

Finanziato dall'Unione europea NextGenerationEU







### UNLOCKING THE COMPUTE CONTINUUM SCALING OUT FROM CLOUD TO H{P,T}C RESOURCES

October 19 - 25, 2024

# CHEP 2024

ICSC Italian Research Center on High-Performance Computing, Big Data and Quantum Computing

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in the middle





### CONTINUUM

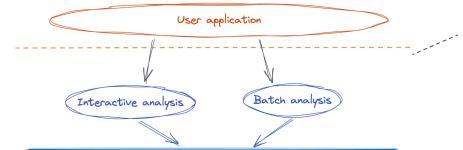


#### Heterogeneous use cases

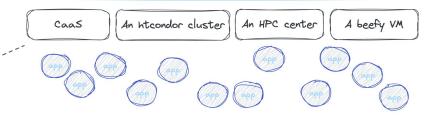
- Coffee-break time long analyses at LHC
  - cloud + scale out in high rate frameworks

### **GPU accelerated HEP simulations and triggers**

- develop on cloud, deploy at scale (on grid?)
- ML training and inference of particle tagging
  - Cloud based data-science frameworks Our continuum idea is here hungry for HPC hosted GPUs
- Well known grid-friendly workflows



Continuum --> I don't care what you do, just do it



<u>Heterogeneous</u> resources managed by <u>heterogeneous</u> systems

HPC  $\rightarrow$  mainly SLURM  $\rightarrow$  any grid batch system (HTCondor, ARC, ...) HTC  $\rightarrow$  Kubernetes Cloud

