



European Support Centre for Scalable AI Research and Deployment

Webinar #1 - MINERVA Services - 2 Dec 2025

Agenda

Webinar #1 - Agenda

- Welcome and project introduction
- Overview of MINERVA services
- How to contact MINERVA and request support
- Additional activities: benchmarking, best practices
- Q&A session

Objectives

MINERVA is a distributed, European-wide HPC-enabled AI application support service

It brings together expertise from EuroHPC Hosting Entities and partners representing major European stakeholders in AI.

MINERVA acts as a central hub for cutting-edge European competences in large-scale ML/AI research and development.

It started in January 2025, and the project's duration is 36 months.



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the European Union





Partners



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What MINERVA Aims To Achieve

What we do

- Offer a rich service portfolio covering several levels and types of support.
- Interact with AI communities to identify needs and update the MINERVA service portfolio.
- Increase competences of AI communities on AI n HPC.
- Ensure models are developed according to ethical and responsible AI regulations.

- **Support:** Operating an EU-wide support centre.
- **Knowledge transfer:** Publishing best practice guides and guidelines.
- **Benchmarking:** Evaluating model performance on supercomputers.
- **Data Access:** Providing information on access to public datasets.
- **Training Programs:** Providing "AI on HPC" training program.
- **Community Hub:** Supporting large-scale open-source ML/AI research and development on HPC.

How we do it

The MINERVA Support

To deliver support that is both structured and flexible, MINERVA recognises that users may need help on different topics and in different ways. We organise our services along those two key principles:

Service Levels (L1-L5)

These define the thematic areas where support is offered, reflecting the kinds of challenges AI users typically face, from porting workflows to HPC (L1), scaling models and workflows (L2), to pre-training (L3) and fine-tuning foundation models (L4), as well as complying with ethical and regulatory requirements (L5).

Support Types (T1-T4)

These describe how the support is delivered – from quick technical help (T1) to long-term engineering collaborations (T4).

👉 *Unlike the Levels, these Support Types are an internal coordination tool to help allocate resources and tailor support formats to user needs.*

SERVICES



Support Type	Goal	Scope	Time to resolution	Delivery	Trigger
T1 - Assistance	Quick resolution of specific issues	Debugging, configuration, usage questions	Few days	💻 Remote only	Ticketing system
T2 - Consulting	Strategic guidance for planning and theoretical questions	Proposals, regulation, architecture, resources	Up to 2 weeks	💻 Remote / 📱 On-site optional	Ticketing system Scheduled request
T3 - Workshop	Accelerate workflows through hands-on engineering support	Profiling, performance tuning, scaling	Up to 2 months	💻 Remote / 📱 On-site optional	Ticketing system Evaluation-based selection & Partner availability
T4 - Embedded Support	Long-term collaboration on advanced AI projects	Scaling, fine-tuning, large model training, data handling	Up to 6 months	💻 Remote (with meetups recommended)	Ticketing system Evaluation-based selection & Partner availability

SERVICES & MORE

L1

Porting AI Applications & Workflows to HPC

L2

Optimization & Scaling of AI Libraries on HPC

L3

Pre-training of Large-Scale Foundation Models

L4

Specialization & Fine-Tuning of Foundation Models

L5

Ethical & Responsible AI Support

Specialised/Advanced trainings for the AI communities
ICIAP – EurIPS

Catalogues of models and datasets

Best Practice Guides on AI on HPC

Benchmarks reports

Who can access MINERVA support

MINERVA services are open to all actors in the European AI ecosystem – regardless of technical background or experience with HPC. If you're working with AI and face technical, scaling, or regulatory challenges, MINERVA is here to help.

We welcome requests from:

- **Academic researchers** (PhD students, postdocs, principal investigators)
- **Public sector teams** applying AI in areas like health, climate, education, or policy
- **Start-ups and SMEs*** developing AI-driven products or services
- **Larger companies*** with advanced or specialized AI use cases
- **AI engineers and developers** aiming to scale models or optimize performance
- **Project consortia** preparing proposals for national or European funding calls

📌 No HPC access? We can guide you through the process of applying for resources, choosing the right infrastructure, and getting started with your first project.



