Frequency Scaling	1.00	1.02	1.02	1.00	0.97	0.74	1.00	1.00	0.98	0.98	0.98	0.89

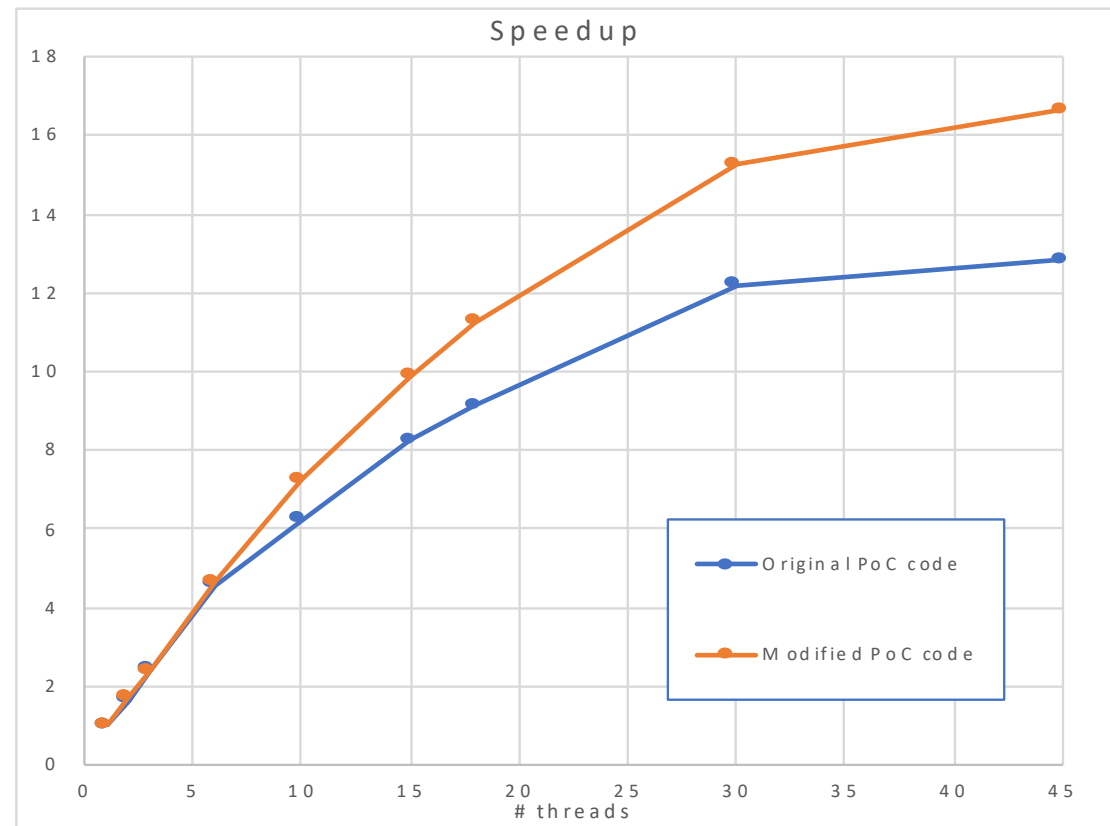
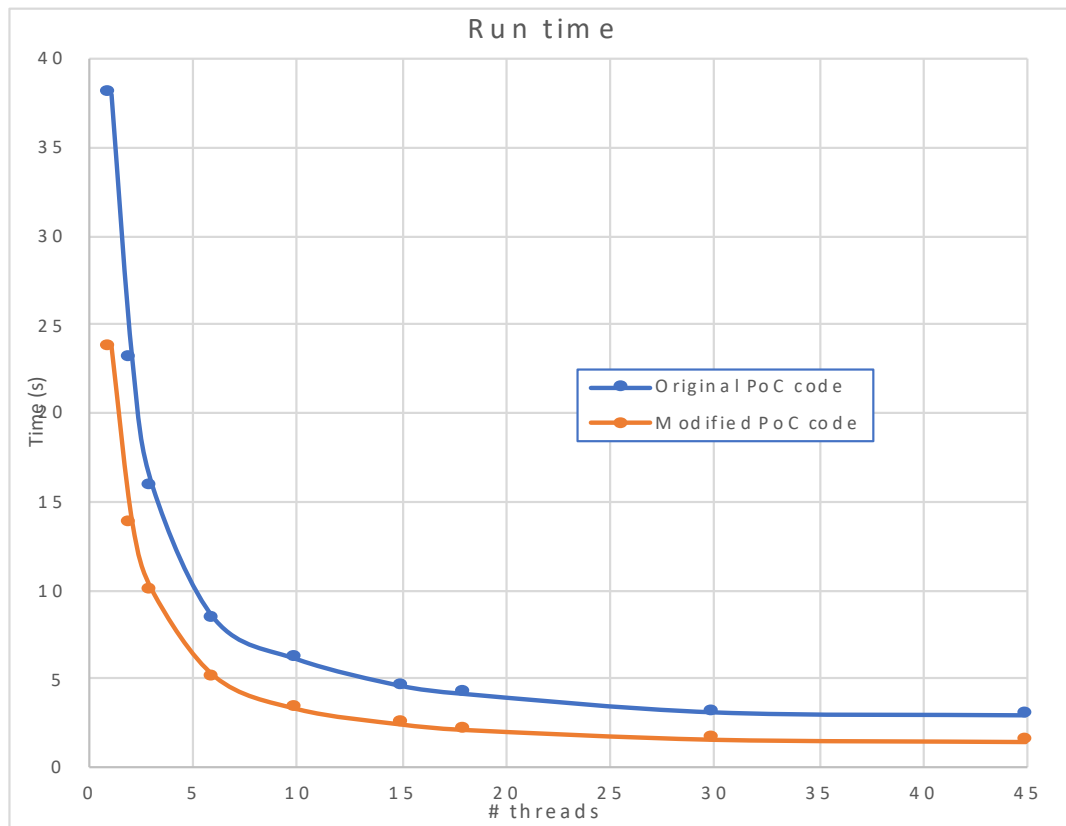
Code refactoring by the POP Proof of Concept service

