

# Monitoring Energy Consumption of Workloads on HPC Vega

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### Agenda

- Introduction
- Related Work and Methods
- Methodology
- Experimental Results
- Conclusion and Future Work
- References



### Introduction



### Introduction

- Insight of first launched EuroHPC JU system (setup, adjustments and mechanisms).
- Management and monitoring of energy consumption within HPC AI workloads.
- Interest from various scientific domains, industry, and SMEs.
- Necessary to introduce effective and essential mechanisms for energy sustainability.
- Differences in systems and architectures lead to different end solutions and policies<sup>1</sup>.
- A challenge for users, lack of knowledge, no unified framework.
- EuroHPC Federation Platform project for integrating current and future systems<sup>2</sup>.
- System administrators & Application Support Teams (AST) provide high-level support across multidisciplinary areas and **best practices and guidelines are available**.



<sup>&</sup>lt;sup>1</sup>Roman lakymchuk et al. Best Practice Guide – Harvesting energy consumption on European HPC systems: Sharing Experience from the CEEC project. 2024. DOI: 10.5281/zenodo.13306639.

<sup>1: 10.5281/</sup>zenodo.13306639.
2 Henrik Nortamo. "EuroHPC FP: A Federated Platform for HPC Infrastructure in Europe". Presented at FOSDEM 2025. 2025. URL: https://fosdem.org/.

### Related Work and Methods



### Related Work and Methods

Research on energy efficient HPC involves mechanisms with focus on pillars framework<sup>3</sup>.

- Infrastructure.
- hardware.
- software,
- optimization.

Despite advances there are still challenges to consider:

- Scalability issues,
- infrastructure limitations and bottlenecks,
- and efficiency trade-offs.



<sup>&</sup>lt;sup>3</sup>Torsten Wilde, Axel Auweter, and Hayk Shoukourian. "The 4 Pillar Framework for energy efficient HPC data centers". In: Computer Science - Research and Development 29 (2013).

### Related Work and Methods

- Power Capping and Constraint
- Baseboard Management Controller (BMC)
- Intelligent Platform Management Interface (IPMI)
- Running Average Power Limit (RAPL)
- System Management Interface (SMI)
- NVIDIA Data Center GPU Manager (DCGM)
- Frequency Throttling
- Checkpoints and Restart (CR)
- Software in HPC
- Slurm Workload Manager
- Monitoring and Metrics Visualization
- Current Setup of HPC Vega



### Power Capping and Constraint

Controlled and limited operation of the system environment (HW/SW)<sup>4</sup>.

- Monitoring and extending consumption according to requirements,
- effective or preventive constraints (hardware failures),
- reducing high operating costs in higher peaks,
- lower priority tasks,
- partially turning off the compute nodes,
- distributed loads,
- in the last resort, jobs may still be killed.



<sup>&</sup>lt;sup>4</sup>Alberto Cabrera et al. "Energy efficient power cap configurations through Pareto front analysis and machine learning categorization". In: Cluster Computing 27 (2023), pp. 3433–3449, ISSN: 1573-7543.

### Baseboard Management Controller (BMC)

**BMC** is a microcontroller integrated into the motherboard of servers, reachable via **Out-of-Band (OOB)** network and enables:

- Remote power management,
- monitoring,
- troubleshooting,
- system administration of server systems,
- independent of the OS.



### IPMI and RAPL

**Intelligent Platform Management Interface (IPMI)** is a set of standard specifications and components.

- Commonly used utility is ipmitool,
- intended for system administration via OOB network.
- Controlling computer subsystems BIOS, CPU, and OS,
- monitoring, logging, recovery, and retrieving data from sensors (SDR, energy, etc.).

Running Average Power Limit (RAPL) is an interface feature introduced by Intel's Sandy Bridge architecture,

- Provides access to real-time measurements of hardware counters based on CPU sockets and DRAM,
- suitable for measurements of power consumption of various HW setups.<sup>5</sup>



<sup>&</sup>lt;sup>5</sup>Howard David and et al. "RAPL: Memory power estimation and capping". In: 2010 ACM International Symposium on Low-Power Electronics and Design. 2010, pp. 189–194. DOI: 10.1145/1840845.1840883.

### **NVIDIA**

- **SMI** is a command (nvidia-smi) utility used for managing and monitoring NVIDIA devices, based on NVIDIA Management Library (NVML) C-based API.
- DCGM is managing and monitoring for GPUs and allows reducing the power consumption of GPUs clock speed in DCs.
- NVIDIA Multi-Instance GPU (MIG) provides up to seven independent slices per GPU (smaller workloads<sup>6</sup>.
- Generic Resources (GRES) such as NVML are available within an additional configuration file (gres.conf)<sup>7</sup>.