Minerva Services

Kamel Guerda
MINERVA WP2 Leader





SERVICES

L1

Porting AI Applications & Workflows to HPC

L2

Optimization & Scaling of AI Libraries on HPC

L3

Pre-training of Large-Scale Foundation Models

L4

Specialization & Fine-Tuning of Foundation Models

L5

Ethical & Responsible AI Support

Guidance and Support on **regulations on ethical and responsible AI**

TYPE 1

Assistance

TYPE 2

Consulting

TYPE 3

Workshop

TYPE 4

Embedded Support



T1 – Assistance: First-Line User Support

Reactive, short interactions to unblock users on concrete issues (jobs, configs, environments, resources).

- **Example with L1 – Porting**
 - ⚠ Problem: Doctoral researcher gets Slurm errors when launching a training job on an HPC cluster.
 - 💡 Support: MINERVA reviews the job script and logs to identify misconfigured directives.
 - 💡 Support: We provide a corrected script and show how to submit and monitor future jobs.
- **Example with L2 – Scaling**
 - ⚠ Problem: SME engineer observes unexpectedly long training times on a multi-GPU setup.
 - 💡 Support: MINERVA suggests profiling tools to locate bottlenecks in the training workflow.
 - 💡 Support: We recommend input-pipeline changes to fix data loading and improve GPU utilization.
- **Example with L3 – Specialization**
 - ⚠ Problem: Team wants to fine-tune a language model but is unsure which model and datasets to pick.
 - 💡 Support: MINERVA leverages ongoing benchmarks and partners' expertise to suggest suitable open models.
 - 💡 Support: We highlight relevant domain datasets and share references to kick-start the fine-tuning effort.



T2 – Consulting: Strategic Guidance Sessions

Structured, planned sessions to provide expert guidance on architecture, scaling, and regulatory questions.

- **Example with L2 – Scaling**

⚠ Problem: Remote-sensing team wants to scale a high-resolution segmentation model to larger datasets and higher input resolution on HPC.

💡 Support: MINERVA performs targeted technology watch on distributed training and I/O bottlenecks for similar use cases.

💡 Support: We give tailored advice on dataset handling, storage layout, and data/model/pipeline parallelism, plus suitable EuroHPC/national resources.

- **Example with L5 – Ethical & Regulatory Compliance**

⚠ Problem: Start-up building clinical decision-support AI needs clarity on how the EU AI Act applies to its use case.

💡 Support: MINERVA reviews documentation and identifies likely risk classification and applicable obligations.

💡 Support: We walk through key requirements, share best-practice examples, and provide a checklist for documentation and future audits.



T3 – Workshop: Hands-On Engineering Sessions

Intensive 1–2 day hands-on sessions to work directly on users' code and workflows on HPC systems.

- **Example with L2 – Optimization**

- ⚠ Problem: Research lab on satellite object detection faces inefficient training on a GPU cluster.
 - 💡 Support: MINERVA reviews training scripts and profiling traces ahead of the workshop.
 - 💡 Support: Together we apply mixed precision, async data loading, and improved GPU scheduling to boost throughput.

- **Example with L3 – Fine-Tuning**

- ⚠ Problem: Private SME with limited AI experience wants to fine-tune a language model on its internal documentation.
 - 💡 Support: MINERVA reviews the existing data, infrastructure, and target use cases before the workshop.
 - 💡 Support: In the session we design the fine-tuning pipeline, apply PEFT (e.g. LoRA), and define evaluation on real internal queries.



T4 – Embedded Support: Medium- to Long-Term Collaboration

Embedded collaboration over several months, with a MINERVA engineer working inside the team on development, optimisation, and research around complex AI workflows.