- Use of OpenMP COLLAPSE clause to improve load balance
- Remove unnecessary calculations
- Use optimal loop ordering with nested loops



Example 3

Performance of modified code



The modified code

- is 1.6x faster on 1 thread due to reduced instruction count
- is 2.1x faster than original on 45 threads
- shows better parallel scaling with a speedup of 16.7 on 45 threads relative to 1 thread



Success stories



success stories

More than 350 services since 2015 across all domains, e.g. engineering, earth & atmospheric sciences, physics, biology and genetics

<https://pop-coe.eu/blog/tags/success-stories>

Success Stories

13
JUL



POP Collaboration with PerMedCoE achieves a 1.45x Speedup in PhysiCell, one of PerMedCoE Core Applications
A collaboration between POP and PerMedCoE started with the performance assessment of PhysiCell.

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25
MAY



Run time halved for OpenMP code
Having already identified the three causes of low efficiency in

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17
MAY



A one-day POP online training for SURF
Jonathan Boyle and Federico Panichi from POP partner NAG (Numerical Algorithms Group) recently pr

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Diversifying the HPC community: boosting the uptake of advanced HPC training by women and underrepresented groups
HPC training is a crucial step in encouraging and building a diverse and inclusive workforce for

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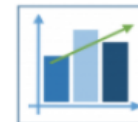
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POP for Astronomy - 40% Reduction in Execution Time for the PIERNIK Code
PIERNIK is a parallel astrophysical fluid simulati

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Performance Improvements by More Than 30% and a Data Race Fixed for CalculiX Code
CalculiX is an open source computational fluid dynamics code.

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588x and 488x Execution Time Speedups of a Volcanic Hazard Assessment Code
The Probabilistic Volcanic Hazard Assessment Work Flow package (PVHA_WF) is a workflow created fo