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ML and GPUs	challenging problem. Reconstructing the tracks from the hits created by those generated particles on the detector layers the indication energy denotation energy denotation and the detector layers at the indication energy denotation and the detector layers at the second secon	You can repeat th
In ATLAS and other high-energy physics experiments, the integrity of Monte-Carlo (MC) simulations is crucial for reliable physics analysis. The continuous evolution of MC generators necessitates regular validation to ensure the accuracy of simulations. We introduce an enhanced validation framework incorporation the Job Execution Monitor (JEM) resulting in	Tori Jeske     Q 23/10/2024, 14:24      Track 2 - Online and real Talk Parallel (Track 2)      Tracking charged particles resulting from collisions in the presence of strong magnetic field is an important and	The's z= ouline and reads— Tak Paralise (track z) The High-Luminosity LHC upgrade will have a new trigger system that utilizes detailed information from muon and track finder subsystems at the bunch crossing rate, which enables the final stage of the Lev Global Trigger (GT), to use high-oracision trigger objects. In addition to cut-based algorithms, novel ma
Mustafa Andre Schmidt (Bergische Universitaet Wuppertal (DE)) U1/10/2024, 15:18 Track 5 - Simulation and Poster Poster session	Within the ROOT/TMVA project, we have developed a tool called SOFIE, that takes externally trained deep learning models in ONNX format or Kersa and PyTorch native formats and generates C++ code that can be easily included and invoked for farst information of the model. The code bases minimal decendency and can be easily interreted into the date 271. ML-Assisted Charged Particle Tracking at GlueX	358. Use of topological correlations in ML-based conditions for the CMS Level-1 Glob upgrade for the HL-LHC & Gabriele Bortolato (Universita e INFN, Padova (IT)) O 24/10/2024, 17:27 Track 2- Online and real- Table   Parallel (Track 2)
management features users can optimise data access to meet specific requirements and compliance standards. Given 20. Streamlining ATLAS Monte-Carlo Generator Validation with PAVER (МОN 33)	© 23/10/2024, 13:30 Track 5 - Simulation and Talk Parallel (Track 5)	quantum computing middleware framework. OlboML incorporates the most commonly used classical Machine Learning frameworks such as Tense
Maker to bata and measure rooten research Amazon S3 is a leading object storage service known for its scalability, data reliability, security and performance. It is used as a storage solution for data lakes, websites, mobile applications, backup, archiving and more. With its	356. Benchmark Studies of ML Inference with TMVA SOFIE	© 24/10/2024, 16:51 Track 5 - Simulation and Tak Parallel (Track 5) We present QiboML, an open-source software library for Quantum Machine Learning (QML) integrated
Andreas Joachim Peters (CERN), Elvin Alin Sindrilaru (CERN), Luca Mascetti (CERN) O 21/10/2024, 15:18 Track 1 - Data and Meta. Poster Poster session	Detector simulation is a key component of physics analysis and related activities in CMS. In the upcoming High Luminosity LHC ers, simulation will be required to use a smaller fraction of computing in order to satisfy resource constraints. At the arms the CMS will be upmarked with the new High Computing Calorimeter (HCCA), which requires	183. QiboML: a full-stack quantum machine learning framework ▲ Matteo Robbiati (Università degli Studi e NKHA Milano (IT))
60. S3 Compatibility: Enabling Seamless Integration with EOS, CERN's Large-Scale Disk Storage System (MON 30)	© 22/10/2024, 16:33 Track 5 - Simulation and Talk Parallel (Track 5)	The LHCb experiment requires a wide variety of Monte Carlo simulated samples to support its physics LHCb's centralised production system operates on the DIRAC backend of the VILCG; users interact with web availation to require and produce samples.
	119. Simulating the CMS High Granularity Calorimeter with ML. Kevin Pedro (Fermi National Accelerator Lab. (US))	© 24/10/2024, 15:18 Track 4 - Distributed Co Poster Poster session
The Set or Simulation and take Paramet (Track 5) The Fair Universe project is organising the HiggsML Uncertainty Challenge, which will/has run from June to October 2024.	The ALICE Collaboration aims to precisely measure heavy-flavour (HF) hadron production in high-energy proton-proton and heavy-ion collisions since it can provide valuable tests of perturbative quantum chromodynamics models and insights into hadronization mechanisms. Measurements of the $\Xi_{c}^{+}$ and $A_{c}^{+}$ insiduction decaying in a proton (c) and	461. LbMCSubmit: Streamlined production and submission of LHCb MC requests (THL Emir Muhammad (University of Warwick (GB))
21/10/224/13:48 Track 5 - Simulation and Talk Parallel (Track 5)	Track 5 - Simulation and _ Poster Poster session	The success and adoption of machine learning (ML) approaches to solving HEP problems has been wi As useful a tool as ML has been to the field, the growing number of applications, larger datasets, and in
380. Fair Universe HiggsML Uncertainty Challenge ▲ Yulei Zhang (Shanghai Jiao Tong University (CN) & APC-Paris (FR))	Z/1)     Maria Teresa Camerlingo (Universita e INFN, Barl (IT))     © 22/10/2024, 15:18	Track 9 - Analysis faciliti Talk Parallel (Track 9)
Detailed event simulation at the LHC is taking a large fraction of computing budget. CMS developed an end-to-end ML based simulation that can speed up the time for production of analysis samples of several orders of magnitude with a limited loss of accuracy. As the CMS experiment is adopting a common analysis level format, the NANOAOD, for a larger	thermal model measurement of the participation	338. Efficiency, Reproducibility, and Portability in HEP Machine Learning Training - ML Facility at Vanderbilt University ▲ Andrew Malone Melo (Vanderbit University (US)) © 24/10/2024. 14.42
© 21/10/2024, 11:30 Plenary Talk Plenary session	Track 5 - Simulation and Talk Parallel (Track 5) Direct photons are unique probes to study and characterize the quark-gluon plasma (QGP) as they leave the collision medium mostly unscathed. Measurements at top Large Hadron Collider (LHC) energies at low pT reveal a very small	Track reconstruction, a.k.a. tracking, is a crucial part of High Energy Physics experiments. Traditional T task, relying on Kalman Filters, scale poorly with detector occupancy. In the context of the upcoming H 110 existence based on Kalman filters and the filter based on the most statement of the second statement of
114. CMS FlashSim: end-to-end simulation with ML Andrea Rizzi	Abhishek Nath (Heidelberg University (DE))     ③ 22/10/2024, 14:24	© 24/10/2024, 13:48 Track 3- Offline Comput. Tak Parallel (Track 3)
Talk Plenary session	293. ML-based classification of photons and neutral mesons for direct photon measurement in ALICE	514. Efficient ML-Assisted Particle Track Reconstruction Designs
Agnieszka Dziurda (Polish Academy of Sciences (PL)), Tomasz Szumlak (AGH University of Krakow (PL))     O 21/10/2024, 09:00	We present an ML-based en-ot-end aigontim for adaptive reconstruction in dimerent F-C detectors, the aigontim takes detector hits from different subdetectors as input and reconstructs higher-level objects. For this, it exploits a memetric graph neural network, trained with object condensation, a graph segmentation technique. We apply this	Although caching-based efforts [1] have been in place in the LHC infrastructure in the US, we show the intelligent prefetching and targeted dataset placement into the underlying caching strategy can improve further. Never prefetching and targeted dataset placement into the ULE LHC and DUNE and an and the ULE of DUNE and DUN
568. Welcome	Track 3 - Offline Comput Talk Parallel (Track 3) We present an ML-based end-to-end algorithm for adaptive reconstruction in different FCC detectors. The algorithm	Track 1 - Data and Meta Talk Parallel (Track 1)
	Dolores Garcia (CERN)     322/10/2024, 14:06	110. ML-based Adaptive Prefetching and Data Placement for US HEP systems ▲ Dr Byrav Ramamurthy (University of Nebraska-Lincoln) © 23/10/2024, 16:51