- Example with L1 – Workflow Adaptation
 - ⚠ Problem: Research group needs to migrate a CPU-based AI workflow to a GPU-accelerated HPC system.
 - 💡 Support: A MINERVA engineer is embedded for three months to review the codebase and refactor the workflow for GPU execution.
 - 💡 Support: Together we build a reproducible pipeline with experiment versioning, leading to a stable, documented HPC setup.
- Example with L3 – Pre-training
 - ⚠ Problem: Public research institution wants to pre-train a legal-domain LLM on EuroHPC resources.
 - 💡 Support: An embedded MINERVA engineer coordinates access, advises on model parallelism, and helps curate and prepare the training dataset.
 - 💡 Support: We integrate logging/monitoring, ensure compliance with EU data-handling rules, and transfer best practices from MINERVA benchmarks.

A typical user journey

How a user can move from first access to HPC to advanced collaboration.

1. First contact – Getting onto HPC (T1 Assistance)

A bioinformatics postdoc contacts MINERVA to access EuroHPC for the first time. Support helps with resource application, environment setup, and first job submissions.

2. Second contact – Becoming efficient (Docs, training, T3 Workshop)

They use MINERVA guides and training to work more autonomously. When performance issues appear, a T3 Workshop is organised to profile the code and optimise GPU and data loading.

3. Third contact – Planning and scaling (T2 Consulting)

As the project grows, they request T2 Consulting to discuss scaling to larger datasets, storage and reproducibility, and relevant EuroHPC calls, leaving with a concise roadmap.



How to Get Support from MINERVA

- For support:
 - minerva4ai.eu/support/#support-form
 - [**support@minerva4ai.eu**](mailto:support@minerva4ai.eu)
- A unified support team will bind expertise and resource overview across all the partners of the project which will act as common main entry for all the users.
- A multi-level, helpdesk support service will have a single point of entry for users, the AISC portal, backed up by a ticketing tool enabling user requests to be processed and the life of a ticket to be tracked through to resolution, while retaining its history.
- The service portal will be integrated with the ticketing system.



Services tools and Benchmarking

David Vicente
MINERVA WP3 Leader





- Implementation of a ticketing system connected to all the Minerva partners and other external entities providing support to AI communities in Europe.
- Generate a collection of benchmarks to measure the performance of AI machines for specific tasks like training and inference .
- Run the benchmarks in the AI systems of EuroHPC to compare hardware and software.
- Generation of best practice guides linked to AI topics

