



Computing infrastructure for Artificial Intelligence

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Florence workshop on Artificial Intelligence for Phys. Sciences

Garbasso, Arcetri, Firenze

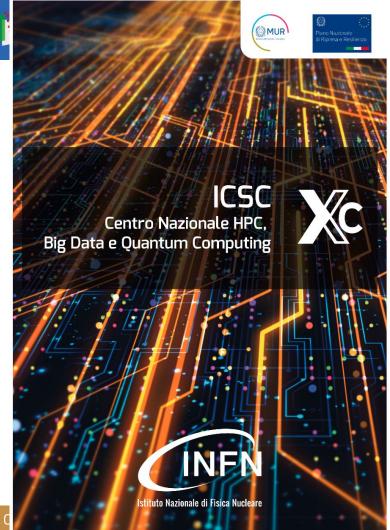






The National Center for HPC, Big Data and Quantum Computing

- Italy has funded, with NRRP (pandemic recovery) funds, 5 large National Centers, for a total of 1.6 Beur over 3 years
- One of them, coordinated by INFN, focuses on modern IT technologies, with the final goal to deploy a long-term (>>3 years) distributed infrastructure available to the reseach and the industrial ecosystems in Italy
- Sept 2022 to Aug 2025





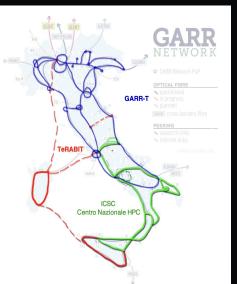






The three pillars of ICSC and Terabit projects

GARR: Tbps-level connectivity between all public data centers



 CINECA: expansion of Leonardo (HPC#6 on top500.org) with Lisa, and deployment of a production level Quantum Computer

GPU Module Lisa

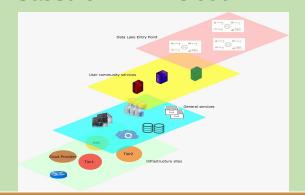
500 compute nodes 4x GPUs / node >100 PF HPL

CPU Module Lisa

1000 compute nodes 2x CPUs + HBM / node >6 PF HPL



• INFN: strengthening of the WLCG infrastructure (1 Tier1 – 9 Tier-2s); acquisition of Cloud resources; implementation of the datalake middleware, based on INFN-Cloud





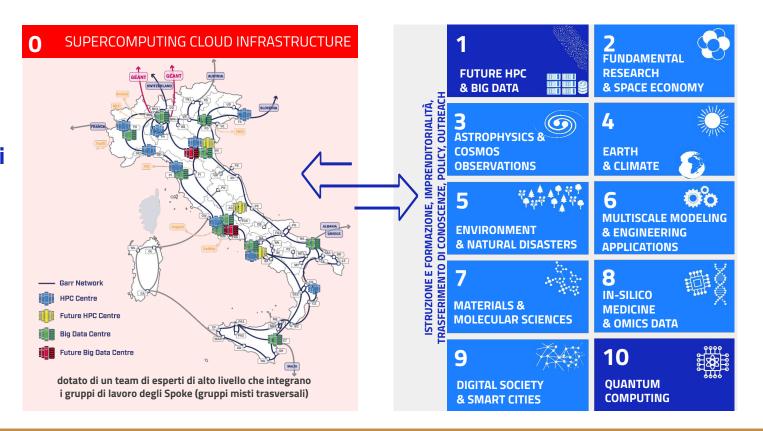




L'ICSC include

10 Spoke tematici
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1 Spoke
infrastruttura









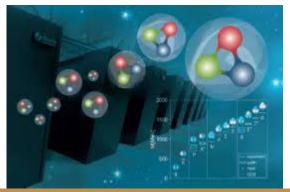


Why our communities need an ICSC

- Since at least 2 decades, research at the fundamental frontier is heavily "computing bound"
 - Latest (and next) generation experiments collect data at the Exabyte and beyond
 - Simulation efforts in theoretical and experimental physics are at the Exascale
 - We have been forced to develop in-house solutions when nothing was available, with a good success. But it is now due time to evolve to a shared infrastructure model
 - The Web, the GRID, ...
- Examples:
 - LHC has has already surpassed the global scale of several Exabytes of Data, and more than 1 Million CPU cores
 - Lattice QCD simulations are, with Meteo, the main driver and benchmark for HPC systems

The Worldwide LHC Computing GRID





Lattice QCD on HPC









From High Performance Computing to Artificial Intelligence

T. Boccali, 1st Hackathon of Machine Learning of Al INFN

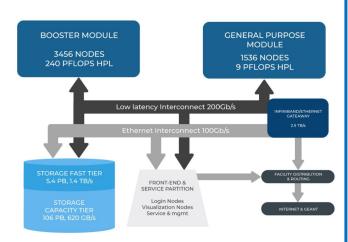
HPC is designed for Simulation

Artificial Intelligence requires data

Leonardo (current machine @ CINECA):

- Not designed for AI, but with standard HPC use cases (lattice QCD, Meteo, computational chemistry, ...)
 - Not al lof them can use GPUs → 2 partitions
- · Leonardo Booster:
 - 3456 nodes, each 3xA100 and 32 CPU cores
- Leonardo GP:
 - · 1536 nodes, each 112 CPU cores
- Fast interconnect between nodes and fast storage

Italy: current and next machine for large (as in LLM?) AI



"It happened" it is also a quite good machine for AI; not automatic!

