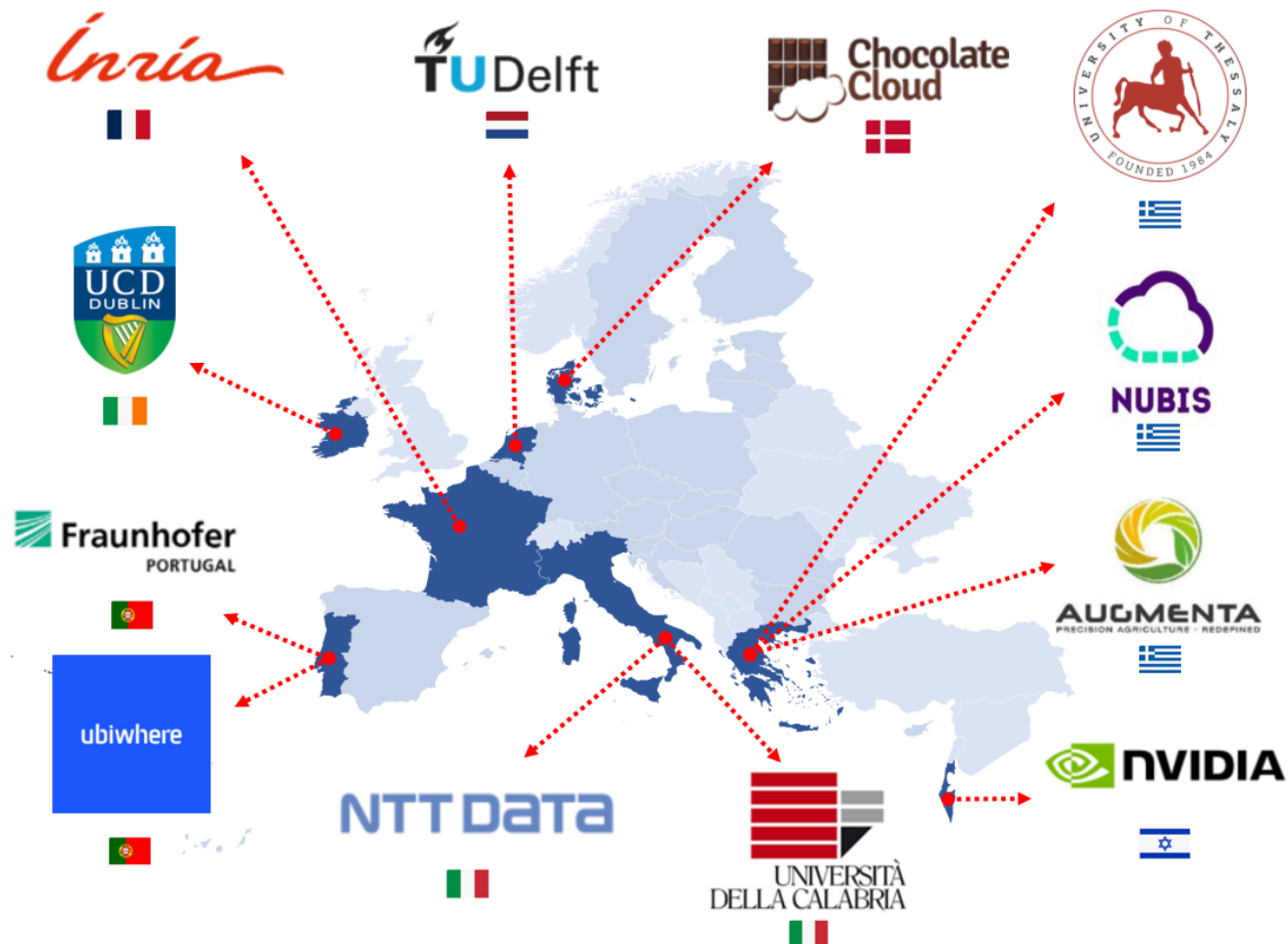


MLSysOps

Machine Learning for Autonomic System Operation in
the Heterogeneous Edge-Cloud Continuum