<sup>&</sup>lt;sup>6</sup>Tirth Vamja et al. On the Partitioning of GPU Power among Multi-Instances. 2025. DOI: 10.48550/arXiv.2501.17752. arXiv: 2501.17752 [cs.DO]." \* Comparison of Comparison

<sup>&</sup>lt;sup>7</sup>Morris A. Jette and et al. "SLURM: Simple Linux Utility for Resource Management". In: Proceedings of the 9th International Workshop on Job Scheduling Strategies for Parallel Processing (JSSPP), 2002. DOI: 10.1007/10968987.3.

### Frequency Throttling

Adjustment of the workload **frequency** and **performance**, which reduces the number of instructions inside the processor, to save consumed energy while not affecting or compromising the performance.

- Is possible within a heterogeneous system (CPU/GPU),
- adjust the core or memory frequency,
- solutions have been developed to automatically adjust frequencies per workload,
- dynamic adjustment without affecting the performance (CPU and GPU),
- newer processors enable performance autonomous scalability on OS level<sup>8</sup>.



<sup>&</sup>lt;sup>8</sup>Kai Ma et al. "Energy conservation for GPU-CPU architectures with dynamic workload division and frequency scaling". In: Sustainable Computing: Informatics and Systems (2016), pp. 21–33, ISSN: 2210-5379.

### Checkpoints and Restart (CR)

CR mechanisms save states of jobs as checkpoints that are restored upon restart9.

- Adopted for fault tolerance in HPC workloads,
- several checkpoints could be established per workload,
- continuing the execution after interruption,
- used for power management to reduce cost,
- offloading a larger workload,
- interconnect errors caused by network congestion,
- various CR implementations i.e DMTCP, BLCR, and CRIU.



<sup>&</sup>lt;sup>9</sup>Ifeanyi P. Egwutuoha and et al. "A survey of fault tolerance mechanisms and Checkpoint/Restart implementations for High performance computing Systems". In: The Journal of Supercomputing 65 (2013). ISSN: 1573-0484. DOI: 10.1007/s11227-013-0884-0.

### Software in HPC

The various application domains resulting in different workloads, preperation and requirements.

- different package managers
- custom-builds from source,
- **containerization** is increasingly popular due to various beneficial advantages such as reproducibility, portability, consistent environment, scalability, and isolation <sup>10</sup>.
- various repositories could be accessible via CernVM-FS<sup>11</sup>,
- tools with power management capabilities i.e. APM, AMD ROCm, **LIKWID**, **Perf**, **PAPI**, EAR, LLview, and **MERIC**<sup>12</sup>.
- importance of efficient code, code optimization, and performance analysis!

<sup>&</sup>lt;sup>10</sup>Barbara Krašovec and Teo Prica. "Secure Usage of Containers in the HPC Environment". In: Nordic e-Infrastructure Tomorrow. Springer Nature, 2025 pp. 96–112. ISBN: 978-3-031-86240-3.

<sup>&</sup>lt;sup>11</sup> Jakob Blomer and et al. "New directions in the CernVM file system". In: Journal of Physics: Conference Series 898 (2017). DOI: 10.1088/1742-6596/898/6/062031.

<sup>&</sup>lt;sup>12</sup>Roman lakymchuk et al. Best Practice Guide – Harvesting energy consumption on European HPC systems: Sharing Experience from the CEEC project. 2024. DOI: 10.5281/zenodo.13306639.

### Slurm Workload Manager

**Slurm**<sup>13</sup> manages resources, job scheduling, measuring, and retrieving data from its database.

- Wide range of plugins, tailored solutions,
- power management is integrated through a plugin,
- either to prevent overload (prevent HW issues) or increase the utilization or consumption,
- commands suspend, hold a running job, or resume to resume a suspended job,
- accounting is done through a Slurm Energy Accounting Plugin,
- GPU management library, IPMI, PM\_Counters, RAPL, and XCC.



<sup>13</sup> Morris A. Jette and et al. "SLURM: Simple Linux Utility for Resource Management". In: Proceedings of the 9th International Workshop on Job Scheduling Strategies for Parallel Processing (JSSPP), 2002, DOI: 10.1007/10968987 3.

### Monitoring and Metrics Visualization

In the facilities such as HPC, monitoring is done via metrics which are collected, stored, and visualized:

- Supply infrastructure,
- system-wide overview of the cluster,
- specific equipment, devices, hardware, and software being monitored.
- electricity consumption, which fluctuates between 700 kW and 1 MW.



