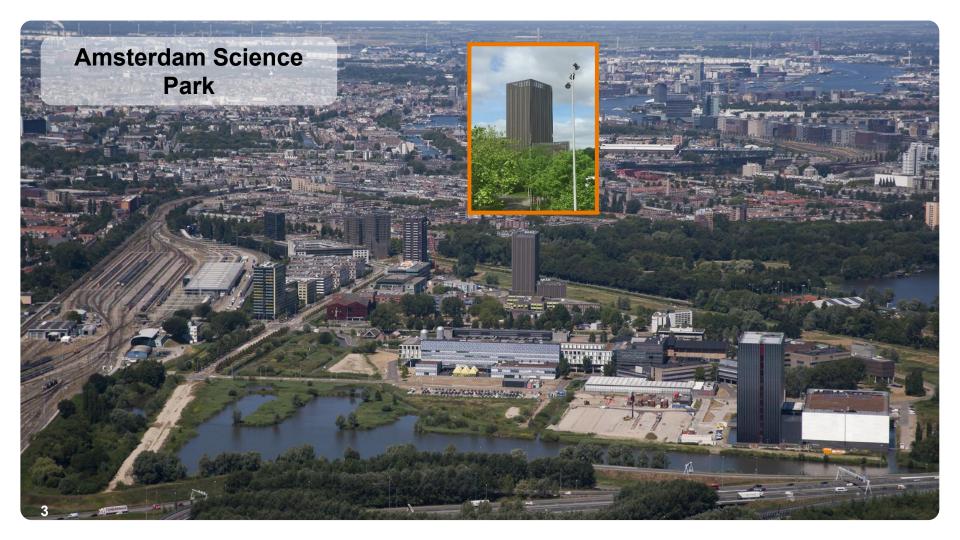
Green Computing at Dutch National Supercomputing Center

EGI : Green Computing webinar Introduction session

Sagar Dolas Program Lead - Future Computing (HPC, Quantum & Networking) SURF Labs

Agenda

- Introduction to SURF, Netherlands
- Energy as a design point for supercomputing operations
- Energy experiments on experimental systems
- EAR design & data collections
- Questions





Reliable, secure and innovative ICT infrastructure Digital innovation and transformation of education and research Knowledge exchange and trainings Services development and integration with EU initiates

Snellius : Dutch National Supercomputer



Snellius

Phase 1:

- 76,832 cores (1.6 ×)
- 144 GPUs (3 Pflop/s, 14 ×)
- Total peak: 6.1 Pflop/s (3.4 ×)

Phase 1 and 2

- Total peak: 11.2 Pflop/s (6.2 ×)
- Full system, based on choice for Phase 3:
 - > 200,000 cores (> 4 ×)
 - Total peak: 13.6 21.5 Pflop/s (7.6 11.9 ×)

Snellius – Energy Consumption

- HPL Energy Consumption (typical use: 85%, idle use: ~ 25%)
 - Snellius
 - Phase 1: 620 kW (0.7 ×)
 - Phase 1 and 2: 1200 kW (1.4 ×)
 - Full system ("worst case": Phase 3 GPU): 1430 kW (1.6 ×)
- Average energy consumption based on phasing
 - 2021–2022: ~ 1 ×
 - 2023: 1.5 ×
 - 2024 and later: 1.6 ×



Datacenter

PUE < 1.19

Waste heat reuse Through hot & cold well

Use of Hvdroelectricitv Infrastructur Level

Rear door heat exchanger + Direct water cooling Hot & cold island

CPU + GPU architecture System Level

Energy & power management, energy capping, Energy accounting Application Level

Application analysis & tuning using EAR User awareness

Platform for energy & performance visualisation

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architecture

Datacenter	Infrastructure Level	System Level	Application	User
PUE < 1.19	Rear door heat	Energy & power	Level	awareness
Waste heat reuse	exchanger + Direct water	management, energy capping,	Application	Platform for energy &
Use of Hydroelectricity	cooling Hot & cold island	Energy accounting	monitoring, analysis &	performance
, ,	CPU + GPU		tuning	

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Waste heat reuse Use of Hydroelectricity			Application monitoring, analysis &	Platform for energy & performance visualisation
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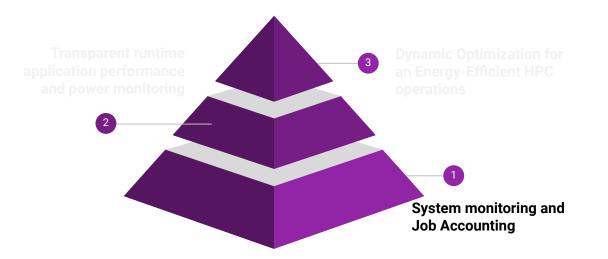
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EAR software



EAS main values





EAR software



EAS main values Transparent runtime 3 application performance and power monitoring System monitoring and **Job Accounting**



EAR software



EAS main values Transparent runtime **Dynamic Optimization for** 3 application performance an Energy-Efficient HPC and power monitoring operations System monitoring and **Job Accounting**



Energy-efficiency process



System Monitoring and Contro

Nodes temperature, power (Node, DRAM. Package, GPU), effective frequency ...

