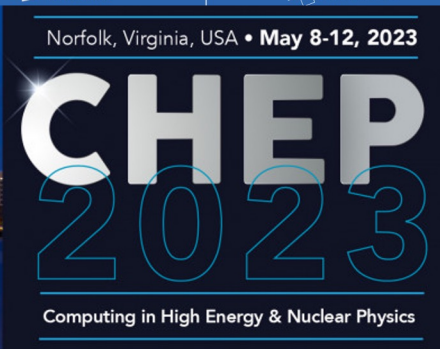




Luca Giommi
Istituto Nazionale di Fisica Nucleare
CNAF

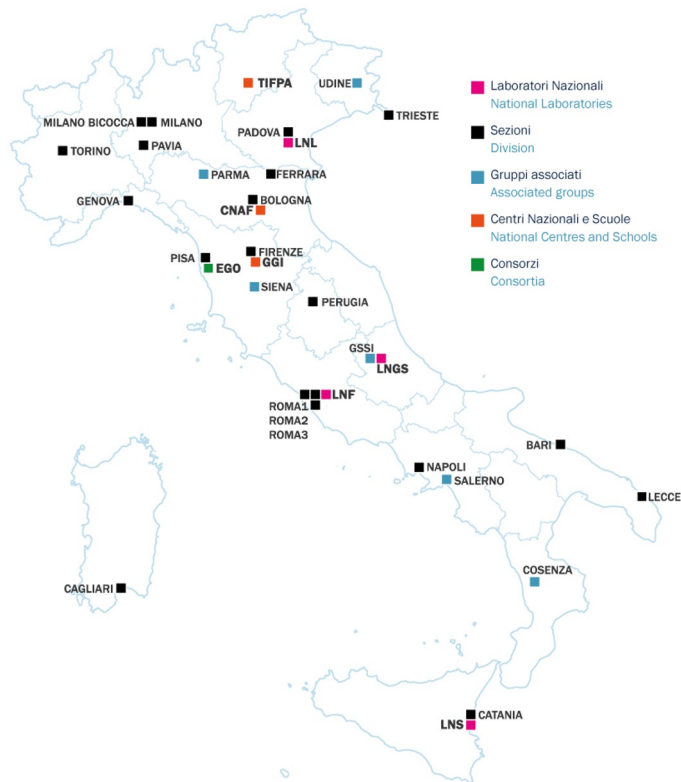


The ML_INFN Initiative





INFN Research and structures



216 activities distributed in 33 structures (labs, groups and divisions)



Particle Physics

17 experiments



Astroparticle Physics

45 experiments



Nuclear Physics

23 experiments



Theoretical Physics

35 initiatives



Technological Research

96 experiments

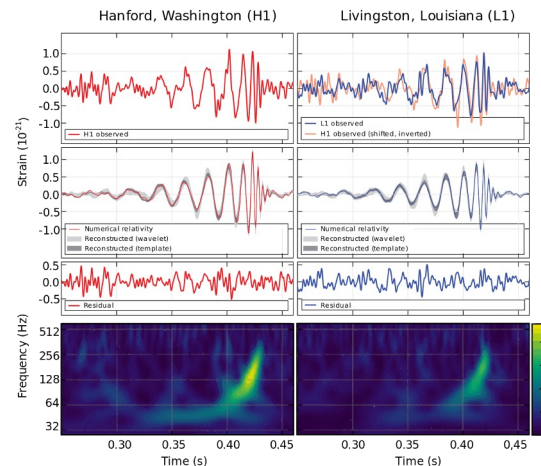


Machine Learning applications in INFN

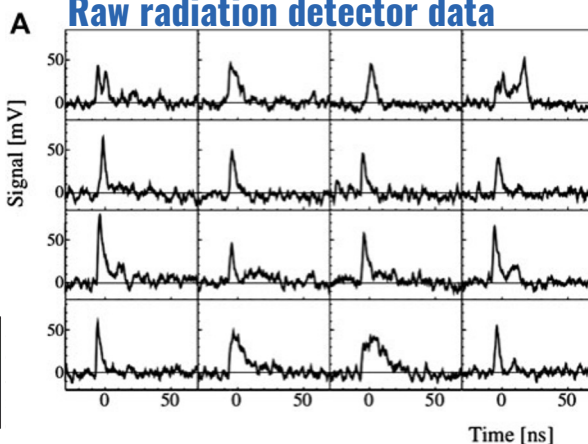
Most of the experiments and initiatives produce, analyse or process digital data.