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Trends

Applications have moved outside the cloud

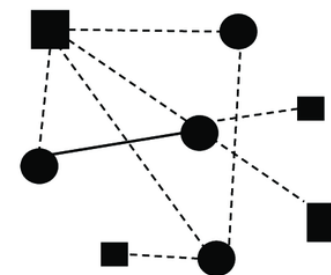
- Involving nodes and resources toward or directly at the edge of the Internet...
- ... and including powerful but also resource-constrained IoT devices

Impossible to monitor/manage large systems by rule-based approaches

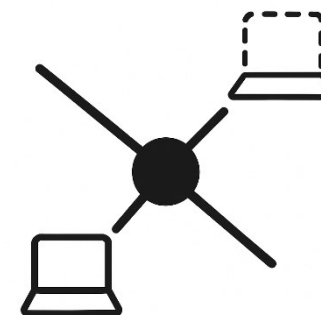
Challenges



Scale



Heterogeneity

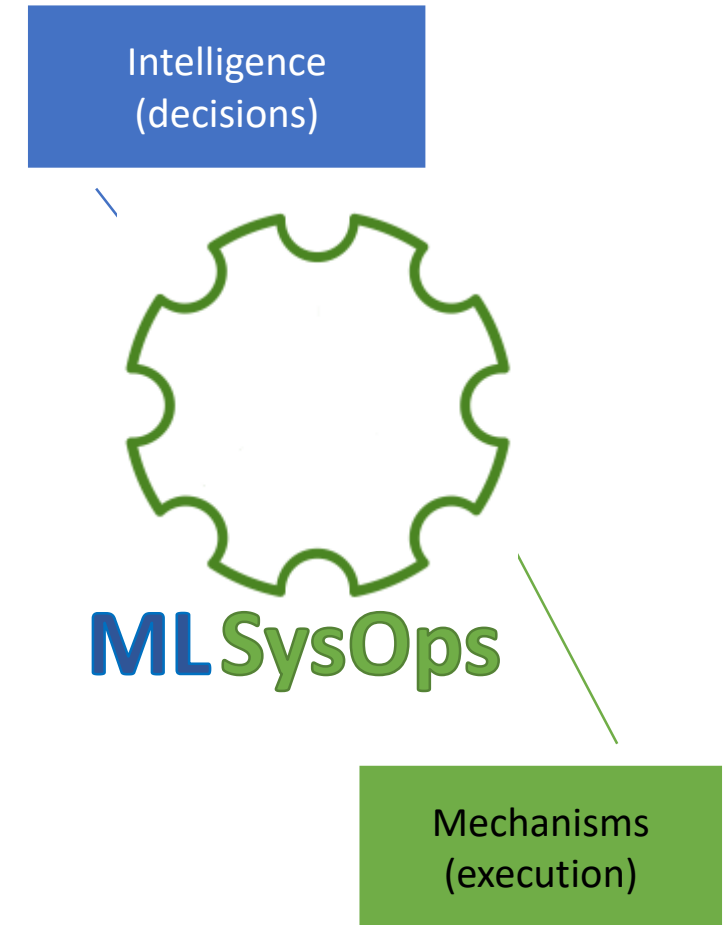


Volatility

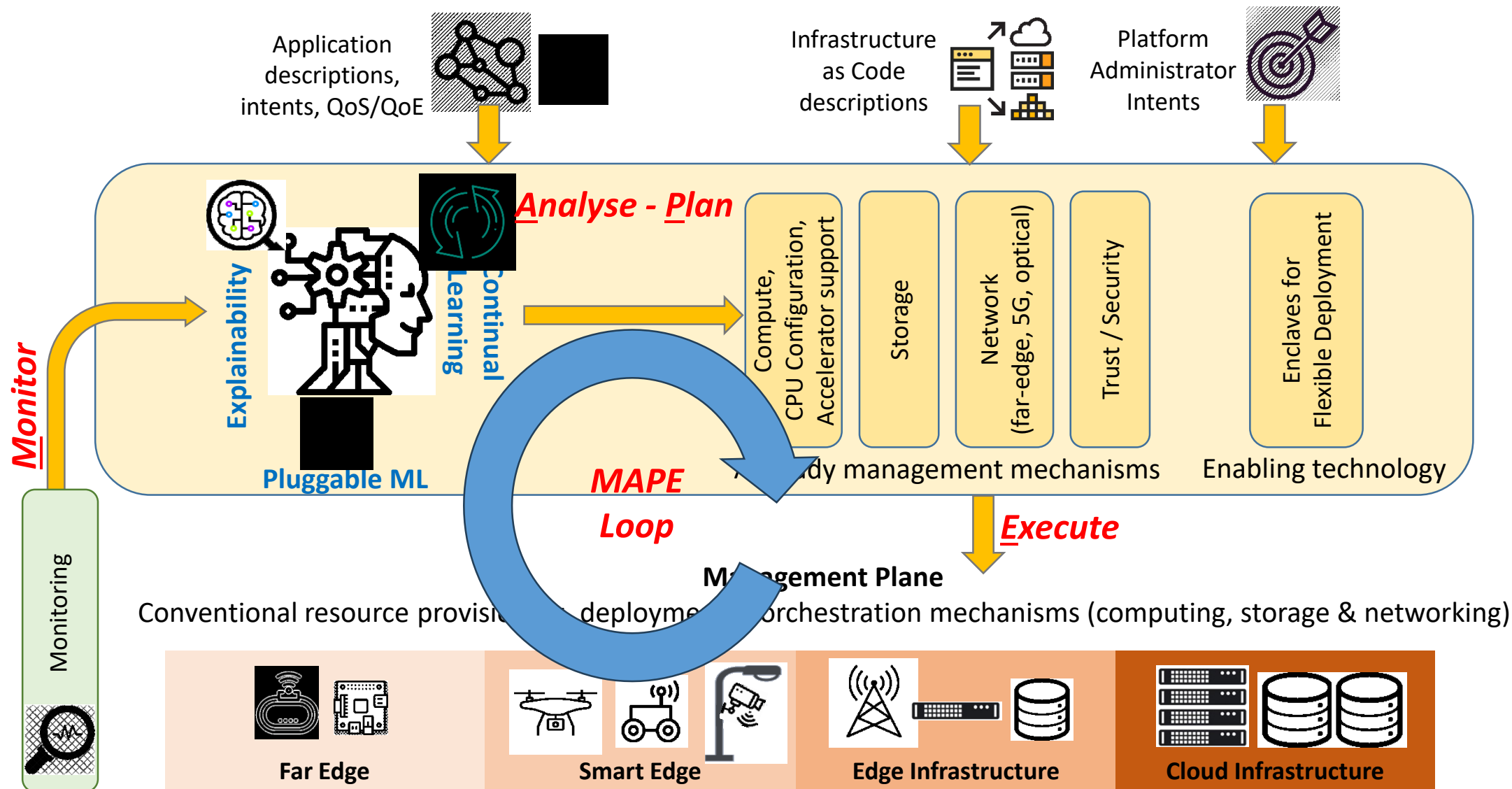
The MLSysOps project

*Autonomic system management and configuration
in the **cloud-edge-IoT continuum** using **AI/ML methods***