## meaningful use cases?

L Jiahui Zhuo (Univ. of Valencia and CSIC (ES)), Volodymyr Svintozelskyi (Univ. of Valencia and CSIC (ES)) 323/10/2024, 15:18 Track 2 - Online and real-Poster Poster session

One of the most significant challenges in tracking reconstruction is the reduction of "ghost tracks," which are composed

integrating

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Training

lespread and fast.

programme.

with the Qibo

Trigger

the calorimeter.

#### he exercise searching for "GPU"...

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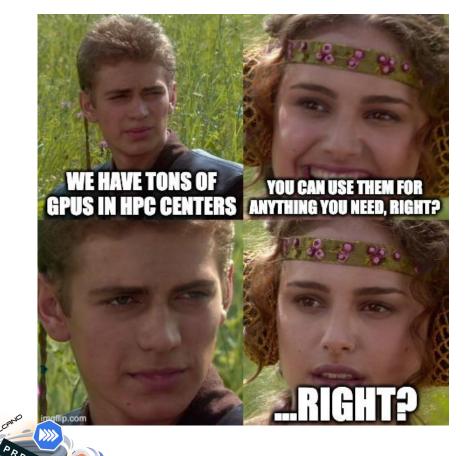
# Where are my resources?

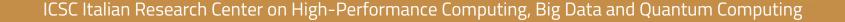


We could/should(?) use HPC supercomputers...

Technically we can already BUT how about making **a painless experience out of it?** 

Can we ignore that most of new data science tools have now a "lingua franca"? Kubernetes API support is arguably a de-facto standard













### Where it all began

A long experience at INFN with site extensions

Two opportunities:

- <u>ICSC</u>
- interTwin EU project
   visit EGI booth to know more!

# <u>Not only HEP</u>, a good solution is an <u>adaptable solution</u>

Everyone should be able to come and plug its own integration logic.

Several heterogeneous science communities: different domains, spanning O(10) to O(1000) people

On a geo-distributed heterogeneous set of computing centers: different resource management (batch, cloud), different size and specialized hardware

We strive to get an efficient interface to match the evolution of analysis frameworks toward cloud-native tools with a seamless resource brokering.



Very similar set of requirements with an even stronger need for cloud native framework integration with HPC centers (kind of a cloud-prohibitive environment)

<u>A Digital Twin Engine should be able to provide a platform capable of distributing</u> <u>DT applications to remote sites, maintaining an interoperable (cloud) interface.</u>







Consortium at a glance



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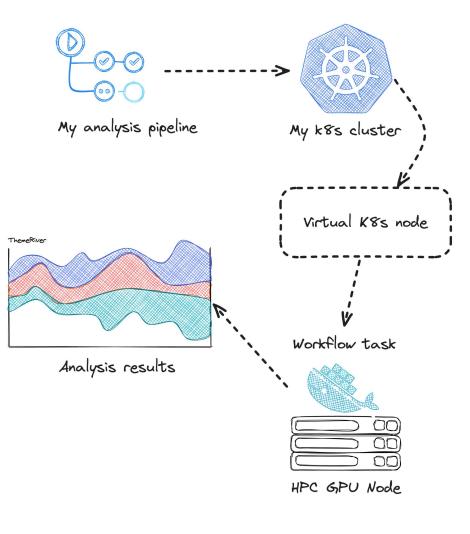


### **Our strategy**

# What if we extend the container orchestration layer (K8s) to support "offloading"\* under the hood? (With little to no knowledge required from the end user perspective)

Exposing the <u>very SAME experience of running a pod</u> on the cloud resources -> never touching the native API layer by K8s

N.B. Using Kubernetes as the workhorse for the "offloading", <u>NOT as the main user interface though</u>



\*"Offloading" refers to the process of delegating the execution of a container to a remote resource instead of a physical node in your cluster.