Enthusiasm on the modern data processing technologies!

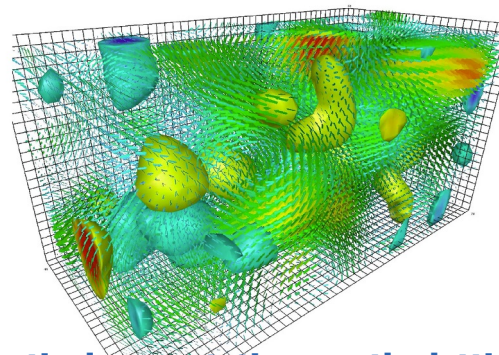
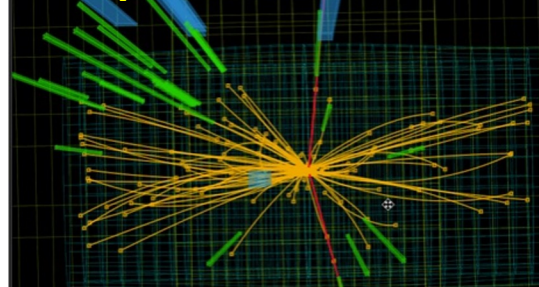
Gravitational wave detection



Raw radiation detector data



LHC experiment data & simulation



Theoretical computations on the lattice

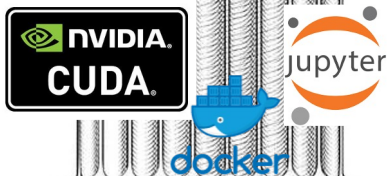




ML_INFN: The structure of the project

Applications of Machine Learning
HEP, MedPhys, GW detection, Theory...

