

Advanced Al for scientists: the Al4EOSC platform approach

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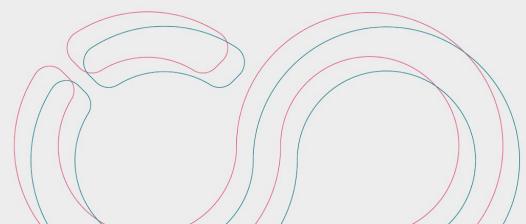
Advanced Computing and e-Science group Institute of Physics of Cantabra (IFCA) - CSIC | UC













Artificial Intelligence for the EOSC

- Evolution of the DEEP Hybrid DataCloud platform
- Runs September 1st 2022 August 2025 (36 months)
- 7 academic + 2 SME + 1 non-profit organization

Advanced features for distributed, federated, composite learning, metadata provenance, MLOps, event-driven data processing, and provision of AI/ML/DL services

















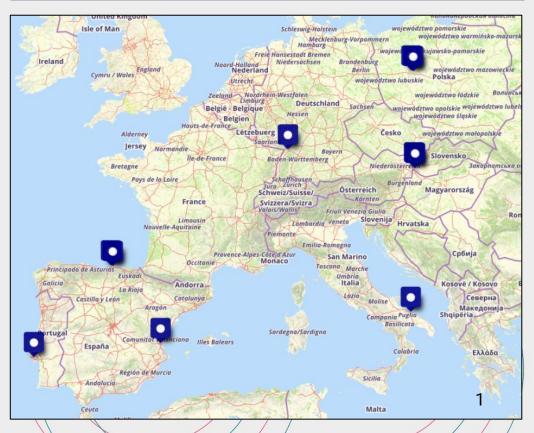






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- Funding: 5M€
- 3 workshops on AI, image processing, federated learning
- 1 external users open call
- 8 peer reviewed publications in high impact journals
- 2 peer reviewed publications in high impact conferences
- Collaboration with several EU funded and INFRAEOSC projects





AI4EOSC main objectives

- 1. Feature rich services and platform to build and deploy custom Al applications in the EOSC
- 2. Support for **building AI systems on distributed datasets**, with a particular focus on **federated learning**
- 3. Services to **compose AI tool workflows**, enabling the development of complex data-driven AI applications
- 4. **Al Exchange Hub** in the context of the EOSC, enhancing and increasing the application offer currently available
- 5. **Extend** the service offer and the **capabilities** being offered through the **EOSC portal**, with focus on Al



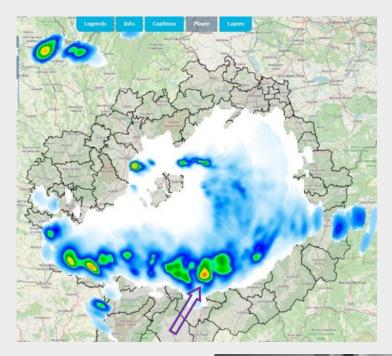


AI4EOSC use cases:

Agrometeorology

Integrated plant protection

Automated thermography









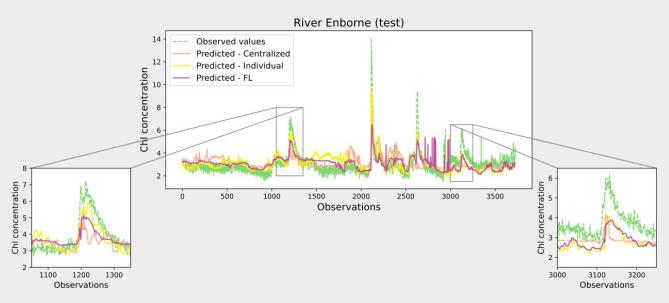




Additional use cases

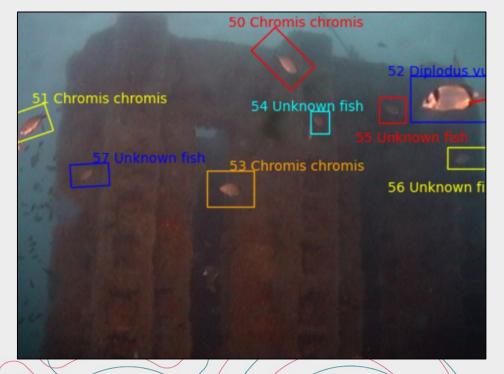
Water quality: predicting Chl concentration