<https://minerva4ai.eu/support/#support-form>

REQUEST SERVICES

Your name

Your email

Your affiliation

Your affiliation type

University or Research Institution

Subject

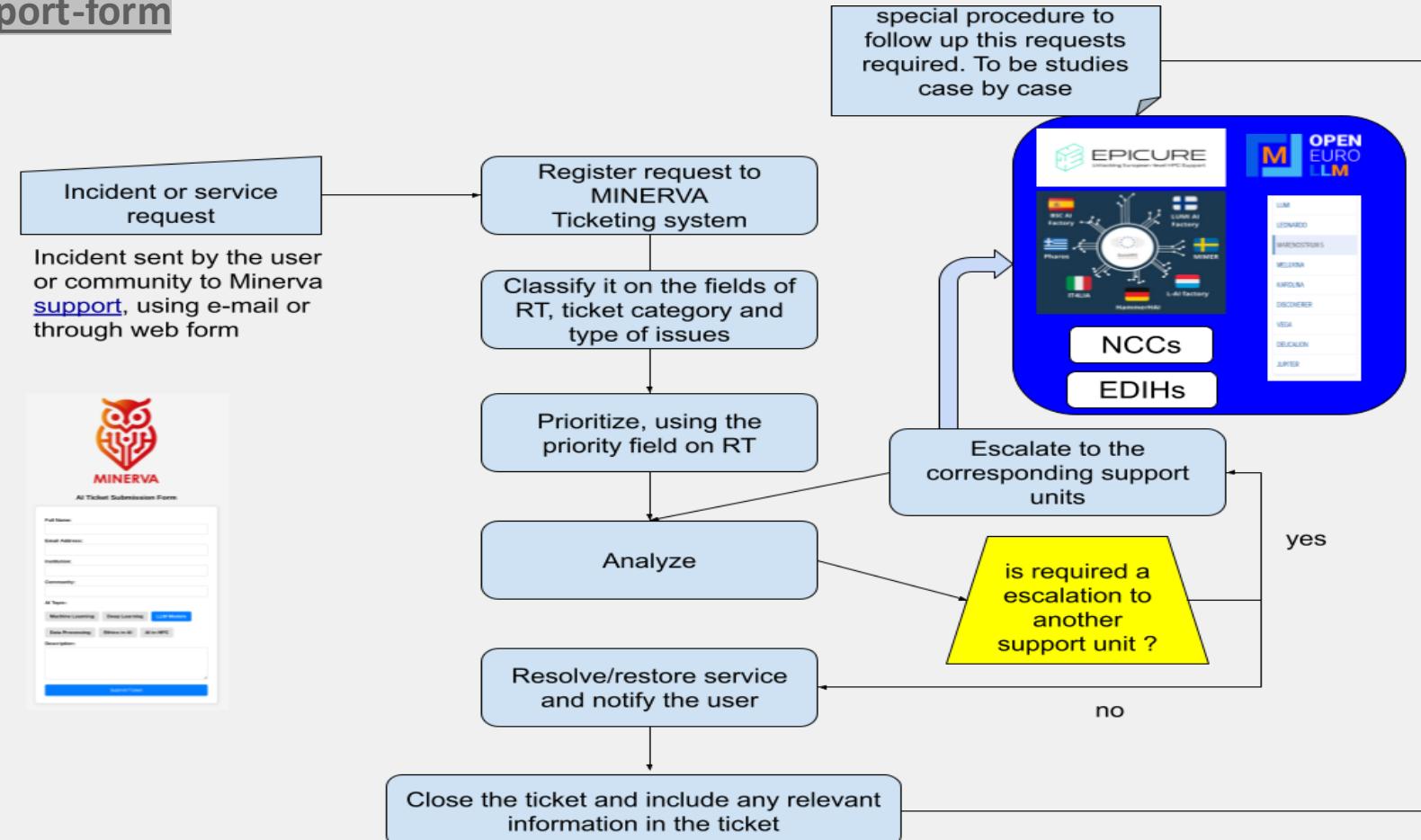
Your message (optional)

*Submitting your enquiry you agree with the submission of your personal data to the partners of the MINERVA consortium and accept their treatment in full respect of the current legislation on privacy – Reg. (UE) 2016/679, as indicated in the Privacy Policy. <https://minerva4ai.eu/privacy-policy-2/>

I give my consent to receive email updates on MINERVA activities or events, even in the form of a periodical newsletter, in the respect of the Privacy Policy <https://minerva4ai.eu/privacy-policy-2/>