#4, Fugaku, is a terrible AI machine: no GPUs, low RAM







From Distributed to Heterogeneous computing

Heterogeneity among computing sites: different hardware, operating systems, batch systems, data infrastructure, networking

The name of the game is build the ability to transparently move applications from site to site.

- Uniform AuthN/Z;
- non-locality of data (DataLake), with local caches
- Containerization and distribution of the applications

Lingua franca #1



Kubernetes for containerized applications

Lingua franca #2



Singularity/Apptainer as runtime in user space







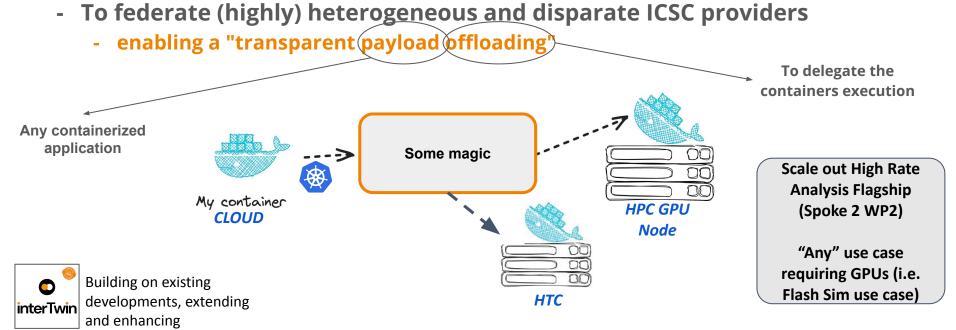


D. Spiga (INFN)
ICSC Annual meeting

InterLink, in general

Transparently extend analysis testbed to run "any application anywhere"

To feed a vote (highly) between a vote and dispersed 1666 avaidance





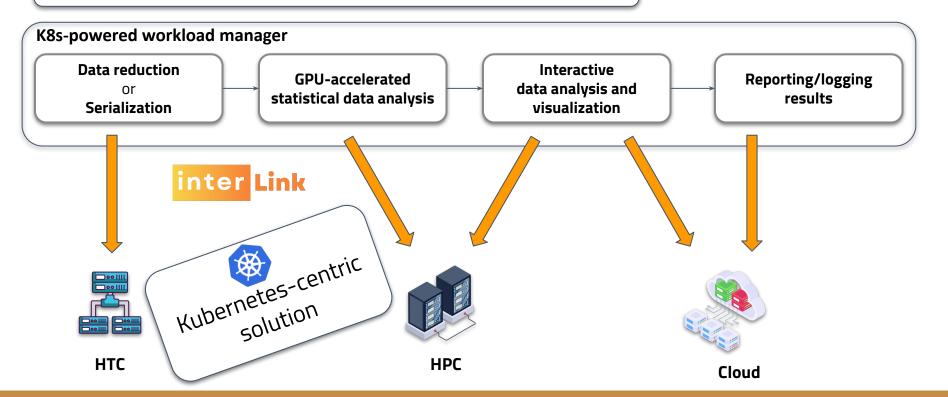






A more "Cloud-native" approach to distribution: *InterLink*

D. Ciangottini, ICSC, CHEP 2024 L. Anderlini, ICSC, SIF 2024











The AI_INFN Platform as a bridge from the Cloud to ICSC









Federated bare-metal resources

Computing resources available to Al_INFN are located at Bologna Technopole within the new CNAF Data Center facility, and managed through a **virtualization layer** (OpenStack of Cloud@CNAF) in **INFN Cloud**:

- 4× servers:
 - 1x 64 CPU cores with 750 GB RAM
 - 3x 128 CPU cores with 1024 GB RAM
- Total local storage: 60 TB of NVMe disk
- GPU cards:
 - 8x NVIDIA Tesla T4
 - 5× NVIDIA RTX 5000
 - 1x NVIDIA A30
 - 4x NVIDIA A100, potentially served as 4x7 MIG slices
- FPGA boards:
 - 2× AMD Xilinx Alveo V70
- 10 GbE connection to CNAF resources











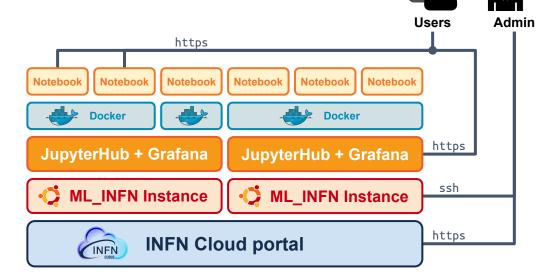
INDIGO IAM

The ML_INFN platform architecture

The ML_INFN outcome:

"

Sharing precious GPUs through the Cloud is feasible and effective!