### **Enabling tech: virtual nodes!**

#### **CNCF** Sandbox project

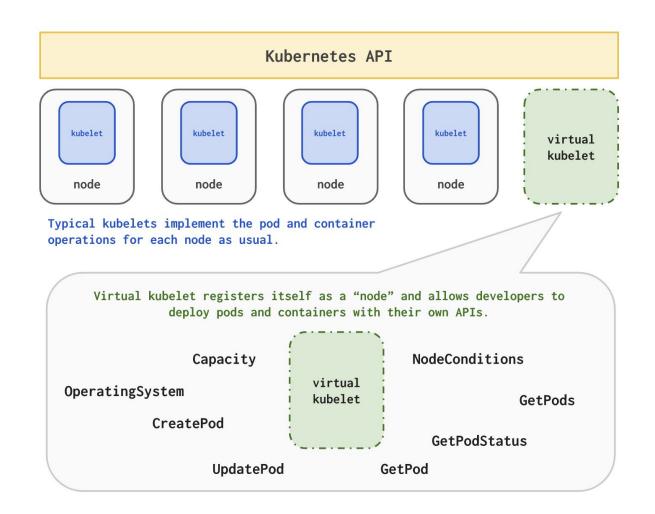


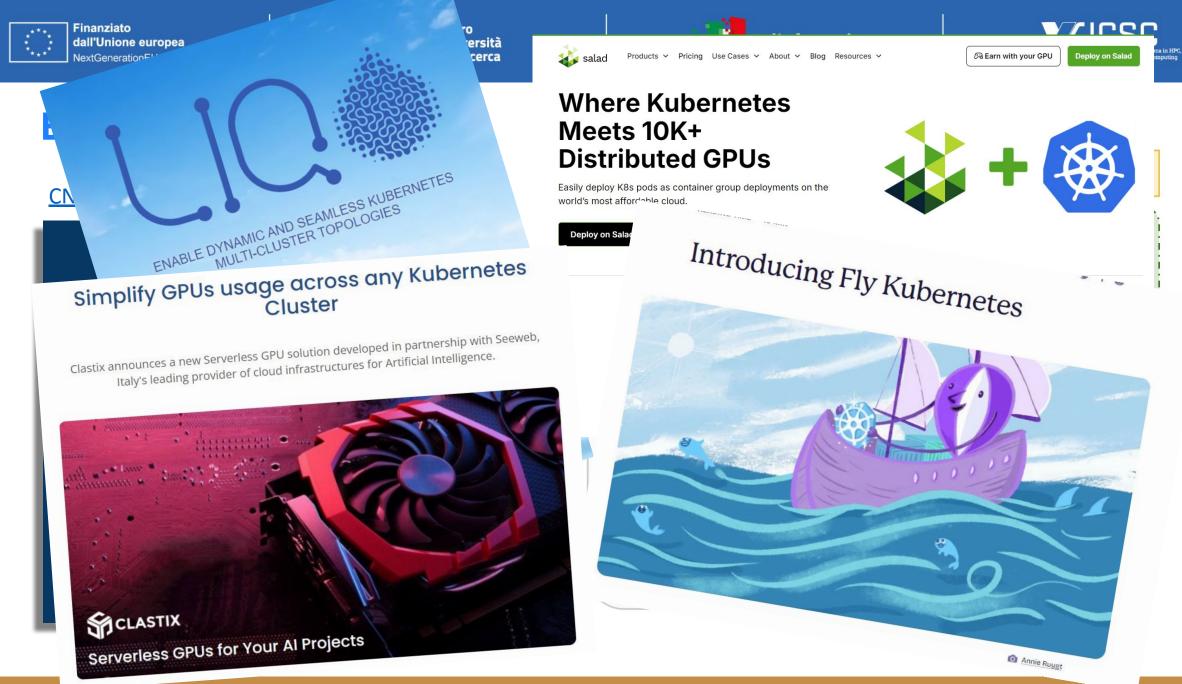
Create a Kubernetes cluster node, but virtual...

