

DAIMOS: Distributed AI Model training Optimization at Scale

Exa-MA Annual Meeting 2026

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Axe 1.2 : HPC pour l'apprentissage haute performance et une IA très grande échelle

Consortium : IRIT-INP, Inria Bordeaux, Sorbonne Université, Météo-France

The Challenge

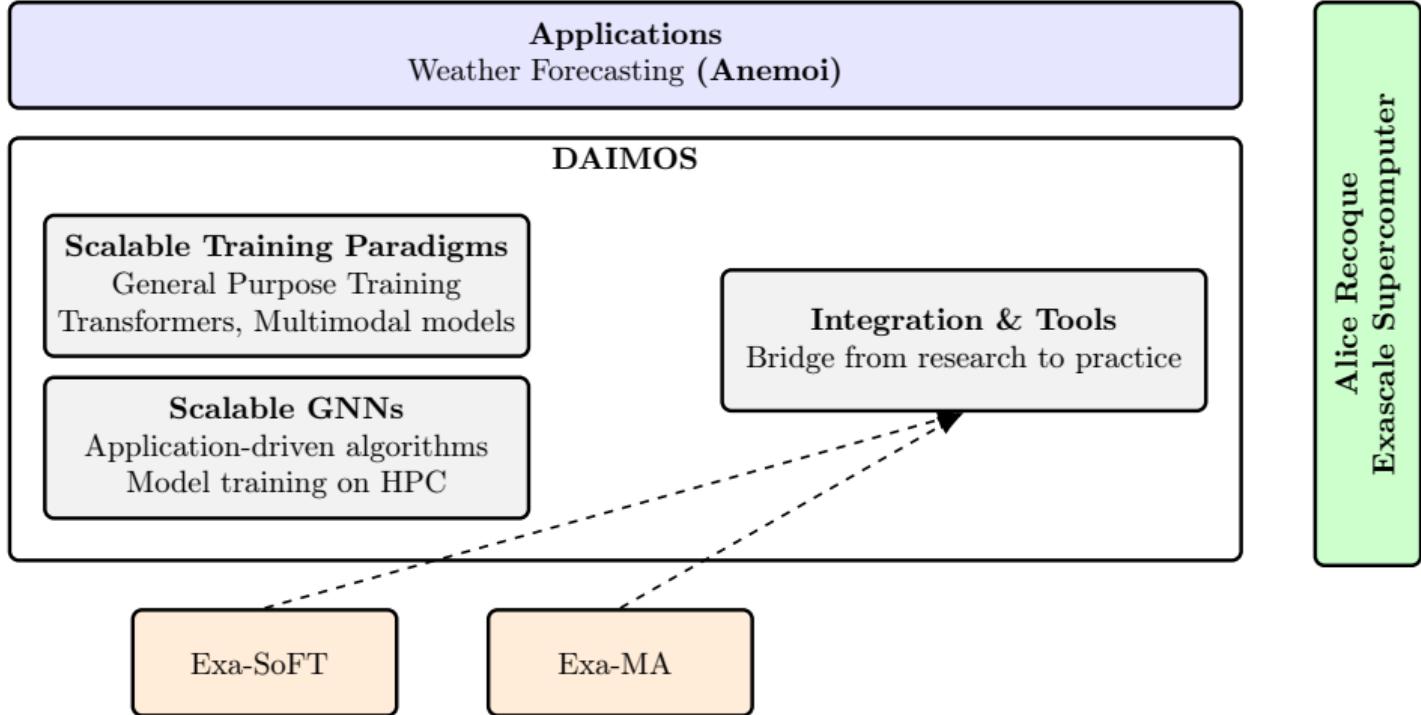
- **Explosion of training cost:** Modern AI models demand massive compute and memory; scaling is hard across heterogeneous, multi-GPU, multi-node systems.
- **Scientific aspect:** Training time, memory bottlenecks, and limited portability slow down innovation.
- **Where it hurts most:**
 - Parallelization techniques (data-, model-, tensor- parallelism) are understood individually but hard to co-optimize.
 - Graph and multi-modal models suffer from irregular structures, *communication* and *memory* bottlenecks.
- **Opportunity:** Harness Exascale HPC systems with smarter *algorithms + systems* co-design.
- **High-impact:** Weather forecasting is a flagship use case: frequent retraining on huge datasets; accuracy benefits from higher spatial resolutions.

Our vision

- **Goal:** Make distributed training of complex AI models **scalable, portable, and resource-efficient** on national HPC systems.
- **Main scientific objectives:**
 - Design efficient distributed training algorithms (including multilevel domain decomposition and alternative to back-propagation).
 - Tackle GNN & multi-modal training bottlenecks (memory, communication, scheduling).
 - Deliver a **modular, reusable software stack** for HPC environments.
 - Integration and evaluation of these contributions in a real-world application scenario, namely, GNN-based weather forecasting
- **Why it matters:** Beyond weather, benefits extend to bioinformatics, networks, climate modeling, and large scientific AI workloads.

DAIMOS – overview

NumPEx



Scalable Training for Modern DL Architectures

Focus HPC for AI: advancing scalability of DL models

- Reduce memory, communication and synchronization bottlenecks
- Extend existing automated optimization tools to multimodal models
- Propose alternative to backpropagation

Contributors Sorbonne (E. Oyallon), Inria (J. Gusak)

Resources 1 PhD, 2 years postdoc



Graph Neural Networks (GNNs)

Focus Exploit specific structure of GNNs to improve scalability

- Load-balancing algorithms targeting the irregular structure of GNNs
- Multi-Level Domain Decomposition training algorithm for GNNs

Contributors Inria (L. Eyraud-Dubois), IRIT (A. Kopanicáková)

Resources 2 PhD



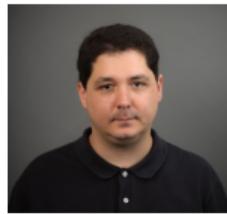
Integration and Reusable Tools

Focus Integration and evaluation of techniques

- New graph partitioner tailored for irregular GNN structures
- Adaptative communication
- Integration of algorithms in **Anemoi**
- Benchmark and demonstrators

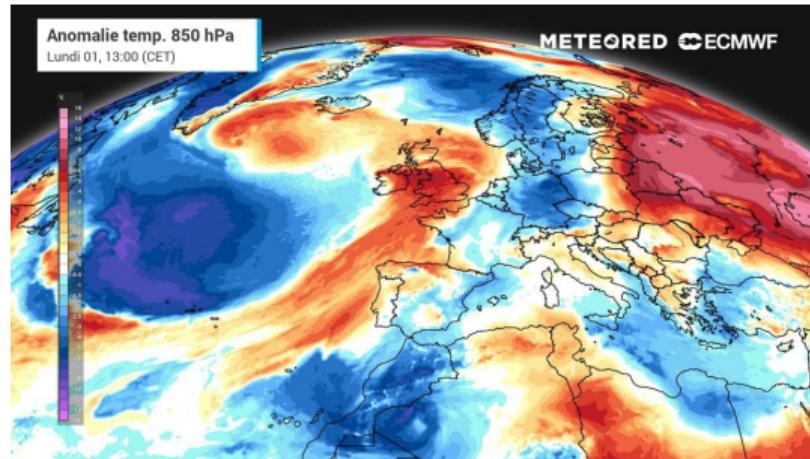
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The logo of Inria, featuring the word "inria" in a large, flowing red script font.

