

Knowledge transfer@ CERN

Nick Ziogas

Knowledge Transfer Group CERN

Overview

- Innovation drivers context briefly
- Our mission
- Tech & competence samples
- Example projects
- How we work together
- Conclusions
- Objectives of the day

FACTS

- The LHC collides protons at unprecedented energy, equivalent to 13,000 times their mass
- 40 Million collisions/sec, one every 25 ns. About 40 collisions per event.
- Thousands of particles emerge from each collision
- 1 MB of data recorded by the detectors at each collision. Too much to be stored
- Only 5% of those are stored after filtering. About 80 Pb of derived data per run.

The NEXT Challenge

- Upgrade of the LHC the CERN flagship accelerator which led to the discovery of the Higgs boson.
- Increase of instantaneous luminosities by a factor of five larger than the LHC nominal value
- Enlarge experiment data sample by one order of magnitude compared with the LHC baseline programme.
- Operational around (2026)-2027

The LHC Big Data Challenge – HL LHC

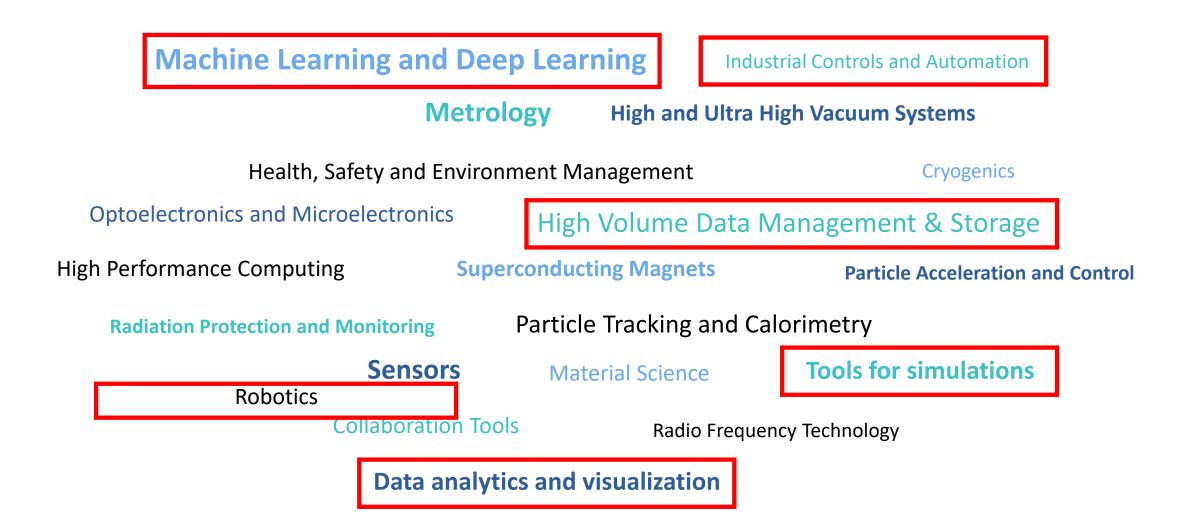
High Luminosity LHC

- 200 collisions per event vs 40 today. Need to disentangle 200 collisions happening at once.
- Event complexity grows non linearly
- A HL-LHC run would need to store about 900 Pb of derived data. A data deluge!
- Even taking into account HW progress (storage & processing), we are off by a factor of 10, projecting to 2027

 Physics <u>always</u> drives CERN Innovation but we need
 Accelerators, Detectors & Computing to underpin the Physics exploration
 R&D is happening now, in many domains