Fish detection and classification



Sáinz-Pardo Díaz, Judith, Castrillo, María and López García, Álvaro. "Deep learning based soft-sensor for continuous chlorophyll estimation on decentralized data." Water Research 246 (2023): 120726. https://doi.org/10.1016/j.watres.2023.120726





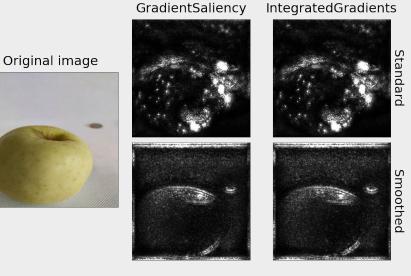
Martinez, Enoc and Valentin Kozlow. OBSEA Fish detector. https://github.com/EnocMartinez/obsea-fish-detection



Additional use cases

Medical imaging using federated learning

Al for precise weight measurement of fruits

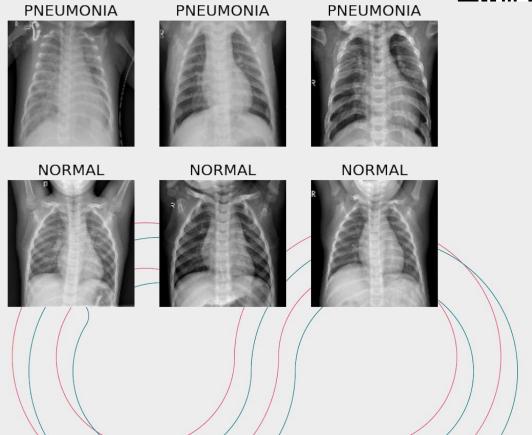


Izquierdo, Pablo, from CSIC DigitalAlimenta project. https://digitalalimenta.csic.es/



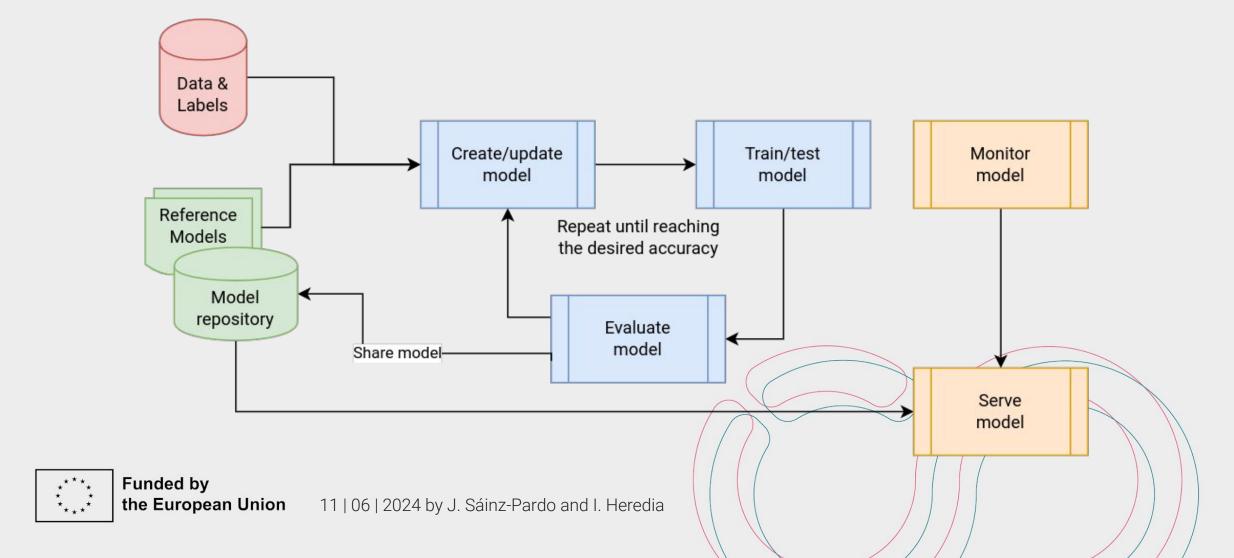
Sáinz-Pardo Díaz, Judith, and López García, Álvaro. "Study of the performance and scalability of federated learning for medical imaging with intermittent clients." *Neurocomputing* 518 (2023): 142-154. https://doi.org/10.1016/j.neucom.2022.11.011