Infrastructure



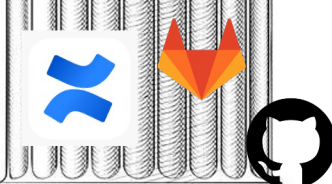
WP 1

Stewardship



WP 2

Knowledge Base



WP 3



Virtualization and orchestration layer
developed and maintained by INFN Cloud



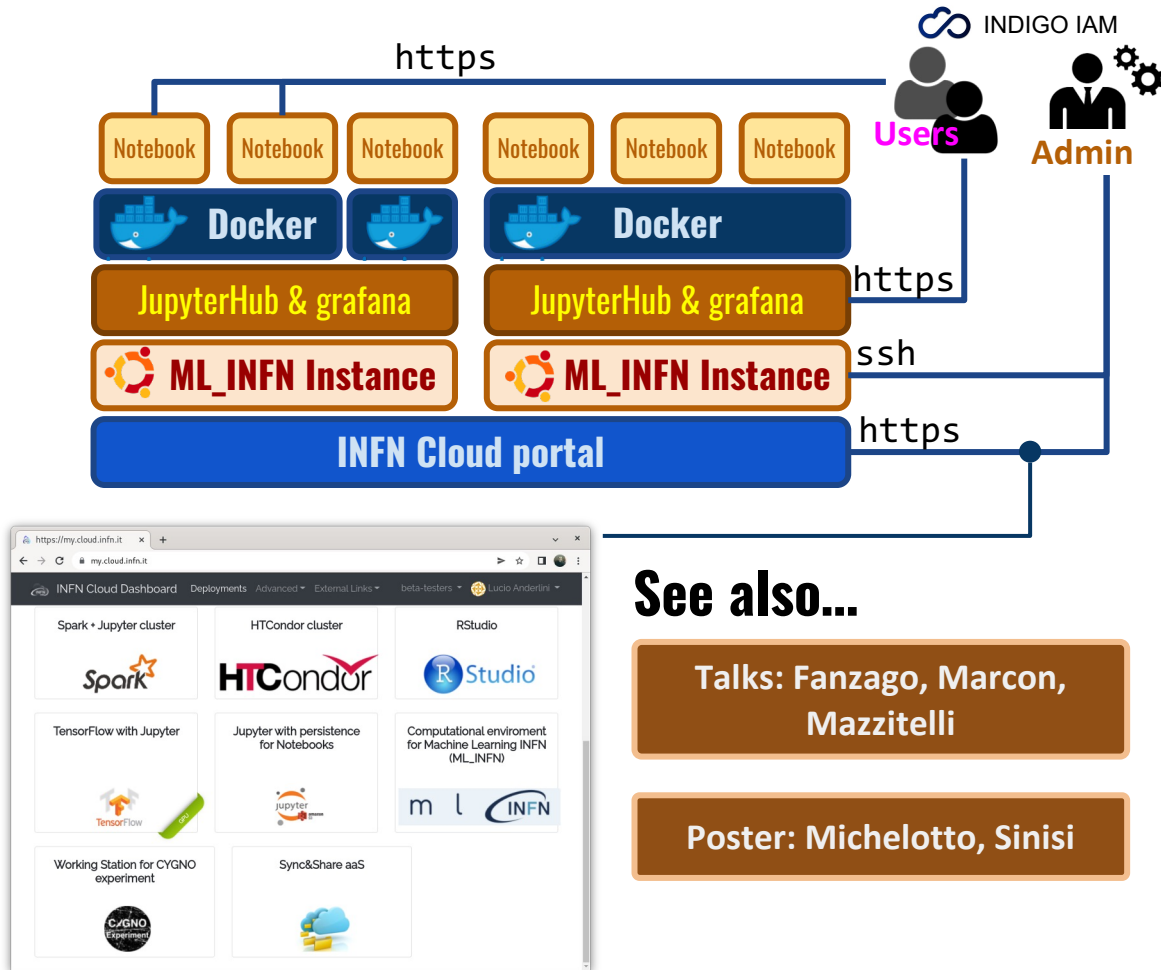


WP1. The infrastructure



INFN Cloud

ML_INFN is built on top of **INFN Cloud**: a data-lake centric, heterogeneous federated Cloud infrastructure spanning over multiple sites across Italy, providing an extensible portfolio of solutions tailored to **multidisciplinary scientific communities**.

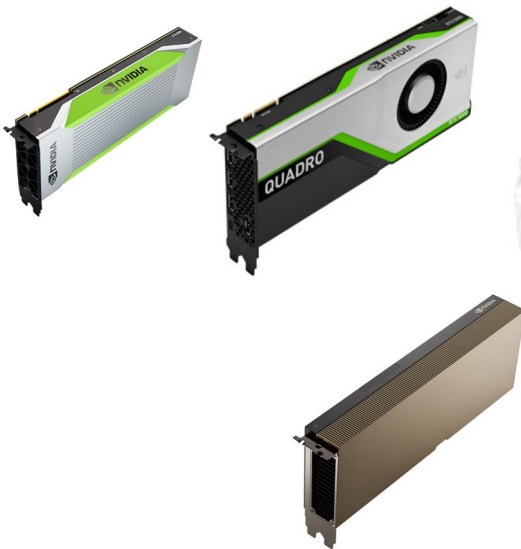




Federated bare-metal resources

1× SuperMicro + 1× E4 servers:

- 1.8 TB RAM
- 64-128 CPU cores
- 36 TB local storage (NVMe)
- 8× **Tesla T4** GPUs
- 5× **RTX 5000** GPUs
- 1× **A30** GPU
- 1× **A100** GPU, served as 7 independent MIG slices
- 10 GbE connection to CNAF resources



Storage solutions

CERN experiments data, contained in INFN Tier-1 storage, are remotely accessed via NFS

Hypervisors integrate Ceph to manage persistent virtual volumes accessed from the VM via POSIX

Federated to CNAF OpenStack and INFN Cloud



WP2. Stewardship



Machine Learning hackathons: *Base and Advanced level*

To foster the adoption of machine learning tools and techniques in INFN community, we organize events to discuss ML algorithm with the time to look at (and hack) the code.

