

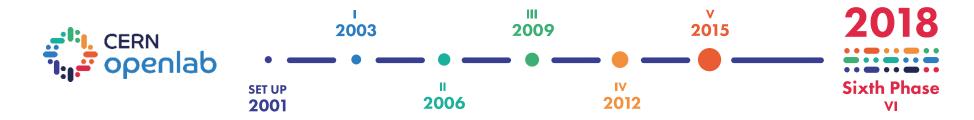
# Partnering with Industry for Machine Learning at HL-LHC

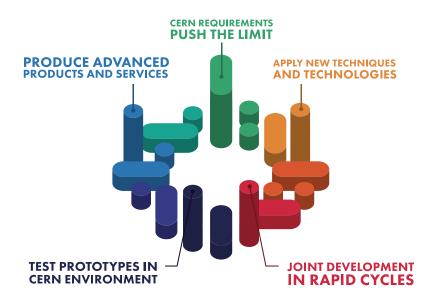
Maria Girone, CERN

CHEP18 Sofia: Track 6

### **CERN OPENLAB**

A public-private partnership between the research community and industry, fostering innovativation





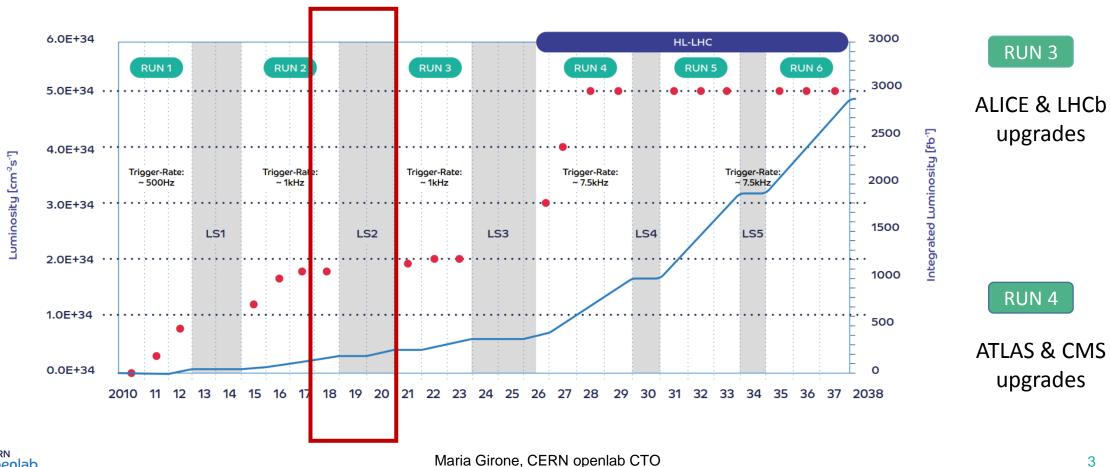
#### **CERN** openlab Collaboration

PARTNERS	CONTRIBUTORS	ASSOCIATES	RESEARCH	
HUAWEI	🕜 rackspace.	<b>♦</b> COMTRADE	VNFN then becomes of their becomes	KINGS College LONDON
(intel)	IBM.	<b>Y</b> andex	<b>╬</b> Fermilab	cimpul. se freelition
ORACLE'	E4	open systems	Newcastle University	EMBL-EBI
SIEMENS	Extreme* Connect Bigured the Network	systems	GSÅ	INNOVATION" VALUE INSTITUTE



### **Upgrades of LHC**

A carefully set out programme of upgrades to increase the scientific reach.