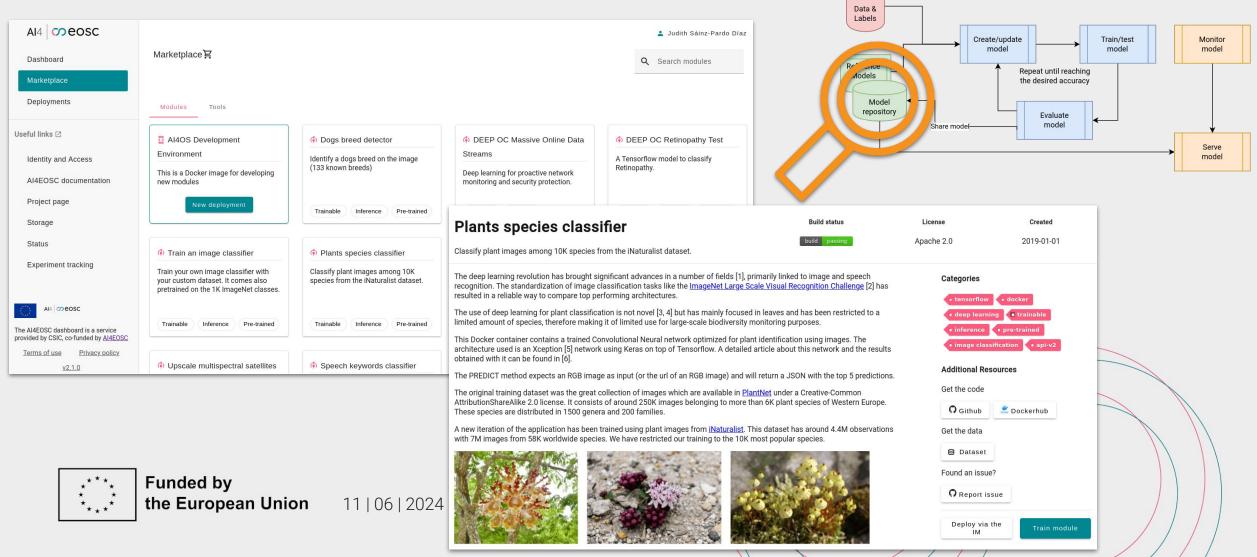


The Machine Learning Lifecycle



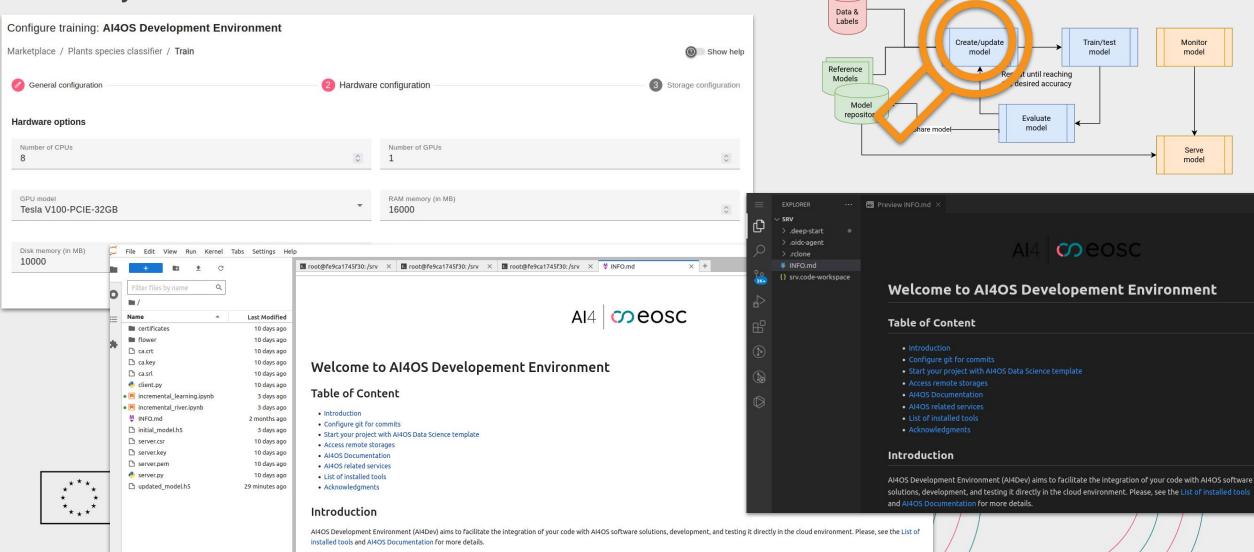


Model marketplace





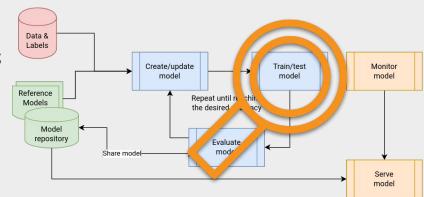
Create your own model





Training: federated learning

- Collaborative and decentralized approach to build ML models
 - No need to centralize a dataset (i.e. technical or privacy restrictions)
- Management of experiments through platform dashboard
- Participating clients both within AI4EOSC platform or external (with authentication)

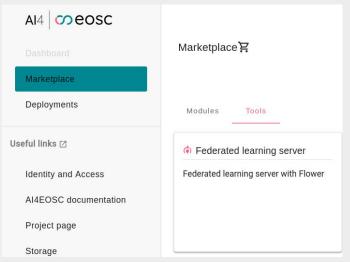


CLIENT 1

CLIENT 2

SERVER

CLIENT N



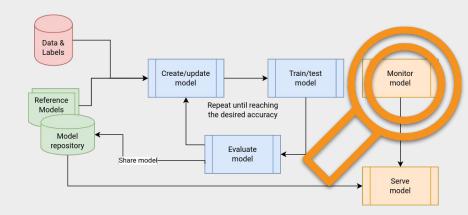
root@8878206726fb:/srv# cd federated-server/fedserver root@8878206726fb:/srv/federated-server/fedserver# python3 server.py INFO flwr 2024-04-18 14:10:23,381 | vault.py:76 | Configured Vault Bearer token INFO flwr 2024-04-18 14:10:23,381 | vault.py:77 | Reading tokens stored in: 'us a3f-0242ac130005/federated Getting tokens from Vault -> users/7d7a87545b700b38b54e2b5b4713084fd2b8d7e5ed82 INFO flwr 2024-04-18 14:10:23,790 | vault.py:79 | Configured Vault Bearer tokens 86bf6d2843ad31a8879a5fab0c1318', '6707ded3dab865271f4a5ac637601cbb38269b808f2029 INFO flwr 2024-04-18 14:10:23,790 | server.py:80 | Token interceptor created INFO flwr 2024-04-18 14:10:23,791 app.py:158 | Starting Flower server, config app.py:172 | Flower ECE: gRPC server runnin INFO flwr 2024-04-18 14:10:23,800 INFO flwr 2024-04-18 14:10:23,800 server.py:91 | Initializing global paramete server.py:282 | Requesting initial paramete INFO flwr 2024-04-18 14:10:23,800 server.py:288 | Received initial parameters INFO flwr 2024-04-18 14:10:49,423 server.py:93 | Evaluating initial parameters INFO flwr 2024-04-18 14:10:49,424 INFO flwr 2024-04-18 14:10:49,424 server.py:106 | FL starting