### Grafana, Prometheus, and Victoria Metrics

- Grafana is an open-source solution for the visualization of metrics obtained from various databases stored into Time Series Database, queries, visualized and monitored using prepared dashboards.
- **Prometheus** is open-source used to collect various metrics, monitoring and alerting and can be added as a source within Grafana.
- Victoria Metrics is an open-source high performance time series database with capabilities to monitor and control the system, added as an additional source to Grafana.
- Node exporter is a tool used to monitor, collect and export various metrics from a client.









### Current Setup of HPC Vega

- Nodes are divided into five partitions cpu, gpu, dev, largemem, and longcpu (1028).
- Energy consumption is retrieved through **IPMI** sensors,
- entries are assigned to parameters within **Slurm** configuration, and
- deployment, configuration distribution cluster-wide, is managed through Ansible.
- Node Health Check (NHC) on nodes.
- Two storages, the **Lustre** parallel file system (1 PB) for scratch and a large capacity storage based on **Ceph** (19 PB) for home, project directories, data storage, and beyond.



### Current Setup of HPC Vega

- Power constraint is managed via various physical sensors in a DC, readings are retrieved from heating, ventilation, air conditioning (HVAC).
- Supervisory Control and Data Acquisition (SCADA) and Data Center Infrastructure Management (DCIM) based system.
- An alert is triggered after deviations of events from given thresholds.
- Triggers upon fail (picked by the management system),
- A signal is sent to Slurm through **snmptrap**, to reduce power consumption via Slurm abilities with **suspend** command and an unsuspend via **resume** command.
- Tailored implementation executed synchronously rack-by-rack, maintained by the vendor.
- Reduction or increase of consumed power and sustaining the lifetime of the components.
- BEO is currently disabled in HPC Vega!<sup>14</sup>



<sup>&</sup>lt;sup>14</sup>Roman lakymchuk et al. Best Practice Guide – Harvesting energy consumption on European HPC systems: Sharing Experience from the CEEC project. 2024. DOI: 10.5281/zenodo.13306639.

### Current Setup of HPC Vega

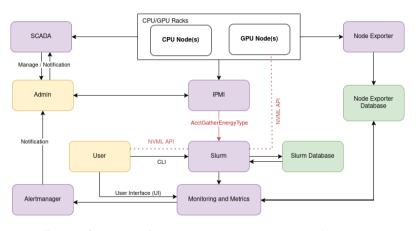


Figure: Overview of measurement, monitoring, and metrics.



### Grafana, Prometheus, and Victoria Metrics



Figure: Grafana dashboard of total power consumption by equipment.



### Differential-Evolution-Based Hyperparameter Otimization

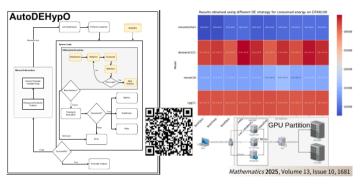


Figure: High-Performance Deployment Operational Data Analytics of Pre-Trained Multi-Label Classification Architectures with Differential-Evolution-Based Hyperparameter Optimization (AutoDEHypO)<sup>15</sup>



<sup>&</sup>lt;sup>15</sup>Teo Prica and Aleš Zamuda. "High-Performance Deployment Operational Data Analytics of Pre-Trained Multi-Label Classification Architectures with Differential-Evolution-Based Hyperparameter Optimization (AutoDEHypO)". In: *Mathematics* 13.10 (2025). ISSN: 2227-7390. DOI: 10.3390/math13101681. URL: https://www.mdpi.com/2227-7390/13/10/1681.

## Methodology



### Slurm Account Gather Configuration

Challenges had to be tackled when enabling the power measurement of consumption via Slurm. Frequencies were adjusted using **AcctGatherNodeFreq** parameter, which was set from **25** to **60** and **JobAcctGatherFrequency** parameter, from **10** to **60**.

```
AcctGatherEnergyType=acct_gather_energy/ipmi
AcctGatherNodeFreq=60
JobAcctGatherFrequency=60
JobAcctGatherType=jobacct_gather/cgroup
```

### CPU and GPU are mapped to the IPMI sensor at different positions:

```
EnergyIPMIFrequency=60
EnergyIPMICalcAdjustment=yes
EnergyIPMITimeout=60
EnergyIPMIPowerSensors=Node=value  # CPU = 13 ; GPU = 20
```