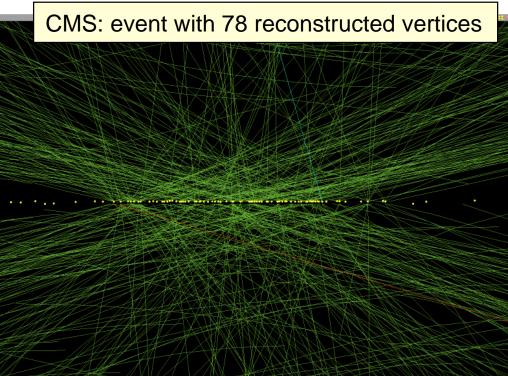


### **Challenges of Complexity**

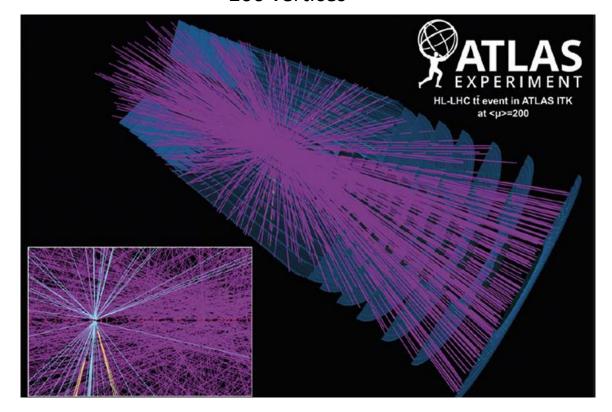
More collisions and more complex data at HL-LHC.

CMS: event from 2017 with 78

reconstructed vertices



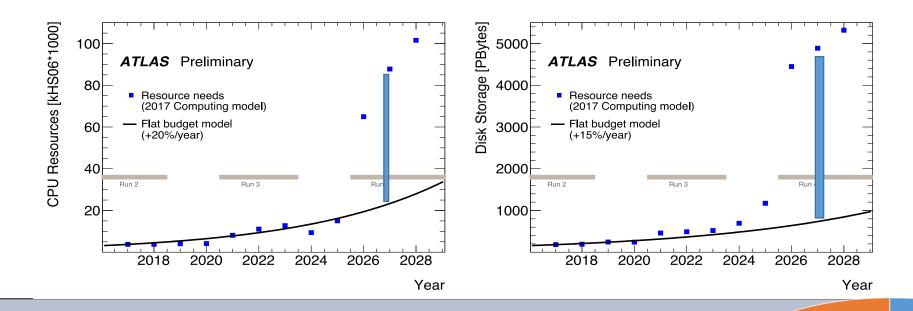
ATLAS: simulation for HL-LHC with 200 vertices





### Resource Gap

Using current techniques, required computing capacity increases 50-100 times



Closing the resource gap in the next decade requires close collaboration with industry.

Technology Evolution

Software innovation,
New Architectures, Techniques
and Methods



### R&D Areas towards Run3 and Run4

The LHC experiments have a multi-pronged approach to closing the resource gap



**Scale out capacity** with public clouds, HPC, new architectures



performance with hardware accelerators (FPGAs, GPUs, ..) optimized software





New techniques with Machine Learning, Deep Learning, Advanced Data Analytics



### JOINT R&D PROJECTS







Yandex





anomaly detection, physics data reduction, benchmarking/scalability, systems biology and large-scale multidisciplinary platforms

Data quality monitoring,









Data Acquisition

Data Analytics, Machine Learning Code modernizatio n Simulation, HPC on the Cloud, benchmarking











Predictive/proactive maintenance and operations



Cloud infra tructures

Cloud federations, containers, scalability







Software Defined Networks, Security



Data Storage

Storage architectures, scalability, monitoring







### **Machine Learning in HEP**

- ML is an area with heavy investment within industry
- The LHC experiments are working closely with industry via CERN openlab
  - Focus on adoption of accelerators (GPUs, FPGAs)
  - Engineering resources dedicated to support the application porting and increase knowhow on deep learning techniques













#### **Data acquisition**

- **Real time event categorization**
- **Data monitoring & certification**
- Fast inference for trigger systems

#### **Data Reconstruction**

- **Calorimeter reconstruction**
- **Boosted object jet tagging**

#### **Data Processing**

- **Computing resource optimization**
- **Predicting data popularity**
- **Intelligent networking**

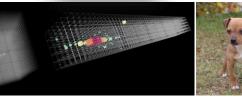
#### **Data Simulation**

- Adversarial networks
- **Fast simulation**

#### **Data Analysis**

- **Knowledge base**
- **Data reduction**
- **Searches for new physics**











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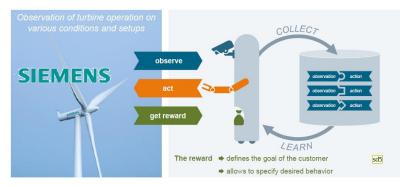
### Monitoring, Automation, Anomaly Detection