Translate a pod request in a custom container execution (e.g. remotely run apptainer/podman)

- 1. KEEP KUBERNETES API
- 2. EXPLOIT EXTERNAL/REMOTE RESOURCES

Quite a good fit 🤒





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### We need a bit more though

With current Virtual Kubelet interface:

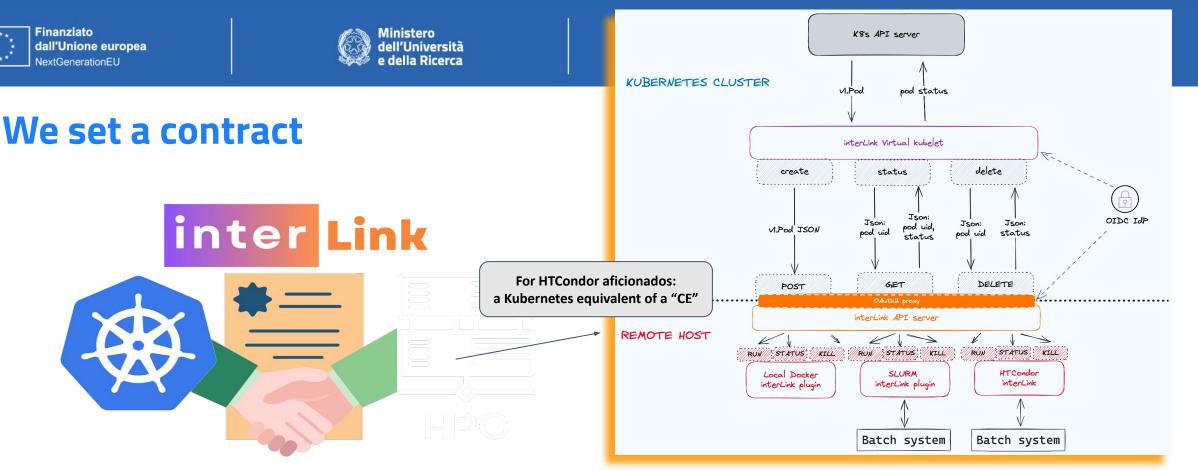
- Deep knowledge of Kubernetes internals is needed
  - When developing a different flavors of virtual kubelet "logic"
- We want providers to be in control
  - abstracting the layer of resource provisioning from the payload brokering itself

Requiring Kubernetes black belt for any provider administrator of ours is a NO GO

The result is the development of <u>a modular interface to create custom</u> <u>Virtual Kubelets:</u>







As a community effort we maintain the complexities

VK-interLink communication, AuthN/Z, VK state machine, caching etc..

### Providers focus on creating value via additional integrations

All of that without a PhD in k8s, leveraging simple SDKs based on interLink OpenAPI spec









### **Our first case studies**



HPC Vega is the first EuroHPC JU supercomputer hosted at the Institute of Information Science in Maribor, in Slovenia.



The Jülich Supercomputing Centre operates one of the most powerful supercomputers in Europe, JUWELS, and JUNIQ the first European infrastructure for quantum computing. UNICORE offers seamless access to the Supercomputers.

**<u>First volunteer HPC provider</u>**, enabling super early prototyping

#### First volunteer for an external plugin based on UNICORE

In both cases the payloads will be submitted to SLURM by the interLink plugin hosted on a VM on the edge of the HPC

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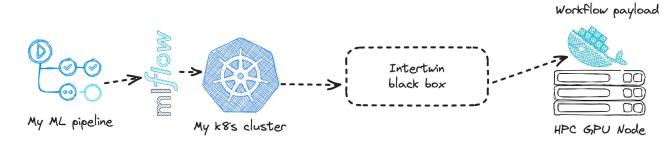
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### Did it work?

First use case driven by **3D GAN algorithm** developed at **CERN** and integrated with **MLFlow** tracking.

**Training task offloaded to GPUs** available at the HPCs as "normal" K8s pods

Pod annotations are a powerful tool to send dedicated requests to the remote container manager!