Verify you are human

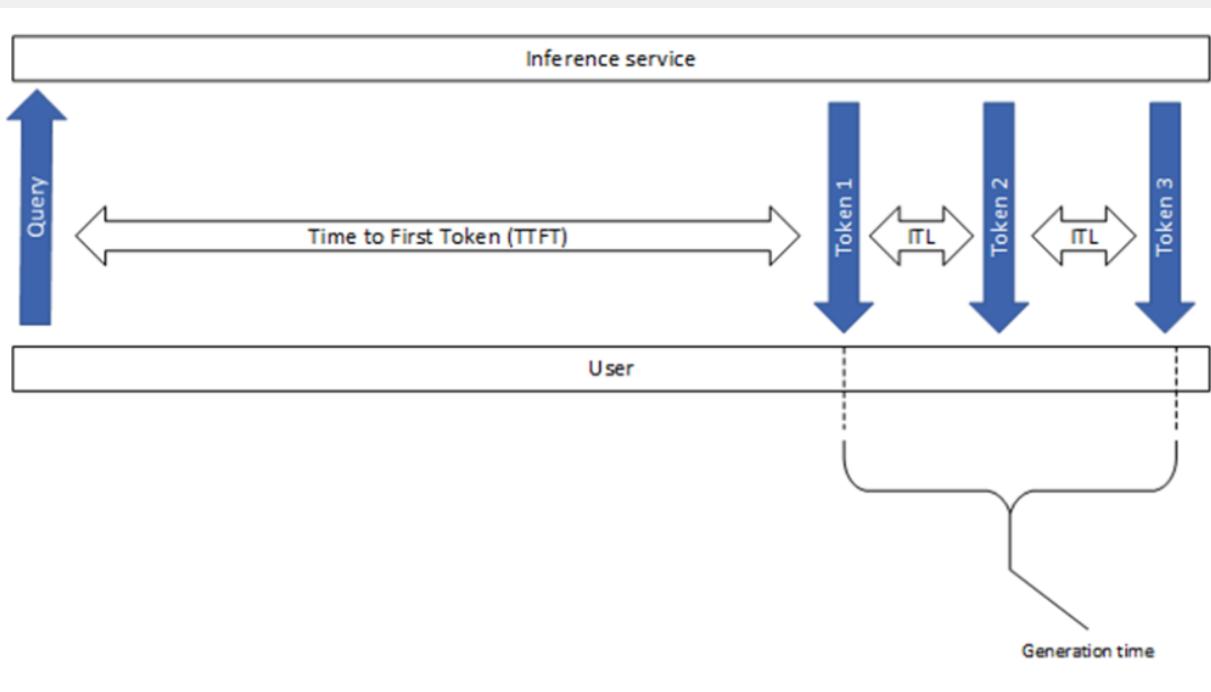
SUBMIT





- Benchmark large AI models during both training and inference using diverse datasets and precision settings.
- Evaluate performance on EuroHPC supercomputers across model sizes and frameworks.
- Develop tools and examples for instrumenting and analyzing foundation model computation.
- Support best practices by contributing to guides and public documentation.
- Utilize consortium-developed and external models as use cases for benchmarking.

- Conda Environments
- Easy to Use Frameworks:
 - vllm
 - DeepSpeed-MII
 - ...
- Examples of Inference



```
iterative-executions-vllm0.6.6.post1 > vllm-0.6.6.post1 > launch-9 > VLLM4BENCH_Nodes_1-GPU
1  Namespace(backend='vllm', base_url=None, host='localhost',
2  Starting initial single prompt test run...
3  Initial test run completed. Starting main benchmark run...
4  Traffic request rate: inf
5  Burstiness factor: 1.0 (Poisson process)
6  Maximum request concurrency: 100
7  ===== Serving Benchmark Result =====
8  Successful requests: 1000
9  Benchmark duration (s): 71.23
10 Total input tokens: 175690
11 Total generated tokens: 193639
12 Request throughput (req/s): 12.05
13 Output token throughput (tok/s): 2718.68
14 Total Token throughput (tok/s): 5185.36
15 -----Time to First Token-----
16 Mean TTFT (ms): 92.31
17 Median TTFT (ms): 40.19
18 P99 TTFT (ms): 813.51
19 -----Time per Output Token (excl. 1st token)-----
20 Mean TPOT (ms): 33.28
21 Median TPOT (ms): 32.93
22 P99 TPOT (ms): 65.98
23 -----Inter-token Latency-----
24 Mean ITL (ms): 32.41
25 Median ITL (ms): 28.43
26 P99 ITL (ms): 94.58
27 =====
```