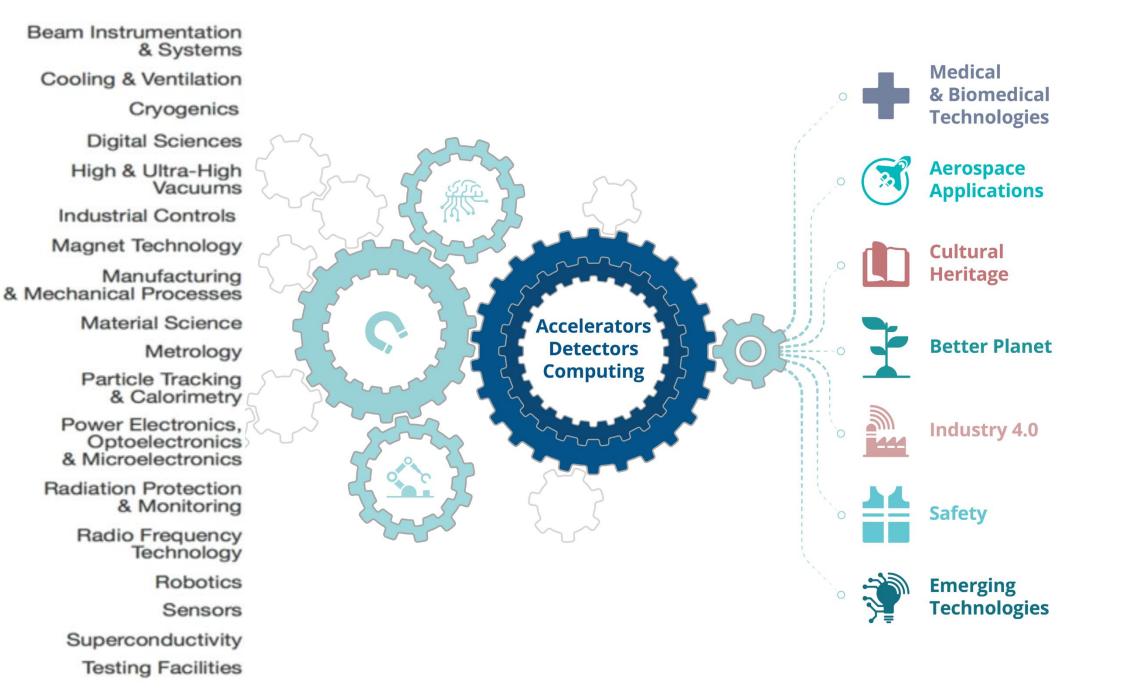
- ✓ In Computing
 - Data Analytics
 - Machine/Deep Learning at detector output & many other areas
 - Quantum Computing

✓ Some of the promising areas to address our challenge



The mission of CERN's Knowledge Transfer Group is to maximise the impact of CERN technology and know-how in society, in particular through industry in the member states.

> *Key words: Dissemination and Impact!*



Key technology: Ultra-fast on-edge neural network inference

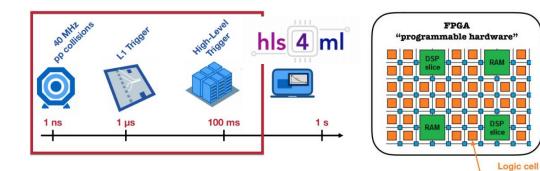
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Know-how in neural network pruning and neuromorphic chips (e.g. FPGA) for ultra-low latency, on-edge inference



CERN needs ultra fast deep learning inference (execution in ~1 microsec) for fast classification of particle collision data, requiring compact code for edge-computing on programmable chips (FPGAs with logic cells)

For this, CERN contributes to an open-source **package (hls4ml) to automatically translate pre-trained neural networks** (as specified by NN architecture, weights, biases) into high-level synthesis code for FPGA architecture, drastically **accelerating prototyping**, **reducing time to results**

- Pruning, quantization (binary, tertiary), compression and parallelization of models by ML experts
- hls4ml integrates with DL libraries: reads as input models trained with Keras/TF, PyTorch, scikit-learn, planned xgboost and outputs hls code; uses Xilinx HLS software (accessible to non-expert, engineers)
- Inference time: <10 microsec
- Comes with implementation of common network components (layers, activation functions, binary NN, ...); example classification network: 16 inputs, 3 layers with 64/ 32/ 32 nodes (ReLU) and 5 outputs (Softmax)