Starting-level Hackathons

Jun 2021, Dec 2021, Jun 2023

- **online events** with no fee
- up to **60 participants**
- **1 tutor per 5 participants**
- INFN Cloud **CPUs** with shared filesystem

Advanced Hackathons

Nov 2022

- **in-person events**
- up to **30 participants**
- **(almost) 1 tutor per participant**
- INFN Cloud **GPUs** with shared filesystem



Base hackathon: *Lecture Program*

Day 1

Lectures

Theoretical introduction to ML

Lectures

Cloud and Cloud Resources

Day 2

Hands-on

Neural Networks

Seminars

**Deep Neural Network
Applications to INFN research**

Day 3

Hackathon

**Exercises
with tutors *continuous support***

Lunch break

Hands-on

Numpy, Pandas and Keras

Hands-on

**Exercises
with tutors *on demand***

Closure

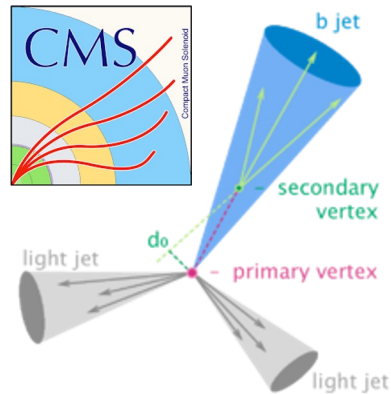
Reports from the students



Hackathon use cases: 10 groups, one tutor per group

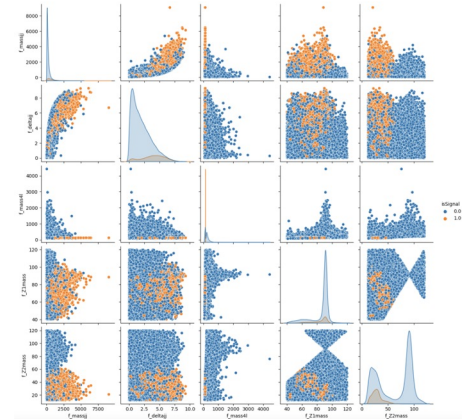
Jet b-tagging at CMS

Recurrent Neural Networks with LSTM



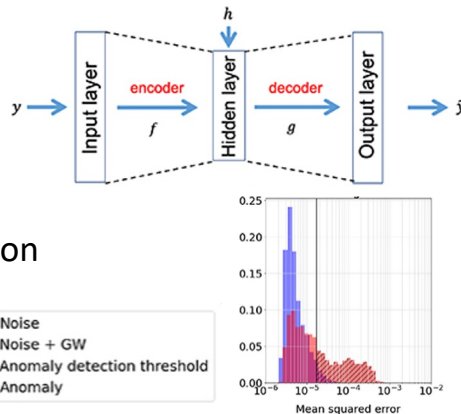
Higgs searches at CMS

Deep Neural Networks and Advanced Keras



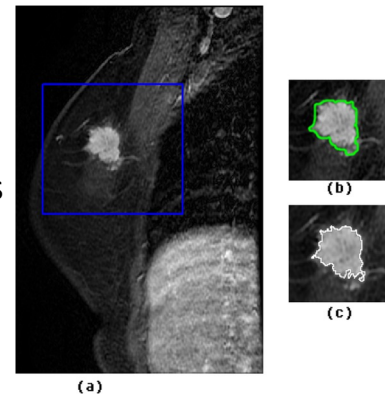
Gravitational Waves with Virgo

Autoencoders, anomaly detection and compression



Segmentation of CT scans

Convolutional Neural Networks
Handling 2D and 3D datasets





Advanced hackathon: *Lecture Program*

Day 1

Day 2

Day 3

Day 4

Lectures

Advanced Models
(U-Nets, GANs, NFs, ...)

Lectures

Advanced Models
(Transformers, GNNs)

Lectures

**Ongoing R&Ds in
Machine Learning**
(FPGAs, HPO, ...)

Hands-on

Beyond Keras
(Coding lower-level ops)

Lectures

Explainability

Hands-on

BondMachine
Compiling NN in VHDL

Lunch break

Lectures

**Cloud Infrastructure and
High Performance
Computing**

Hackathon

**Exercises
with tutors *continuous
support***

Hackathon

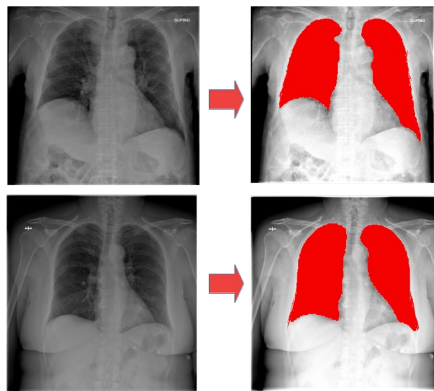
**Exercises
with tutors *continuous
support***