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Where to find us

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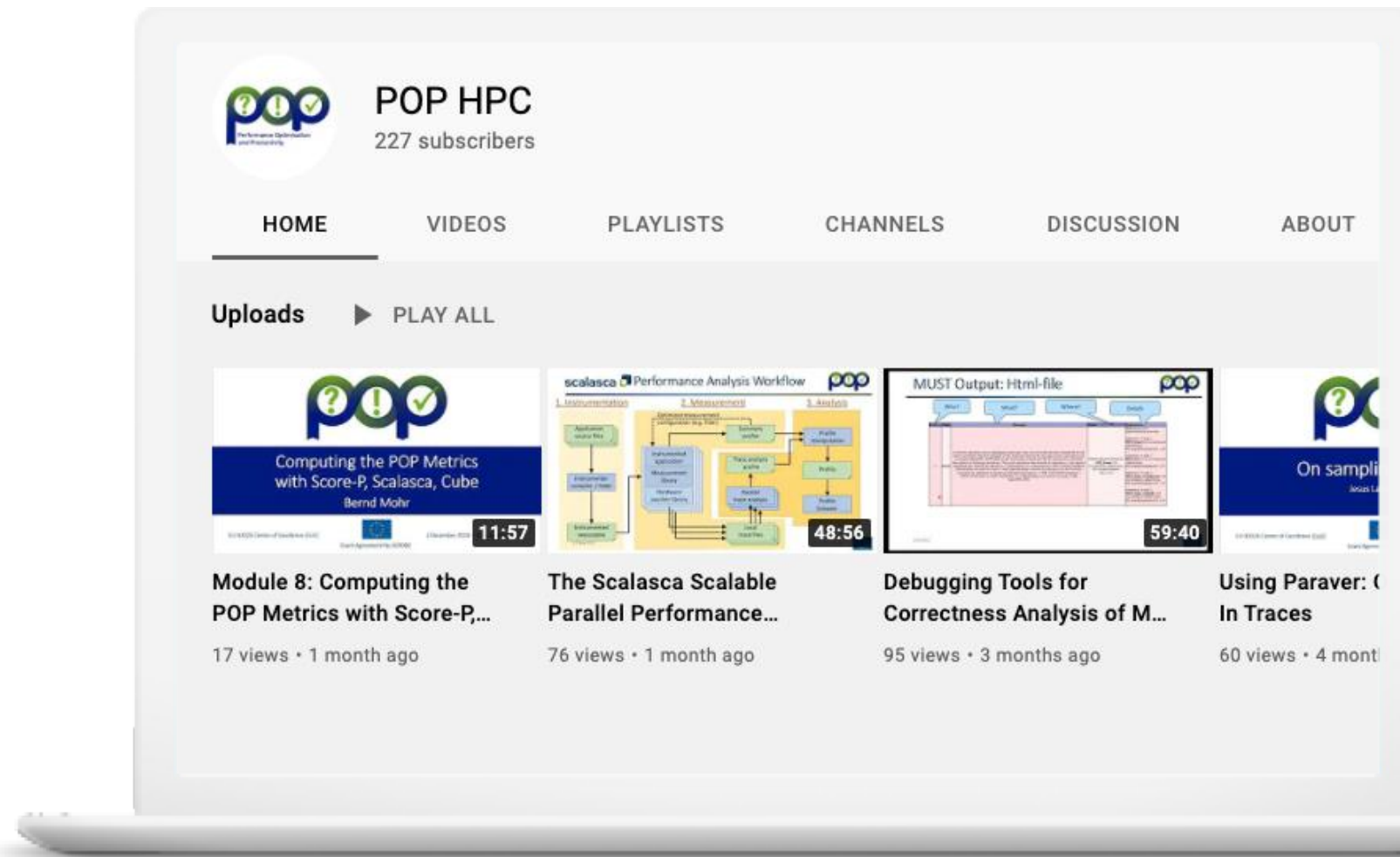
POP Website