### Generic Resources Configuration

The modification was also made to the **GRES** configuration file. The line commented below represents the state before switching to **NVML**:

```
AutoDetect=nvml
Name=gpu Type=ampere File=/dev/nvidia[0-3]
```

```
#NodeName=gn[01-60] Name=gpu Type=ampere File=/dev/nvidia[0-3]
```



### Consumed Resources

To retrieve consumed resources of finished jobs from Slurm via **sacct** command:

Similarly, to retrieve consumed resources of finished Slurm jobs with **sacct** command and parameter **TRESUsageInAve** with enabled **NVIDIA NVML**:

```
sacct -j $%jobid$ -u $%USER$ -Pno TRESUsageInAve
```

Monitoring of the GPU utilization, memory, and power consumption via nvidia-smi command:

```
nvidia-smi -l 1 --query-gpu=index,utilization.gpu, \
   utilization.memory,power.draw --format=csv,nounits > output &
```



### **Experimental Results**



### Obtained and Measured Results

Retrieved data from ipmi-sensors command was executed on GPU compute node.

	tricved data	nom ipmi bombe	T D COIIIII	iuiiu	was excedited on or o
ID	Name	Туре	Reading	Units	Event
10	Watchdog	Watchdog 2	N/A	N/A	l 'OK'
11	SEL	Event Logging Disabled	N/A	N/A	1 'OK'
12	CWG_LM5066I_IIN	Current	24.11	l A	'OK'
13	LM5066_IIN	Current	4.38	l A	, ok,
14	RDST_INST_PWR1	Power Unit	290.00	l W	, OK,
15	RDST_INST_PWR2	Power Unit	340.00	l W	'OK'
16	RDST_INST_PWR3	Power Unit	260.00	l W	'OK'
17	RDST_INST_PWR4	Power Unit	260.00	l W	'OK'
18	LM5066_PIN	Power Unit	240.00	l W	'OK'
19	AVG_PWR	Power Unit	240.00	l W	I 'OK'
20	CWG_LM5066I_PIN	Power Unit	1335.00	l W	l 'OK'
21	CWG_LM5066I_VIN	Voltage	54.98	l V	l 'OK'
22	LM5066_VIN	Voltage	54.90	l V	1 'OK'
23	CPU0_Status	Processor	l N/A	N/A	'Processor Presence detected'
24	CPU1_Status	Processor	l N/A	N/A	'Processor Presence detected'
25	CPUO_TEMP	Temperature	52.00	I C	1 'OK'
26	CPU1_TEMP	Temperature	57.00	I C	1 'OK'
27	DIMMGO_TEMP	Temperature	49.00	I C	1 'OK'
28	DIMMG1_TEMP	Temperature	49.00	l C	1 'OK'
29	DIMMG2_TEMP	Temperature	48.00	l C	l 'OK'
30	DIMMG3_TEMP	Temperature	47.00	I C	1 'OK'
31	CPUO_DTS	Temperature	48.00	I C	1 'OK'
32	CPU1_DTS	Temperature	43.00	I C	1 'OK'
33	GPU1_TEMP	Temperature	68.00	I C	1 'OK'
34	GPU2_TEMP	Temperature	68.00	I C	1 'OK'
35	GPU3_TEMP	Temperature	68.00	I C	1 'OK'
36	GPU4_TEMP	Temperature	62.00	I C	1 'OK'
37	CWG_LM5066I_TEMP	Temperature	51.00	I C	1 'OK'
38	LM5066_TEMP	Temperature	51.00	i c	l 'OK'
39	MB_TEMP1	Temperature	50.00	İC	l 'OK'
40	MB_TEMP2	Temperature	53.00	İC	, ok,
/11 i	CUC TEMP1	LTomporaturo	I 54 00	ic	יאחי:



### Power Managment Signal

### Power management signal is received within Slurm per rack (GPU/CPU):

```
NTR: _recived_power_management_signal_at_Thu_lum_5_09:26:14_CEST cn[0136-0199] 091-0926/2084-0212; 091-0928, 0928-0923, 0923-0923, 0923-0925, 0925-0925, 0925-0925, 0925-0925, 0925-0925, 0926-0928, 0926-0928]
NRC: Check Secu-hotpatch returned 1
NRC: _recived_power_management_signal_at_Thu_lum_5_09:26:19_CEST cn[0387,0389,0409,0419]
NRC: Check Secu-hotpatch returned 1
NRC: _recived_power_management_signal_at_Thu_lum_5_09:26:19_CEST cn[0587,0589,0409,0419]
NRC: _recived_power_management_signal_at_Thu_lum_5_09:26:19_CEST cn[0587,0589,0409,0419]
NRC: _recived_power_management_signal_at_Thu_lum_5_09:26:19_CEST cn0038
NRC: _recived_power_management_signal_at_Thu_lum_5_09:26:19_CEST cn0038
NRC: _recived_power_management_signal_at_Thu_lum_5_09:31:21_CEST cn0038
NRC: _recived_power_management_signal_at_Thu_lum_5_09:31:21_CEST cn0018
NRC: _recived_power_management_signal_at_Thu_lum_5_09:31:21
```