Advanced hackathon use cases

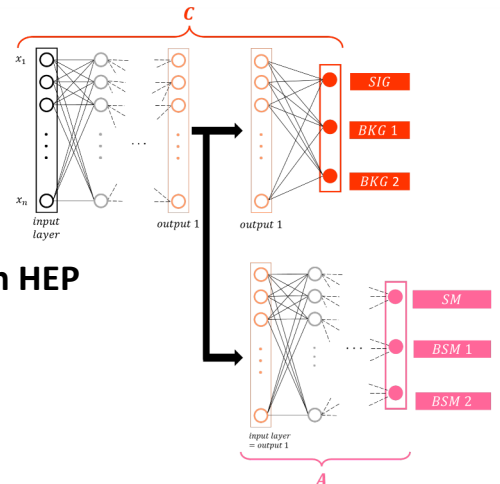
Lung Segmentation with U-Nets

U-Nets, custom loss, custom data loaders



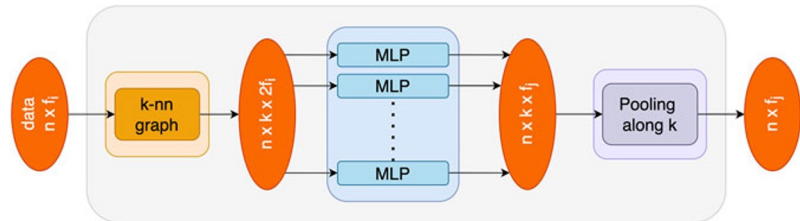
Domain Adaptation in HEP

Adversarial Training, Gradient Tape



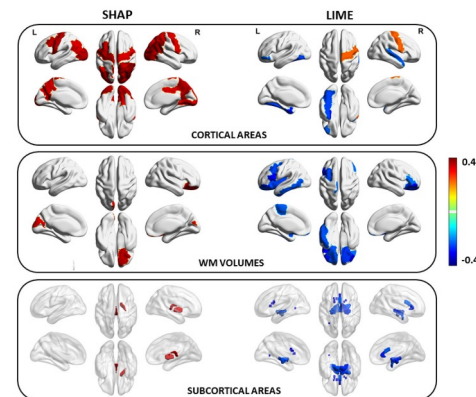
GNN and Transformers in HEP

Application to Jet Tagging



Explainability

Shapley and GradCAM





WP3. The Knowledge Base



Confluence Knowledge Base

Atlassian Confluence was used to build a **Knowledge Base** reporting several machine-learning use cases, including those discussed at the hackathon.

Each entry includes:

- Runnable **example** as a jupyter notebook or a git repository
- **Contact information** of one or more experts

Machine Learning Knowledge Base

This section of the ML-INF N Confluence Space contains the Knowledge Base of fully implemented use cases. This has been created in order to provide new users getting close to Machine learning with concrete examples, with step by step guides for reproducibility.

The division into categories is multidimensional

- Dimension 1: per Machine Learning technology (CNN, Auto encoders, LSTM, GraphNet, ...)
- Dimension 2: per scientific field (High Energy Physics, Gravitational Waves, Medical Physics, ...)
- Dimension 3: per type of used tool

and is implemented via Confluence labels.

Table of Use cases

Name and Link	ML Technologies	Scientific Field	ML Tools	Comments
Tagging in CMS (templated version)	CNN, LSTM	High Energy Physics	Keras + Tensorflow	Realistic application
LHCb Masterclass, with Keras	DE, MLP	High Energy Physics	ROOT + Keras + TF	Introductory tutorial
MNIST in a C header	MLP		Keras	Free-styling tutorial
LUMIN: Lumin Unifies Many Improvements for Networks	CNN, RNN, GNN	High Energy Physics	PyTorch	Package use examples
INFERN: Inference-Aware Neural Optimisation	NN	High Energy Physics	Keras + Tensorflow	Technique application example
An introduction to classification with CMS data	Fisher, BDT, MLP	High Energy Physics	Scikit-learn, TF2	Tutorials for Master

Seminars organized in 2023

- 1) Improving parametric neural networks for high-energy physics (and beyond)
- 2) Cell counting with cell-ResUnet
- 3) A Neural-Network-defined Gaussian Mixture Model for particle identification applied to the LHCb fixed-target programme
- 4) Deep-learning emulators and hierarchical Bayesian inference: application to gravitational-wave astronomy
- 5) New Physics Learning Machine (NPLM): a tool for statistical anomalies detection in presence of systematic uncertainties
- 6) Machine Learning as a Service for High Energy Physics (MLaaS4HEP): a service for ML-based data analyses
- 7) Ante-hoc explainability methods: the ProtoPNet architecture and its application on DBT images