The AI_INFN platform architecture

indigo iam

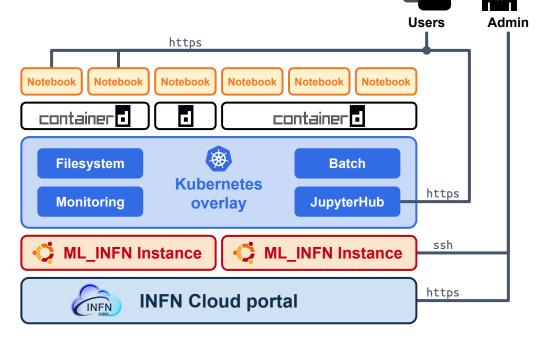
The **ML_INFN** outcome:

"

Sharing precious GPUs through the Cloud is feasible and effective!

AI_INFN improves the sharing capabilities:

- addition of an abstract and elastic overlay powered by Kubernetes
 - login via AAI → INDIGO IAM
 - distributed filesystem
 - managed environments for ML
 - monitoring & accounting
- data decoupled from computing resources with a filesystem shared across the VMs
- adding and removing VMs enables manual horizontal scaling

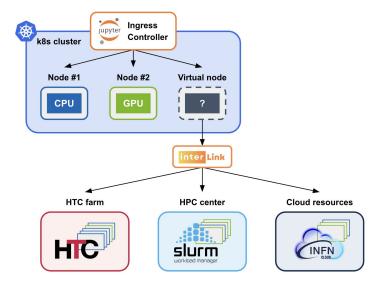








Enabling offloading using interLink as Virtual Kubelet provider



extension of the Al_INFN platform through the VK mechanism

Once AI models are developed, researchers often seek to scale them **beyond development-dedicated resources**

The Al_INFN platform is exploring a solution to transparently extend the resource pool accessible to Kueue using the <u>Virtual Kubelet</u> (VK) mechanism:

- VKs provide k8s cluster with "Virtual Computing Nodes" that have no networking towards the API server or other services
- VKs are ideal for batch processing, where the connection between the cluster and the working node is only needed at job submission and retrieval

The <u>interLink</u> protocol offers a batch-system native backend for Virtual Kubelets (e.g., SLURM, HTCondor, or other Kueue instances)

Developed as part of the Flagship Activities of ICSC Spoke 2







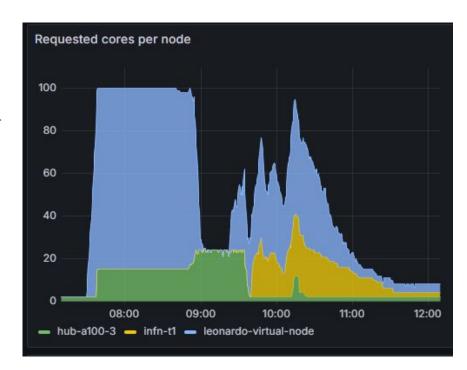


First workflow with INFN Tier-1, CINECA Leonardo & AI_INFN resources

On October 25th, we run a first workflow combining CPU resources from CNAF Tier-1, Leonardo and a local node.

Test limited to CPU-only payloads.

Offloading to Leonardo booster (with GPU payloads) coming soon.











Some(!!!) applications of machine learning studied within ICSC

<u>Fast Reconstruction of the ATLAS experiment</u>

Flash Simulation of the CMS experiment

Lamarr, Flash Simulation of the LHCb experiment

Fast Simulation of cascades from cosmic rays in atmosphere

Physics Informed Machine Learning for AI-based simulations

FADeR, Real-time track reconstruction of the LHC experiment

Machine Learning for Predictive maintenance

SAIFIN, Satellite data and Artificial Intelligence for FINtech

Event reconstruction in Super-Kamiokande

Fast detector simulation for 3D Silicon Pixel detectors

<u>Super-resolution surrogate model for accelerated Geant4 simulations</u>

Satellite data management for advanced environmental applications

Particle ID reconstruction for Future Collider Experiments

Searches for low-frequency Gravitational Waves

Model sources of gravitational waves

Processing of 3D Medical Images (e.g. CT Scans)











Advanced Hackathon of Machine Learning (AI_INFN)

Bari 2022, Pisa 2023, Padova 2024

School of Open Science in the Cloud (SOSC)

Perugia 2022, Perugia 2023, Bologna 2024







Seminar on Software for Nuclear, Subnuclear and Applied Physics

Alghero 2022, Alghero 2023, Alghero 2024







Conclusions?

ICSC/Terabit are entering its third (last NRRP-funded) year of scientific activity.

It fostered a huge effort on digital technologies, and in particular on Machine Learning, enabling access to an unprecedented amount of computing resources.

The call for sustainability of the action is now pressing.

The recent selection of Bologna for one of the AI factories confirms that Machine Learning will remain central.