- Industrial monitoring and predictive failures
  - Looking at optimizing performance of complex systems
    - Minimize costs and improve resource utilization

#### **Reinforcement learning**



cern openlab







**Y**andex

- LHC magnets, industrial controls, ...
- Detector Health
  - Complex system monitoring to minimize downtime and reduce operations costs
- Resource Utilization (scheduling, data placement, I/O optimization)







### **Data Quality Monitoring**

Monitoring the data continuously is effort intensive and critical

data is monitored by shift teams looking for anomalies in distributions

CMS uses supervised learning on reconstructed data with multihead NN to predict a probability of anomaly in separated channels

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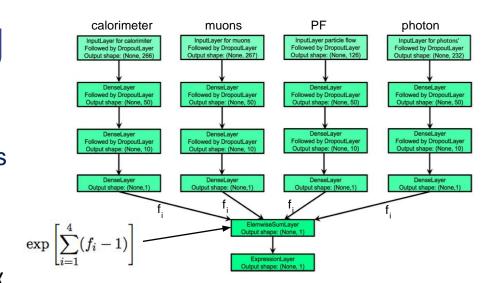
Results are combined to establish the quality of the data

CMS is also working with IBM on automated online monitoring at sub-detector data and metadata level

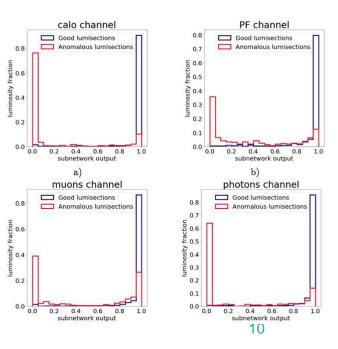
- Predict anomalies in Ecal and Hcal using deep learning recursive NN
- Goal is to integrate detector control systems for a comprehensive monitoring overview

V. Azzolini: Improving the use of data quality metadata via a partnership of technologies and resources between the CMS experiment at CERN and industry









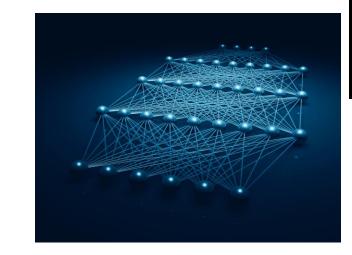
### **Event Reconstruction**

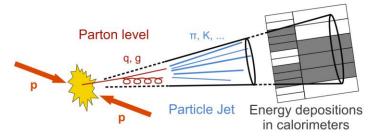
With current software and computer an event like HL-LHC takes 10s of seconds

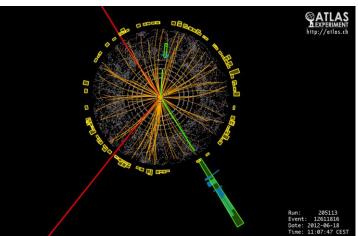
Investigating CNN

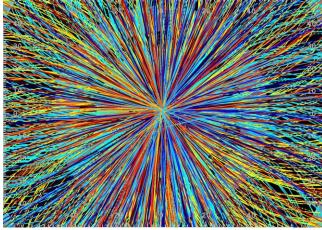
Examine the detector hit information and use 3D image recognition techniques to identify objects

- Recognize physics objects from learned patterns
  - Most ML recognition techniques are designed for regular coordinate systems