Publications

Measuring Analytic Gradients of General Quantum Evolution with the Stochastic Parameter Shift Rule WOS:000613656700001 10.22331/q-2021-01-25-386	QUANTUM-AUSTRIA	2021
The novel Mechanical Ventilator Milano for the COVID-19 pandemic WOS:000632649400001 10.1063/5.0044445	PHYS FLUIDS	2021
How to enhance quantum generative adversarial learning of noisy information WOS:000655340700001 10.1088/1367-2630/abf798	NEW J PHYS	2021
Deep learning method for TomoTherapy Hi-Art: prediction three-dimensional dose distribution WOS:000709667204219	RADIOTHER ONCOL	2021
Model compression and simplification pipelines for fast deep neural network inference in FPGAs in HEP WOS:000714374500003 10.1140/epjc/s10052-021-09770-w	EUR PHYS J C	2021
Generalization in Quantum Machine Learning: A Quantum Information Standpoint WOS:000718152600001 10.1103/PRXQuantum.2.040321	PRX QUANTUM	2021
Model compression and simplification pipelines for fast deep neural network inference in FPGAs in HEP (vol 81, 969, 2021) WOS:000726095200001 10.1140/epjc/s10052-021-09875-2	EUR PHYS J C	2021
Calorimetric Measurement of Multi-TeV Muons via Deep Regression WOS:000749246700002 10.1140/epjc/s10052-022-09993-5	EUR PHYS J C	2022
A Neural-Network-defined Gaussian Mixture Model for particle identification applied to the LHCb fixed-target programme WOS:000770368300012 10.1088/1748-0221/17/02/P02018	J INSTRUM	2022
Tau Lepton Identification With Graph Neural Networks at Future Electron-Positron Colliders WOS:000836356800001 10.3389/fphy.2022.909205	FRONT PHYS-LAUSANNE	2022
Applications of artificial intelligence in stereotactic body radiation therapy WOS:000837374300001 10.1088/1361-6560/ac7e18	PHYS MED BIOL	2022
Model independent measurements of standard model cross sections with domain adaptation WOS:000869847800002 10.1140/epjc/s10052-022-10871-3	EUR PHYS J C	2022
Robust quantum classifiers via NISQ adversarial learning WOS:000890324500009 10.1038/s43588-022-00359-1	NAT COMPUT SCI	2022
Toward artificial-intelligence assisted design of experiments WOS:000908434600001 10.1016/j.nima.2022.167873	NUCL INSTRUM METH A	2023
Hyperparameter Optimisation of Artificial Intelligence for Digital REStoration of Cultural Heritages (AIRES-CH) Models WOS:000916462800007 10.1007/978-3-031-10536-4_7	LECT NOTES COMPUT SC	2022



Summary

The ML_INFN initiative has been providing many INFN experiments with the hardware and the knowledge base to assess the potential **benefit of machine learning to their research** for three years.

The **ML_INFN** project relies on **INFN Cloud** solutions and it federates resources optimized for ML performance in interactive and batch-like usage patterns (high-end professional GPUs, NVMe disks, many-core high-RAM systems)

A series of national training events (**machine learning hackathons**) and a collection of tutorials and real applications within the INFN community (**knowledge base**) contribute to building **a network of experienced and enthusiast machine learning practitioners**, lowering the skill gap to benefit from machine learning developments.



Thanks for the attention

Questions?



Backup



The numbers of ML_INFN

12 INFN **structures** involved in the developments, training activities and hackathons

79 **researchers** devoting a fraction of their time to promote ML techniques for research

14 professional **GPUs** made available and accessible through the INFN Cloud Interface