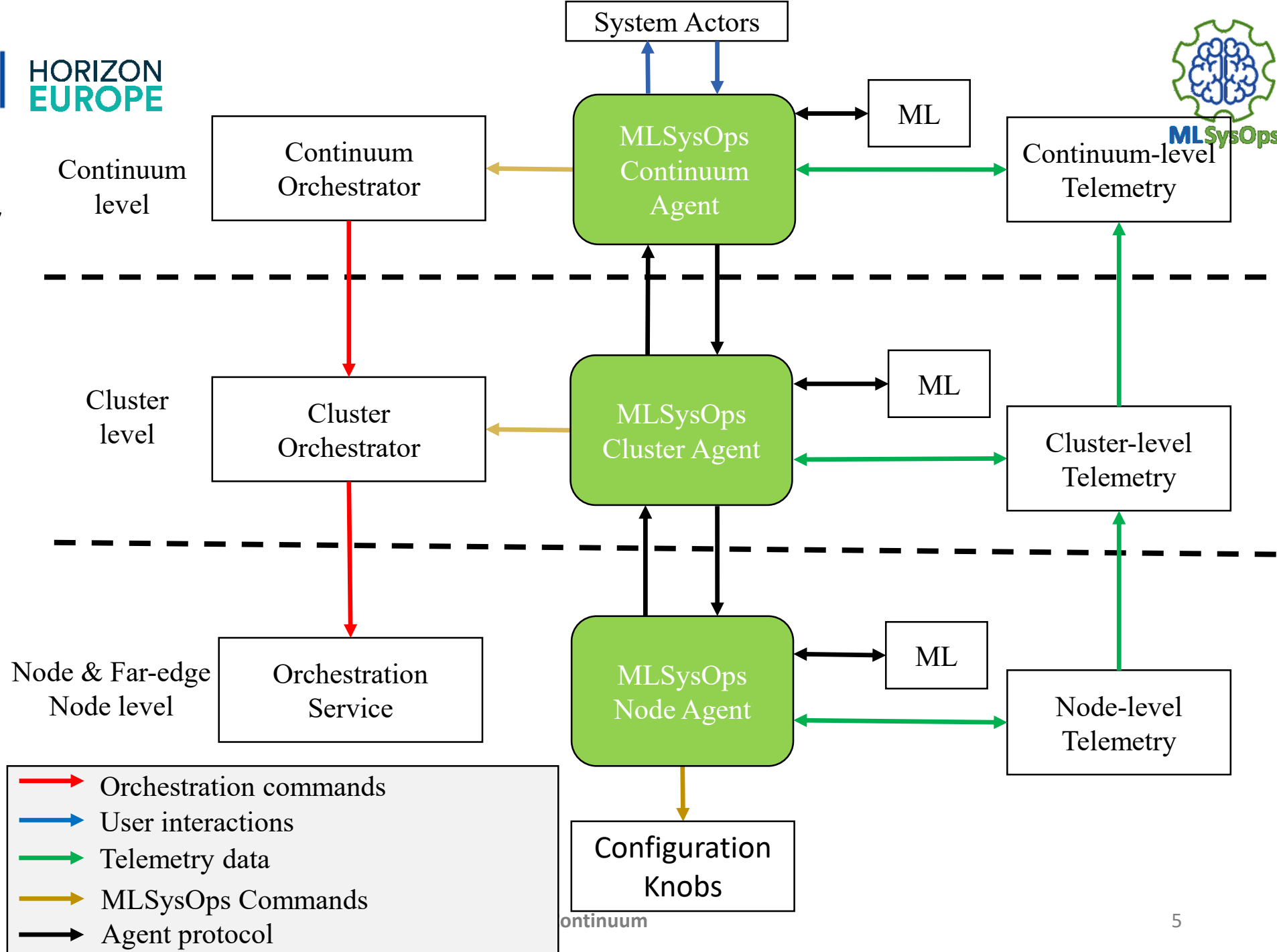
- Focus on modular, distributed applications
 - Comprised out of independently executable components (containers)
- Explore different management aspects
 - Deployment, computing, storage, communication/networking, trust
- Disassociate management from control
 - Develop AI/ML-ready (policy-neutral) mechanisms
 - Take decisions using suitable ML models
- Key AI/ML properties
 - Distributed / hierarchical approach
 - Continual learning / efficient model retraining
 - Explainability
- Extensive data generation evaluation via real-world application testbeds, research testbeds and simulators



How? Concept



How? Hierarchy



Two real-world application use cases

Smart Cities



Improve existing system via AI/ML-driven control to manage/configure application modules that can be deployed on smart lampposts and/or to datacenters.