<https://pop-coe.eu/services>

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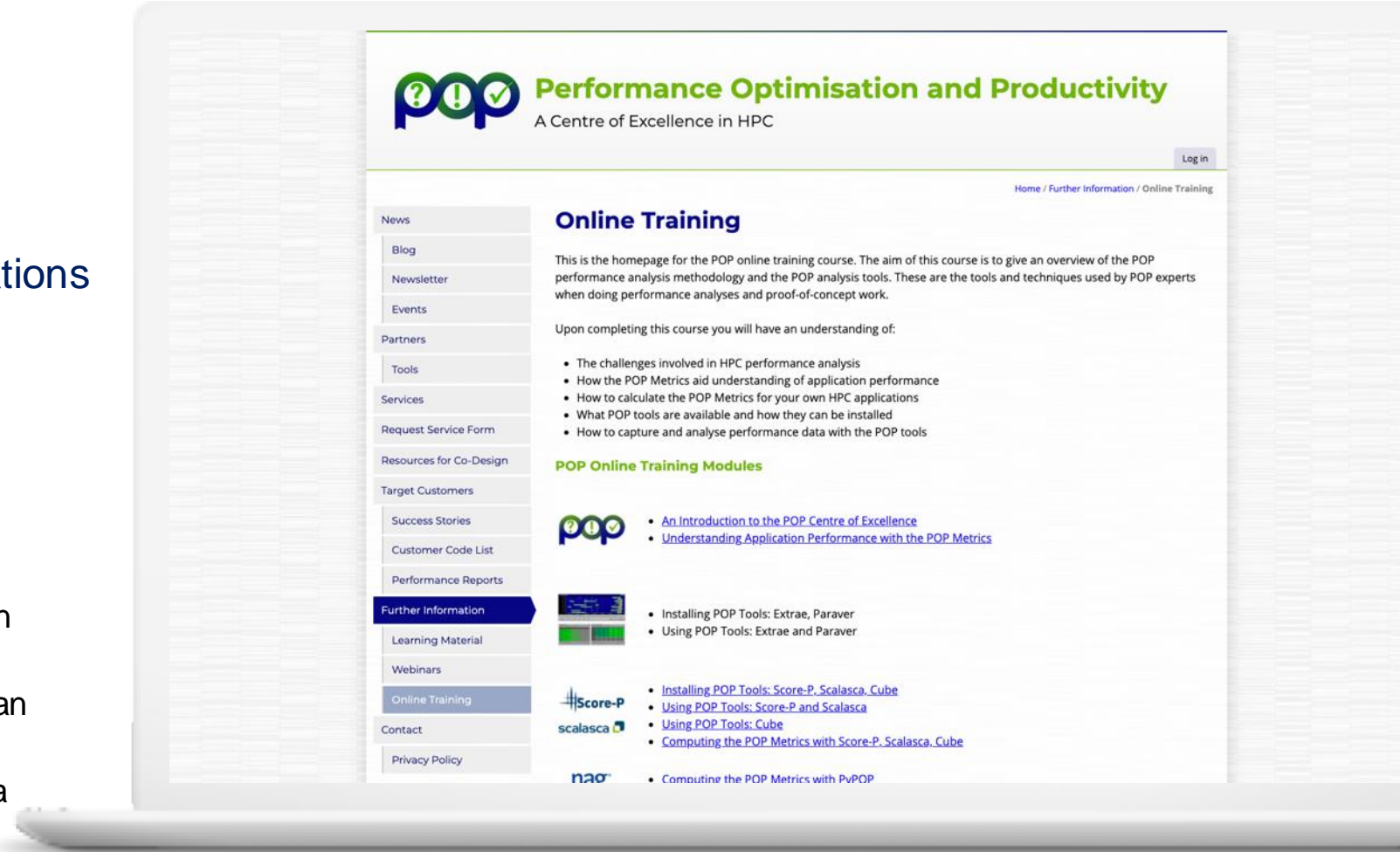


What you can learn

A series of self-study modules
For those with limited experience in
performance analysis of HPC applications

Learning Objectives

- The challenges involved in HPC performance analysis
- How the POP Metrics aid understanding of application performance
- How to calculate the POP Metrics for your own HPC applications
- What POP tools are available and how they can be installed
- How to capture and analyse performance data with the POP tools





What you can learn

POP Performance Metrics

- Build a quantitative picture of application behavior
- Allow quick diagnosis of performance problems in parallel codes
- Identify strategic directions for code refactoring
- So far metrics for MPI, OpenMP and Hybrid (OpenMP + MPI) codes

POP works

- Across application domains, platforms, scales
- With (EU/UK) academic and industrial customers including code developers, code users, HPC service providers and vendors
- To apply for a POP service go to <https://pop-coe.eu/services>

POP CoE

- Promotes best practices in parallel programming
- Encourages a systematic approach to performance optimization
- Facilitates and invests in training HPC experts



Performance Optimisation and Productivity



**HPC
Best Practices
for Research
and Education**

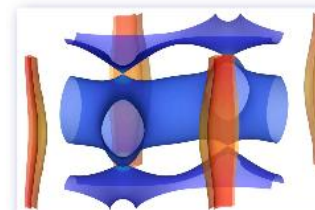
**Collaboration with POP
to achieve academic
excellence**

- Performance optimisation for parallel research software, allowing better usage of universities' resources and creating capacity for solving more complex problems
- Learning materials and training workshops suitable for MSc level, Ph.D students and Postgraduate researchers.



POP achieved 10-fold scalability improvement for EPW (Electron-Phonon Coupling using Wannier interpolation), a materials science code developed by researchers at the University of Oxford. Important optimisations included:

- Load imbalance issues were addressed by choosing a finer grain configuration
- Specialized routines were written for one part of the simulation to avoid unnecessary calculations
- Vector summation operations were optimised
- File I/O was optimised, bringing down seven hours of file writing to under one minute.



EPW, University of Oxford

Your parallel code: better

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Performance Optimisation and Productivity

Contact:



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pop@bsc.es



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