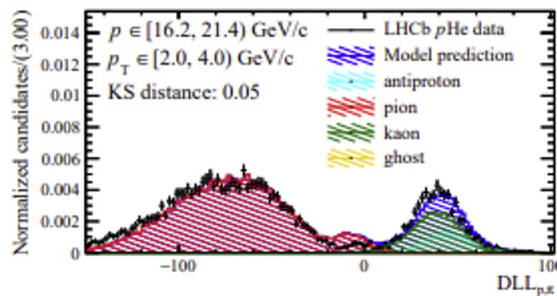
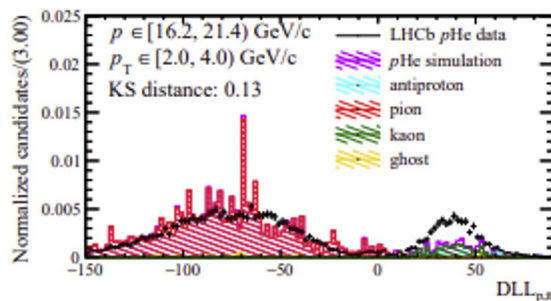
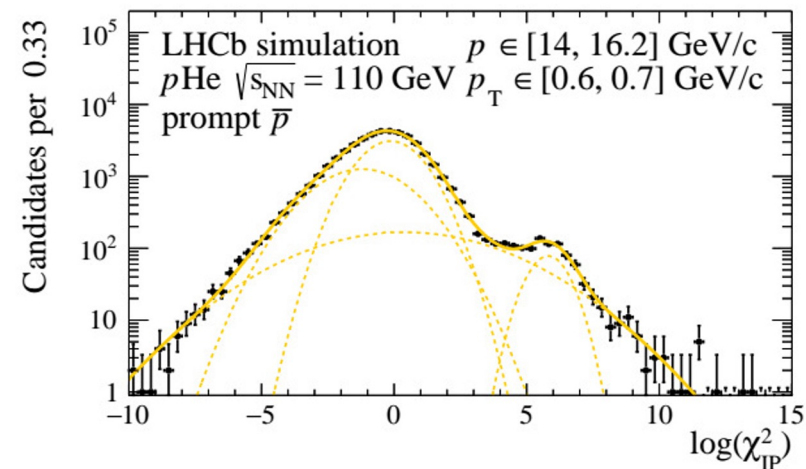
143 **participants** to the **hackathons**, ranging from students to permanent staff members



Stories of success [1]: building template models for LHCb

ML_INFN infrastructure was used to develop a model for the Particle Identification response of the LHCb detector as a Gaussian-Mixture model.

With Gaussian parameters inferred with a Deep Neural Network.



S. Mariani et al,
“A Neural-Network-defined Gaussian Mixture Model for particle identification applied to the LHCb fixed-target programme”, JINST 17 (2022) P02018