Smart Agriculture



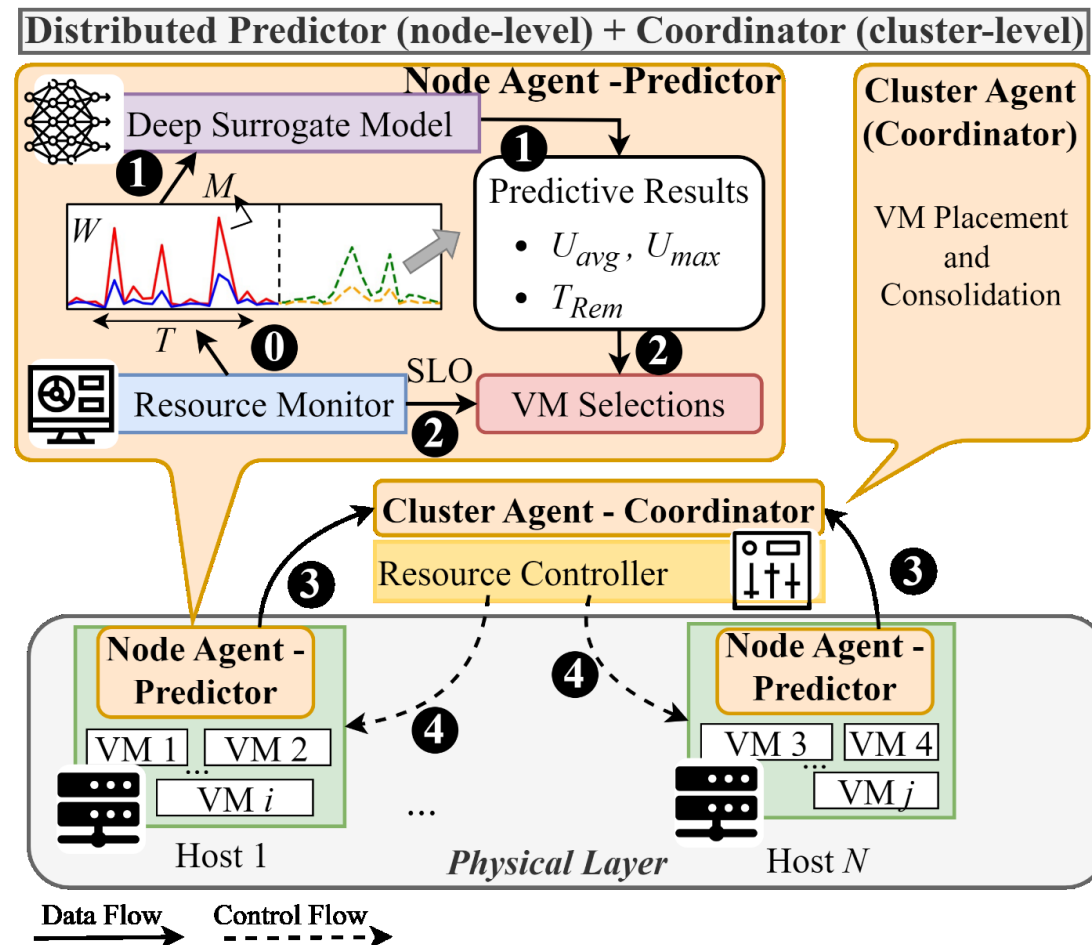
Improve existing system via AI/ML-driven control to jointly manage/configure the image processing application pipeline on a tractor and a companion drone.

One of the many outcomes

Can we do better with ML-based resource
management for the Cloud?

Solution Overview

- Node-level ML model
 - Runs locally on each node
 - Predicts future CPU utilization and VM lifetime
 - Drives VM management decisions
- Cluster agent
 - Receives predictor output
 - Applies a heuristic to globally optimize cluster resources



Results

- Migration counts
 - **Decrease of 41.33%** compared with SoTA conventional (non-ML) policy
 - Only 7.5% over oracle (Ground Truth)
- SLO violations
 - **Lower by 34.98%** compared with SoTA conventional policy
 - A mere 1.1% more than oracle (Ground Truth)

Coming soon...

- Workshop organization

- 2nd ML4ECS @ HiPEAC 2026 (Jan. 26, 2026)
- In collaboration with other projects of the cluster

<https://ml4ecs.e-ce.uth.gr/>

- Hackathon organization

- University of Calabria, Rende, Italy (Sep. 26-27, 2025)
- Familiarize developers with the MLSysOps framework
- Foster application & ML models development



Open-Source Release

Open-source components of the MLSysOps framework

<https://github.com/mlsysops-eu/mlsysops-framework>

Core Components

- Hierarchical Agent Architecture: Interfaces with orchestration/control systems and exposes an ML-model API for plug-and-play explainable/re-trainable models.
- Telemetry & Control Knobs: Collects metrics across the continuum and adjusts configuration (e.g., compute, network, storage, accelerator usage) dynamically.
- Distributed FaaS-style Executor: Enables function offloading across tiers to optimize latency, energy, and performance.
- Explainable ML & Reinforcement Learning Module: Offers transparent decisions, highlighting input factors influencing agent actions.
- Use-cases: Includes real applications focusing on smart cities and agriculture.



<https://mlsysops.eu>



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Newsletter

<https://mlsysops.eu/communication/>

Thank
you!