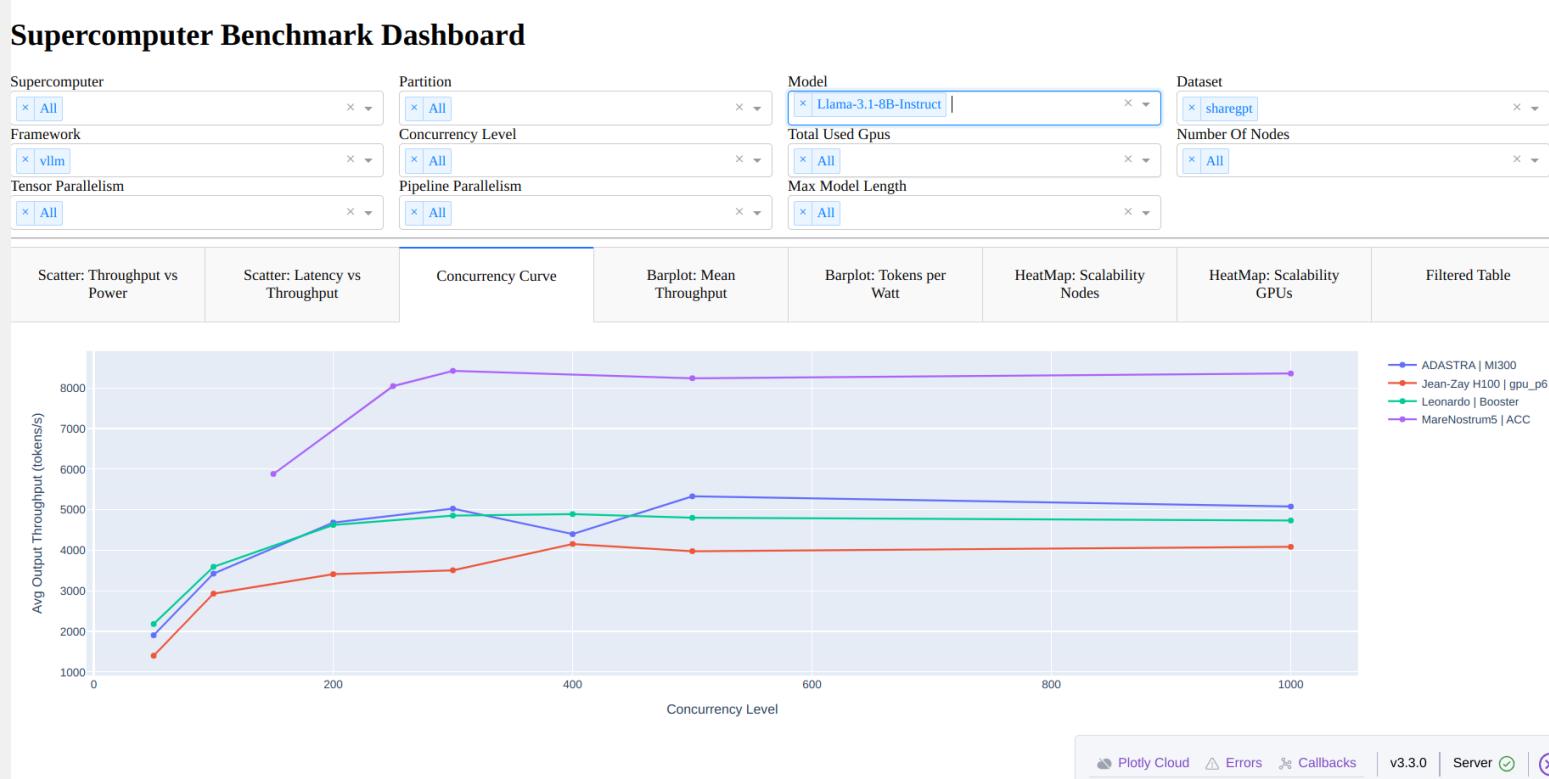
- Conda Environments
- Easy to Use Frameworks:
 - pytorch
 - accelerate
 - deepspeed
- Examples of Training

```
{'loss': 0.0375, 'grad_norm': 0.3546024560928345, 'learning_rate': 7.985285092744795e-06, 'epoch': 5.02}
{'loss': 0.0502, 'grad_norm': 0.3198966979980469, 'learning_rate': 7.018595479690593e-06, 'epoch': 5.04}
{'loss': 0.0295, 'grad_norm': 0.30464619398117065, 'learning_rate': 6.104648424250026e-06, 'epoch': 5.06}
{'loss': 0.0275, 'grad_norm': 0.3265032172203064, 'learning_rate': 5.24612469060774e-06, 'epoch': 5.07}
{'loss': 0.0251, 'grad_norm': 0.46727272868156433, 'learning_rate': 4.4455424767962215e-06, 'epoch': 5.09}
{'loss': 0.0311, 'grad_norm': 0.2630062699317932, 'learning_rate': 3.705250028387089e-06, 'epoch': 5.11}
{'eval_loss': 1.0595293045043945, 'eval_runtime': 8.7248, 'eval_samples_per_second': 229.233, 'eval_steps_per_second': 0.917, 'epoch': 5.11}
{'loss': 0.0244, 'grad_norm': 0.43495360016822815, 'learning_rate': 3.0274187506822483e-06, 'epoch': 5.12}
{'loss': 0.0238, 'grad_norm': 0.3657640516757965, 'learning_rate': 2.4140368396080223e-06, 'epoch': 5.14}
{'loss': 0.0232, 'grad_norm': 0.4130260944366455, 'learning_rate': 1.8669034499939063e-06, 'epoch': 6.01}
{'loss': 0.0333, 'grad_norm': 0.3001957833766937, 'learning_rate': 1.3876234183414994e-06, 'epoch': 6.03}
{'loss': 0.0247, 'grad_norm': 0.3097127377986908, 'learning_rate': 9.776025555625068e-07, 'epoch': 6.04}
{'loss': 0.0209, 'grad_norm': 0.5588397979736328, 'learning_rate': 6.380435234931342e-07, 'epoch': 6.06}
{'loss': 0.0182, 'grad_norm': 0.3872393071651459, 'learning_rate': 3.6994230727969705e-07, 'epoch': 6.08}
{'loss': 0.0191, 'grad_norm': 0.3198922276496887, 'learning_rate': 1.740852939825899e-07, 'epoch': 6.09}
{'loss': 0.0268, 'grad_norm': 0.38139650225639343, 'learning_rate': 5.104696596753278e-08, 'epoch': 6.11}
{'loss': 0.0233, 'grad_norm': 0.7401021122932434, 'learning_rate': 1.1882158498205398e-09, 'epoch': 6.13}
{'eval_loss': 1.0987142324447632, 'eval_runtime': 8.7434, 'eval_samples_per_second': 228.743, 'eval_steps_per_second': 0.915, 'epoch': 6.13}
{'train_runtime': 28931.0222, 'train_samples_per_second': 26.546, 'train_steps_per_second': 0.104, 'train_loss': 0.2665035105148951, 'epoch': 6.13}
```