- A car manufacturer teamed up with CERN to develop ML for fast classification of image (computer vision) and lidar sensor data from self-driving cars using FPGAs
- A **developer of wireless or provident tech** jointly with CERN developed quantized ML to gain specified and a energy with minimal impact on accuracy
- Addition 155 cases: ultra-fast, on-edge, energy-efficient ML incl. fast triggering and classification of events, search engines and chatbets (FPGAs used in BING and SIRI), privacy-compliant inference (object classification without storing camera data)
- CERE, can help organizations design optimized neural networks for neurophic chips
 Adapt neural network code for edge inference on FRGA architecture with hls4ml
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 Adapt neural network code for edge inference on FRGA architecture with hls4ml
- Design neural networks for fast and efficient inference through pruning, quantization, compression, parallelization
 Entry of ficient, reduced heat, reduced latency, ultrast statements
- Understand Jatest developments, define strategy for adoption and R&D
- → Advised opinion on fast ML and the second of the entry by CERN experts

ZENSEACT (Volvo Cars Company) teams up with CERN on fast machine learning using FPGAs.

Collaborative R&D

- General issue
- Jointly find solution

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• Jointly develop solution

Collaborative R&D

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CEVA and CERN joined R&D on neural network weight and activation compression algorithms aiming make them run more efficiently. Wireless comms & computer vision applications

Key competence: Big data classification and anomaly detection

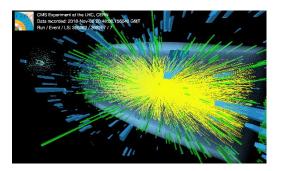
CERN researches and operates and highly sensitive ML models for the detection of weak signals in very large datasets

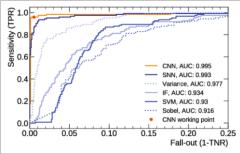
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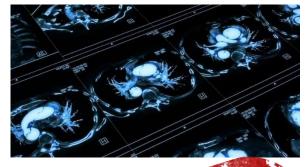


CERN generates and stores large data volumes (1,000 observations/ sec. corresponding to 1 GB/ sec. and 10,000 TB/ year). A single physics analysis typically involves millions of observation examples to reach science-grade results requiring strong classification and anomaly detection algorithms.

For this, CERN researchers develop deep neural networks to effectively reject background noise from weak signals as well as a modular big data software framework (ROOT) for data handling, analysis, and visualization.

- Data • FCN, spac • CNN (sen imag
- Data: 2bn recorded events/ year with 100m dimensions per example
 - FCN, AE, LSTM¹ for unsupervised anomaly detection: comparing latent space representation (AEs)/ prediction (LSTMs) with observed data
 - CNNs for anomaly detection through image recognition: plotting (sensor) read-outs as image and training CNN to recognize anomalous images for e.g., data quality monitoring (reached AUC = 0.995)
 - Weak signal detection where signal occurs at ratio of 1 in 10³ to 1 in 10⁶
 - Example NN: parameters: 2.3m, epochs: 100, examples: 100k 1m
 - Boosted Decision Trees (ROOT, XGBoost) to improve data resolution





- An institute for commodity risk management teamed up with C B to support regulators to detect trading anomalies from stock market data.
- Knowledge Transfer supports the development anomaly detection and classification algorithms for medical image analysis to diagnose cancer and Covid-19 pneumonia
- Additional use cares, sign policed on in large-scale, noisy, high-dimensional data such as identified on essecurity and fraud attacks, detecting dangerous goods in logistic det, detecting energy consumption anomalies, pharma quality control
- CERN can help organizations use its mode
- Adopt or design and train fit for purpose models to detect in (supervised, unsupervised, semi-supervised)
 - u e direfatiodel development

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- Use CERNC hata as testbed for development of anomaly detection models
 Assu to odel quality by benchmarking with high-quality development of gata



Collaboration with CORMEC and WUR to support national banks and regulators to detect trading anomalies in stock market

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Contract Research

- Use case and requirements by the company
- Code contributed to the OS project
- Development @CERN, benefit for HEP applications

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ROCHE is using CernVM-FS for application and library distribution worldwide. Contract Research for a Company in the financial services sector. Strong interest in this tech for fast reliable worldwide file distribution.