Figure: Received power management signal for power constraint CPU racks.

```
== Availability ==
Total drained 528 (518 cn nodes ; 10 gn nodes)
Total available 492 (442 cn nodes ; 50 gn nodes)
Overall 48.23% ; CN 46.04% ; GN 83.33%
```

Figure: Availability of drained and availiable compute nodes (319 power constrained).

```
== Availability ==
Total drained 209 (198 cn nodes ; 11 gn nodes)
Total available 811 (762 cn nodes ; 49 gn nodes)
Overall 79.50% ; CN 79.37% ; GN 81.66%
```



Figure: Availability of drained and availiable compute nodes (0 power constrained).

To retrieve consumed resources of finished jobs from Slurm with disabled NVML:

```
cpu=10:00,energy=47340,fs/disk=158664, \
   mem=1136584K,pages=0,vmem=1179592K
```

To retrieve consumed resources of finished jobs from Slurm with enabled NVML:

```
cpu=10:00,energy=74032,fs/disk=158791,gres/gpumem=19882M,\
      gres/gpuutil=100,mem=1136576K,pages=0,vmem=1146816K
```





### Obtained and Measured Results

- TensorFlow, PyTorch, and NVIDIA NeMo frameworks were used.
- Official and optimized **Singularity** containers<sup>16</sup>.
- Containers built from the public NVIDIA NGC catalog.
- Evaluation benchmarks GLUE, SuperGLUE and wikitext-2.
- Executed on single node with up to four GPUs per each workload.
- Limitation of available amount of memory per GPU (20 epochs, batch size 32).
- Models and datasets are available within the Transformers and Datasets package.
- Results obtained from Slurm and IPMI.



<sup>&</sup>lt;sup>16</sup>Barbara Krašovec and Teo Prica. "Secure Usage of Containers in the HPC Environment". In: Nordic e-Infrastructure Tomorrow. Springer Nature, 2025, pp. 96–112. ISBN: 978-3-031-86240-3.

### Obtained and Measured Results

Table: Obtained and measured results from HPC AI workloads.

Framework	Model	Dataset	Nodes /	Counter	Elapsed	Consumed
			GPUs		Time (M)	Energy
						(1)
TensorFlow	distilbert-base-uncased	glue (mrpc)	1/1	IPMI	00:36:04	360.48K
TensorFlow	distilbert-base-uncased	glue (mrpc)	1/4	IPMI	00:17:28	284.36K
PyTorch	gpt2-medium	wikitext-2	1/1	IPMI	00:40:45	1.63M
PyTorch	gpt2-medium	wikitext-2	1/4	IPMI	00:25:11	1.43M
NVIDIA NeMo	distilbert-base-uncased	super_glue (rte)	1/1	IPMI	00:13:04	18.45? *
NVIDIA NeMo	distilbert-base-uncased	super_glue (rte)	1/1	IPMI	00:12:50	226.84K
NVIDIA NeMo	distilbert-base-uncased	super_glue (rte)	1/4	IPMI	00:10:52	123.59K

<sup>\*</sup> Incorrect reporting on consumed energy was detected

Obtained results are not directly comparable due to the different models and datasets used, the results demonstrate a successful analytical data monitoring on HPC Vega.



### Conclusion and Future Work



### Conclusion

We presented management and monitoring of the consumed energy of **HPC Vega** workloads.

- Initially, power management was configured through **BEO** and **RAPL** (both disabled).
- A cluster-specific solution is used to manage power constraints.
- Resource consumption is recorded and logged through Slurm, IPMI, and NVML (GPUs).

This setup gained the capability to monitor and track resources and energy consumption of workloads.



### Future Work

Activities towards emphasizing energy efficiency and sustainability uncovered several possibilities for further evaluation of these essential mechanisms.

- Transition from BEO to Smart Energy Monitors (SEMS).
- RAPL (and MERIC) assessment is already in progress.
- A calibration with a professional power meter to cross-check and validate the data.
- Colledted data will be made available for users through a **portal**.



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# Questions?



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