Traditional method based on reweighted MC

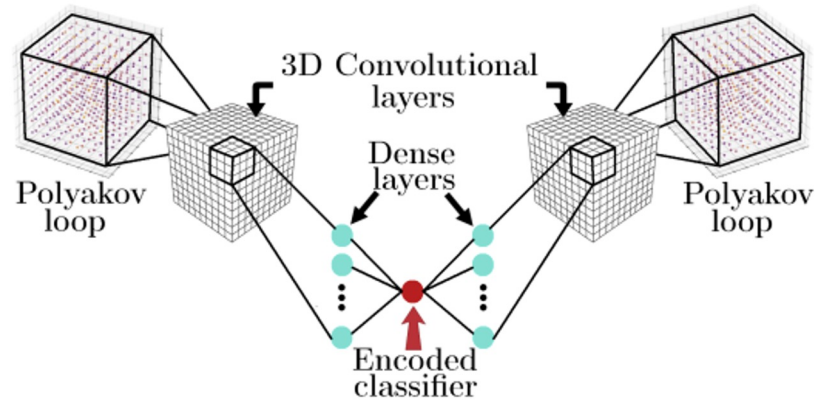
Deep Learning model



Public repository
PID4SMOG

Stories of success [2]: studying LQCD with CNNs

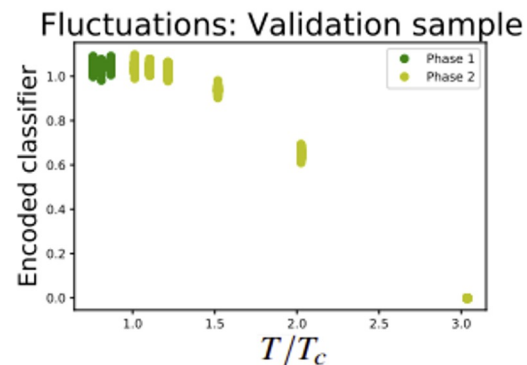
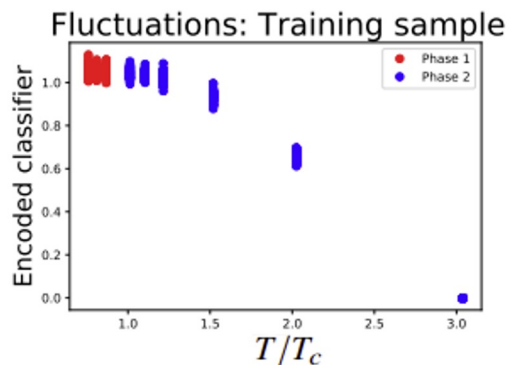
A Deep Neural Network is trained in a semi-supervised manner to define an effective order-parameter for Gauge theory where a real order-parameter is not defined.



The study was made possible thanks to the GPUs provided by the ML_INFN initiative.

A. Palermo, M.P. Lombardo *et al.*

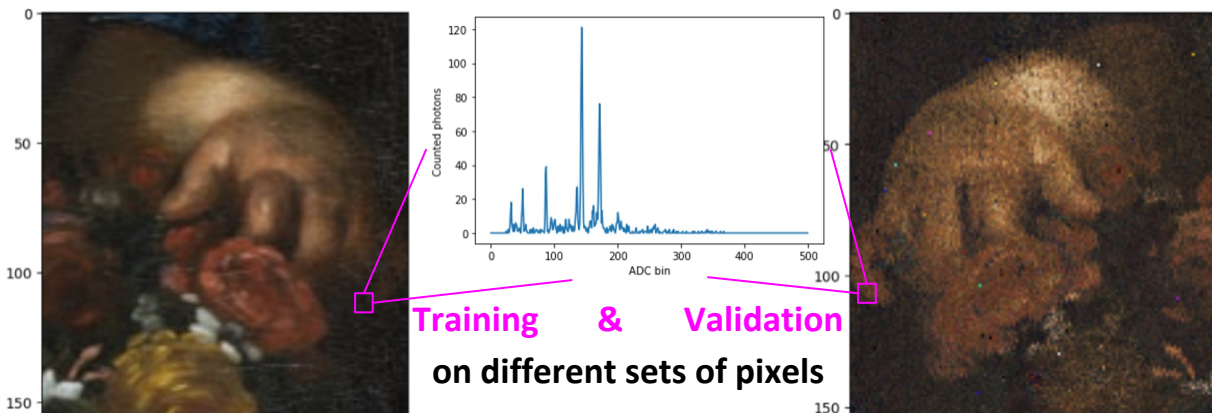
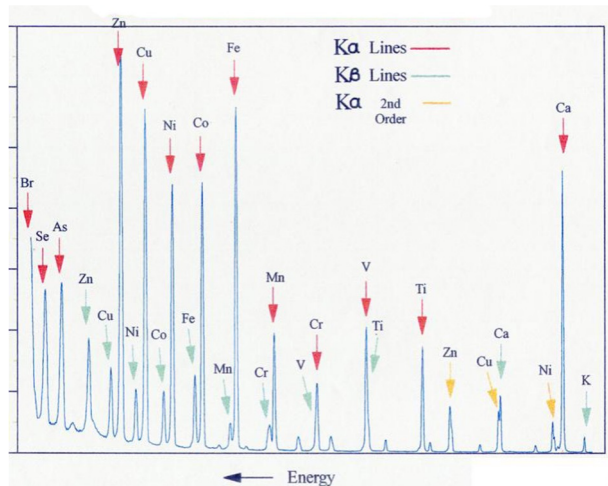
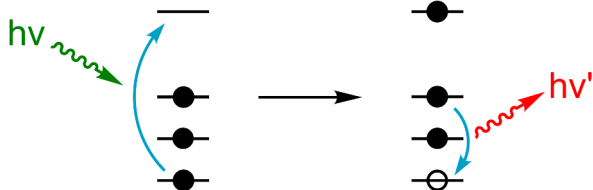
“Machine learning approaches to the QCD transition”, Proceeding of LATTICE21





Stories of success [3]: X-rays to visible colors for CH

X-ray fluorescence spectroscopy widely used for Heritage Conservation and non-invasive probe of pictorial artworks.



Deep Neural Network models are trained to reconstruct the image from the XRF scan of the pixels.