Consultancy/Service

- Specific issue
- Time of experts
- Time of facilities

Bundesdruckerei GmbH works with CERN on next generation ideas for identity management and cryptography and data handling

Collaborative R&D

- General issue
- Case study
- Jointly develop solution

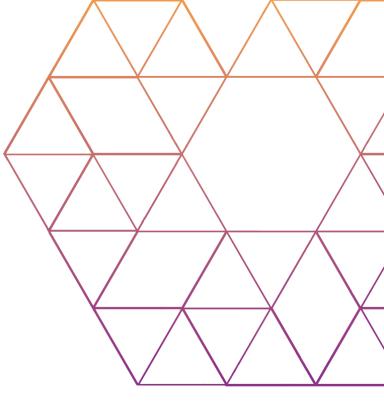
ABB teams up with CERN to build a digital twin of our cooling and ventilation system in order to optimize energy usage

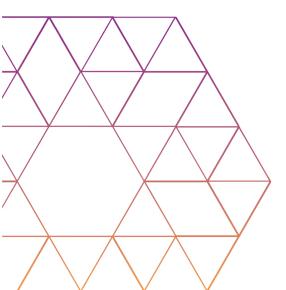


Intellectual Property Management

R&D collaborations Patent portfolio Licence, service & consultancy agreements Advise / 2ND Opinion / Tech Challenge from CERN Expert team

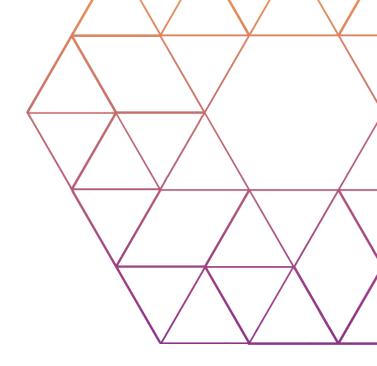
> Collaborative R&D / Codevelopment agreements on specific topic of mutual interest





Challenge Based Innovation program with CERN Experts and/or universities to address specific issue Using CERN labs / CERN openlab for joint R&D, prototyping, benchmarking, testing of software and equipment

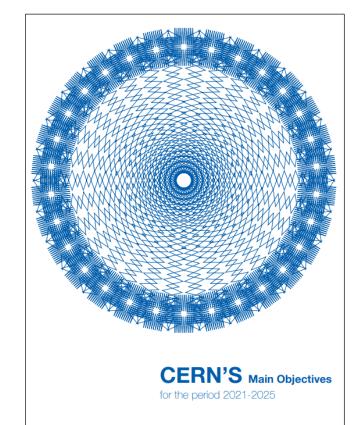
Licensing of CERN technology for commercial use / Support or training on using Open Source Hardware / Software



Facilitation of Knowledge Exchange by sponsoring PhD / Allocation of company resource at CERN / Use of Alumni Network

Environment: a clear priority for CERN

Three main development directions have been identified for environment and sustainability:



Minimise the Laboratory's impact on the environment by implementing CEPS (CERN Environmental Protection Steering) recommendations and defining a Green Procurement strategy Identify and develop CERN's technologies that may contribute to mitigating the impact of society on the environment



Enrico Chesta | CERN Innovation Programme on Environmental Applications

CERN Technology Impact Fund



A mechanism for seeking **donor funding** to support the further development of **CERN technologies** that have high potential to positively impact one or more of the 17 **United Nations Sustainable Development Goals (SDGs)**

- > CERN personnel proposal for a high potential project that creates societal impact.
- Funding sought through the CERN & Society Foundation.
- Partnerships with external organisations in academia, the public sector and industry to maximise the chances of a successful technology transfer to society.

Further information about the SDGs: <u>https://sdgs.un.org/goals</u>

Conclusions

- $\checkmark\,$ CERN innovation culture is inherent to its mission.
- ✓ Knowledge Transfer at CERN objective is to move the innovations and expertise from the lab to society.
- ✓ We have many ways and a lot of flexibility but we need an equally motivated partner to accomplish this mission.
- Some techs and expertise are more readily transferable while others require adaptation via partnerships and funding.

Objectives of the day

- ✓ Improve CERN's understanding of KUKA's needs. What is the generic context of the projects.
- ✓ Improve KUKA's understanding of what CERN can offer in terms of technology and expertise.
- ✓ Identify synergies and precise areas of interest to work together.
- ✓ Clarify how we can work together.
- $\checkmark\,$ Agree on follow up actions and timelines.

Thank you!

Nick Ziogas@cern.ch cern.ch/kt



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