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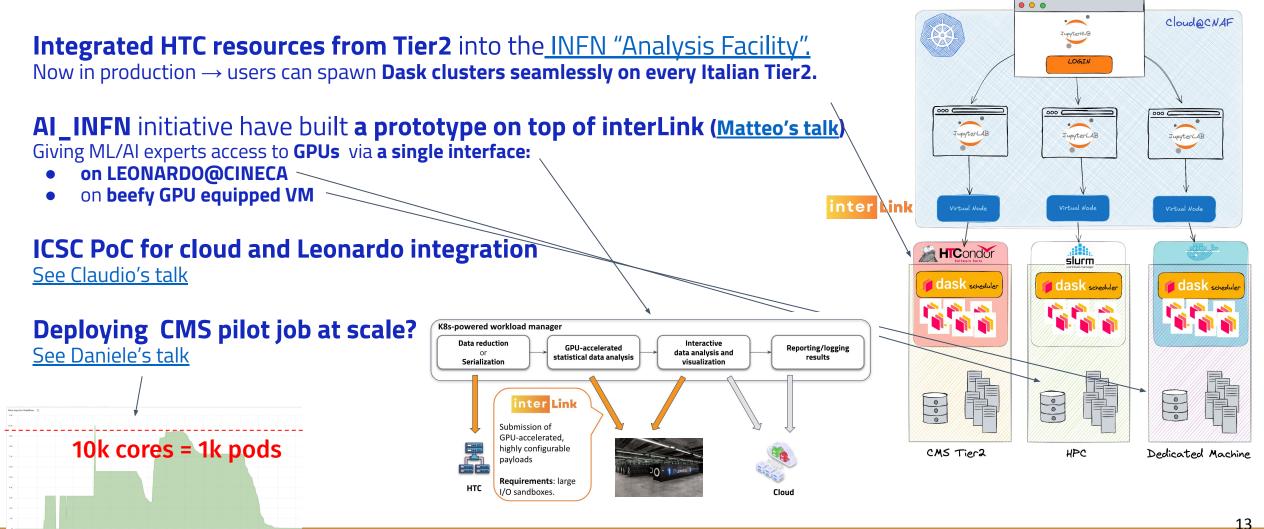








### Not just on SLURM @ INFN and ICSC



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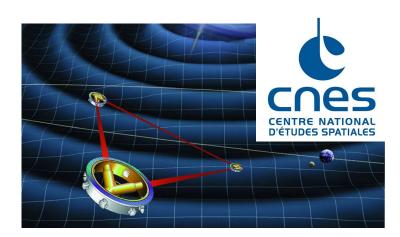


### Toward a "community" software

ArgoWorflows from LISA interferometer are being evaluated to offload argo tasks to HPC centers running SLURM (INFN-CNES interactions)

We are in implementation phase for a offloading prototype with **CERN KubeFlow** based platform.

<u>"Non-scientific"</u> case studies are also joining the evaluation with their own plugins (<u>NuNet</u>)





Wite Initial spec for nunet interlink plugin.

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#### Ss-nowered workload manage HOW IT'S GOING? Not too bad! Data reductio Interactive data analysis and visualization GPU-accelerated Reporting/loggin results ubmission of GPU-accelerated, highly configurabl vloads **GPU accelerated HEP Grid-friendly workflows** Requirements: large simulations and ML training and inference of particle 10k cores = 1k pods triggers tagging Workflow payload <u>Ө-</u>Ө Intertwin black box CMS My K8s cluster HPC GPU N User application CMS inter Link We started here in the Batch analysis Interactive analysis middle, remember? "Coffee-break-duration" long full analyses at LHC Micondor slurm I dask schedule dask schedul dask schedul 666 1.1 î î î 000 000 000 CMS Tier2 Dedicated Machine GPU cloud machine

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### **Summary**

We demonstrated how interLink can extend our resource provisioning model...

for ANY container-based use case!

#### HPC-Slurm is only the beginning

The inclusion of "any" kind of remote resource can work (CaaS, fat nodes, HTC etc..)

#### Providers can offer an effortless K8s style access to resources

without internals deep knowledge required, develop your own provider plugin!

NI I I

<u>Next steps:</u>

#### Streamline communities onboarding

documentation and development SDKs

#### Consolidate through use cases feedback loop

community driven development is key

#### Widening community adoption

applying for contributing to Cloud Native Computing Foundation Sandbox



### https://intertwin-eu.github.io/interLink/





This work is also partially supported by the FAIR "Future Artificial Intelligence Research" NRRP Project











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