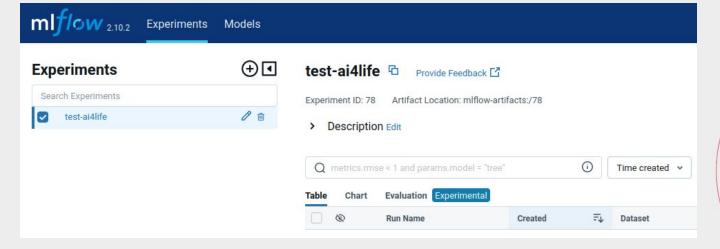


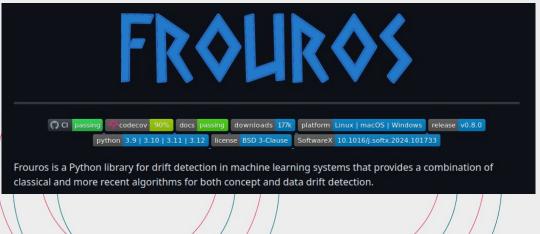


Monitor the model: MLOps and drift detection

- Monitoring of models in production is not enough
 - Model learns from data, data is not stationary
 - Concept learnt by the model may change over time
- Data and concept drift detection
 → essential to build more robust models
- Frouros: state-of-the-art Python library for drift detection in ML problems: https://github.com/IFCA/frouros
- MLOps is an engineering practice that aims to automate and streamline the ML lifecycle
- MLflow tracking server: https://mlflow.cloud.ai4eosc.eu









Serve the model: inference

https://inference.cloud.ai4eosc.eu/ui/

We use <u>OSCAR</u> to serve AI models for inference (AI as a Service). It supports two serverless event-driven execution modes:

- Asynchronous mode: Files uploaded to the object-store
 automatically trigger the invocation of a data-processing script,
 that is run inside a container (out of user-defined Docker image)
 within a scalable Kubernetes cluster (e.g. batch jobs).
- **Synchronous mode**: Scalable HTTP-based endpoints (based on KNative). Direct requests to the model.

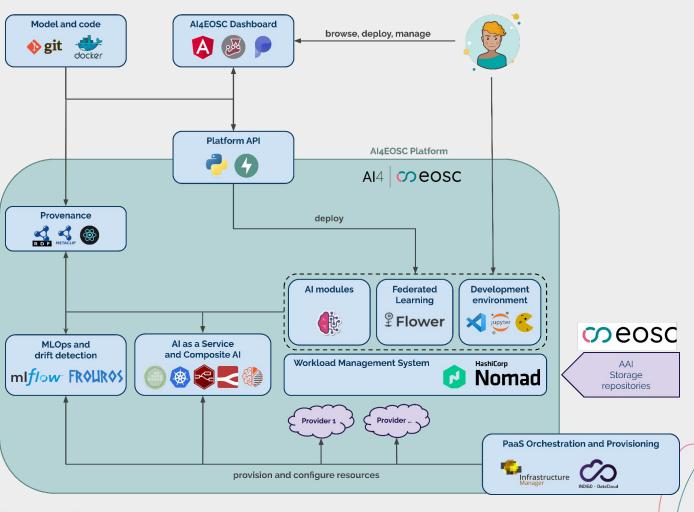
On top, we support building AI workflows using <u>Flowfuse</u> and <u>Elyra</u> through <u>AI4Compose</u>.



Labels Train/test the desired accuracy Model repository model AI4 cosc **OSCAR** Password IGN IN VIA EGI CHECK-IN



AI4EOSC high level architecture



Interactive C4 diagrams available here.







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Showcasing the AI4EOSC platform



https://dashboard.cloud.ai4eosc.eu





Conclusions and collaboration

AI4EOSC empowers scientific research by...

- Providing users with advanced AI tools:
 - Model retraining (iterative learning, fine tuning)
 - Federated learning (including client authentication)
 - Parallel training in multiple GPUs (distributed training data parallelism)
 - o Model monitoring: MLOps, drift detection
 - Model inference
- Providing a simple and intuitive IDE for developing AI/ML/DL models (VSCode or JupyterLab)
- Allowing seamless access to computational resources to accelerate model development
- Deploying your models in production in a serverless environment

Funded by the European Union

More info:

https://ai4eosc.eu/ https://docs.ai4os.eu/



Thank you for your attention!

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