Automatic reporting of run time hardware issues Node powercap: heterogeneous nodes, power balance Cluster powercap : Hierarchical architecture for system scalability

Job Accounting

Performance metrics monitoring: CPI, CPU Gflops, Memory bandwidth, MPI hints, IO bandwidth, GPU utilization

Granularity: jobid, stepid, "loop", user, node

Analysis

Hints for application analysis and optimization Hints for system analysis and optimization Reporting system, extensible based on plugins Trace file generation Energy accounting and reporting commands included Graphana visualization

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EAS

Runtime application energy optimization Transparent, dynamic and lightweight runtime library with no user intervention required Automatic energy savings according to energy policies





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Energy optimization



- Common to all the policies
 - At runtime, loop is detected and loop signature is computed
 - EAR uses time and power models for frequency selection
- EAR-min_time policy
 - $\circ\,$ Applications start at default policy $\,$ frequency lower than nominal
 - Loops with "enough" energy efficiency are accelerated (compute bound)
 - $\circ\,$ Policy detects changes and applies again the policy
 - EAR- min_energy policy
 - Applications start at default policy frequency (nominal)
 - Frequency is reduced up to a maximum performance penalty



BSC



Energy optimization



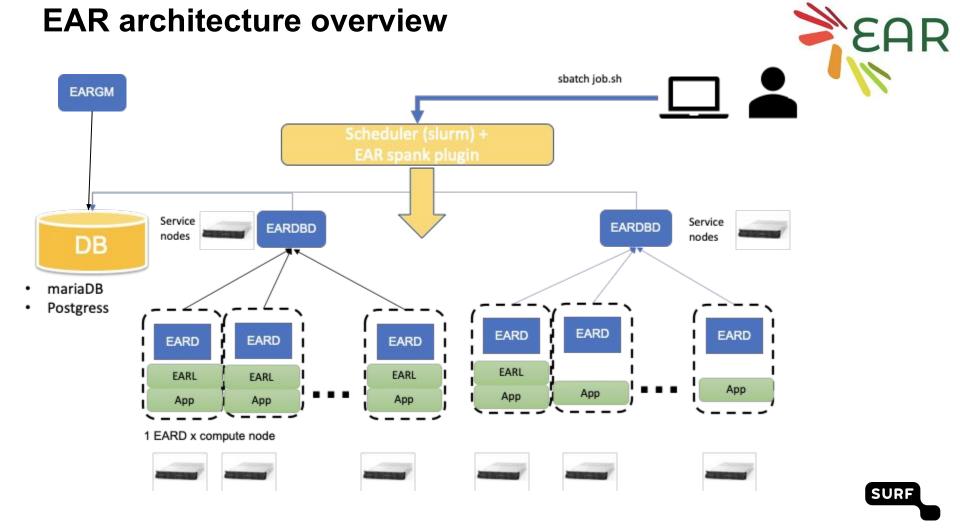
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BSC

reserved





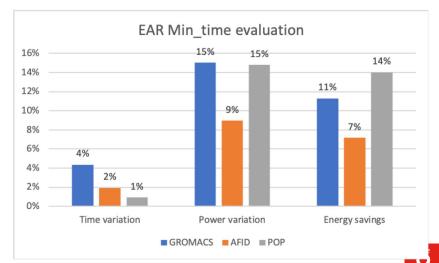
Experimental System for energy experiments

- Experiments system have 4x Lenovo SR650 system
- Skylake 6148 @2.5GHz 40c nodes with 100 g HDR
- 2x V100 nodes for GPU tuning experiments
- Default frequency=2.5GHz
- Easybuild software stack
- Jenkins for automated software builds
- User access for community
- EAR as energy management framework



Energy experiments with real use cases

- •Application use cases provided by SURF Open Innovation Lab (SOIL) collaboration.
 - Experiments executed in Lenovo SD530 system
 - Skylake 6148 @2.4GHz 20c nodes with EDR network
 - Default frequency=2.0GHz
 - GROMACS 640 processes. 16 nodes
 - AFID 600 processes. 15 nodes
 - POP 400 processes. 10 nodes





Average energy savings of 10%

SURF

voua

Visualisation Prototype



Other approaches for achieving energy efficiency

- ML based acceleration for HPC application
- Optimizing the number of simulation required to achieve particular outcome on HPC system
- Exploring Neuromorphic computing for scientific analysis & experimentation
- Introduce the concept of energy accounting



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- 2. Large number of experiments will be carried out to fully understand tunable parameters
- 3. Assessing impact, operational and user requirements
- 4. Pilots will be carried out at SURF infrastructure jointly with member institutions.
- Most important : knowledge dissemination : Joint hackathons, workshops and more.



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