Investigate train-on-demand services to be executed on HPC





F. Pantaleo: Distributed training of deep NN models M. Kiehn: TrackML: the Kaggle HEP tracking challenge



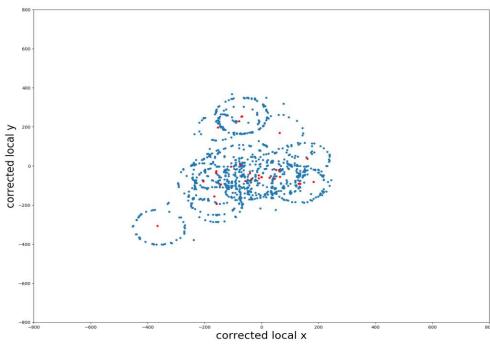
### **Object Identification**

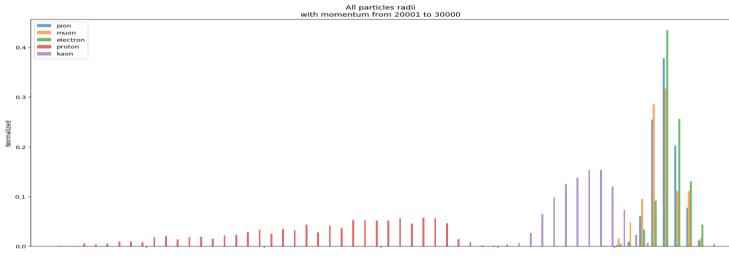
LHCb is exploring particle identification in the RICH detector

- Convolutional neural networks to classify particles based on the radius
- Comparing several modern frameworks: Keras, TensorFlow, and Caffe
- Two ongoing projects in openlab with <u>E4</u>











## Software-based Filtering and real-time Reconstruction

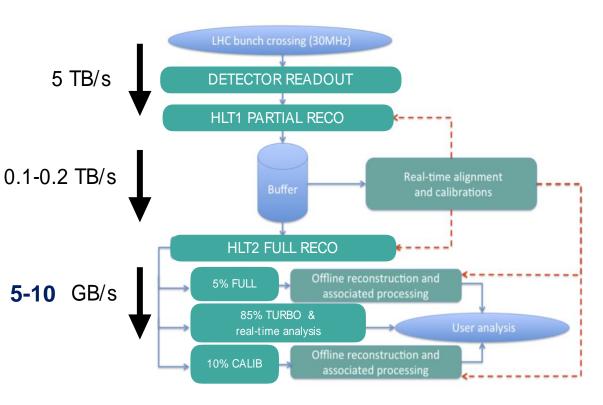
E4 (inte

The ALICE and LHCb experiments will increase their data acceptance rates for Run 3

 LHCb is investigating FPGAs and GPUs to allow reconstruction of 5 TB/s of events in real time.
 Deep learning techniques under investigation

At HL-LHC higher data rates will require more selective triggering and faster reconstruction

- CMS is porting heavy "offline" tasks to real-time processing for HL-LHC
  - Integrate GPUs in the HLT farm to give highquality reconstruction in 200 msec latency (as opposed to tens of sec)



F.. Pantaleo Patatrack

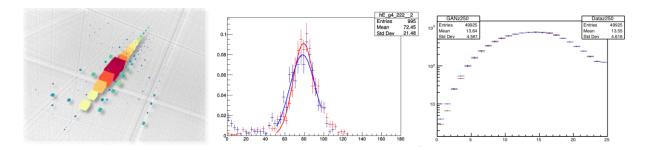


### **Event Simulation**

- Simulation is one of the most resource-intensive computing applications.
- Main R&D areas:
  - Adapting the existing code to new computing architectures
  - Replacing complex algorithms with deeplearning approaches (FAST SIMULATION)



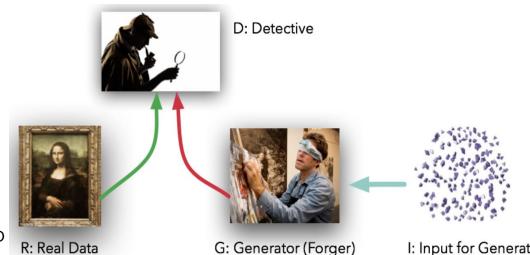




S. Vallecorsa: A machine learning tool for fast simulation

Looking at generative adversarial networks to improve speed, without giving up accuracy of simulated events

- One network attempts to simulate events that match a data distribution (Generator G)
- While a second network tries to distinguish data and simulation (Discriminator D)





### Outlook

- Machine Learning has heavy investments by industry and rapid development cycles
- CERN openlab has active projects with industry in key areas such as data acquisition, processing and analysis
  - Help closing the resource gap in the LHC Run3 and Run4
  - Machine learning is one of the primary focus activities of CERN openlab phase VI
- We have made good progress towards adopting ML for automating data quality monitoring and making faster simulation (without giving up on accuracy)
  - Object identification and full event reconstruction are active areas of investigation
- ML is a fast moving field and we are hoping for more breakthroughs