Benchmark Expansion

- Integrate standardized benchmarks:
 - MilaBench – Advanced LLM Training benchmarking suite.
 - MLPerf – Industry-standard ML performance benchmarks for Inference/Training.
 - Others
- More diverse workloads:
 - Reinforcement Learning.
 - Prioritize emerging European models and datasets for upcoming benchmarks.
- Public Release & Collaboration:
 - Publish results on MINERVA github (<https://github.com/minerva4ai-eu/minerva-benchmarks/>).
 - Open-source selected tools and configs for community reproducibility.
 - Prepare materials for Best Practices documentation.



Output filter features:

Dynamic Filters: Supercomputer, Frameworks, Models, Datasets, Parallelism Type, # GPUs, # Nodes.

Inference:

- Throughput vs Concurrency Level
- Latency vs Throughput
- Scalability Heatmaps

Training/Fine-Tuning:

- Throughput Scaling
- GPU efficiency
- Total Training Time comparison



Best Practice Guidelines – 2 collections

Soon available on MINERVA website

"Optimizing AI Models on European Infrastructure"

-  *"Infrastructure Foundations for Scalable LLM Training"*
-  *"Efficient and Reliable LLM Pretraining"*
-  *"Post-Training: Alignment and User-Centered Objectives"*
-  *"Explainable AI"*

Status: Writing

Publication: June 2026

"Dataset Management for European AI Ecosystems"

-  *"Dataset Sourcing: Where and How to Collect Data"*
-  *"Cleaning, Preprocessing, and Curation"*
-  *"Quality, Ethics, and Compliance"*
-  *"Publishing, Reuse, and Maintenance"*

Status: Writing

Publication: December 2027



Questions?

LinkedIn



Survey

Reach out to us @

info@minerva4ai.eu

Thank you



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