



Joint Public/Private Research Projects

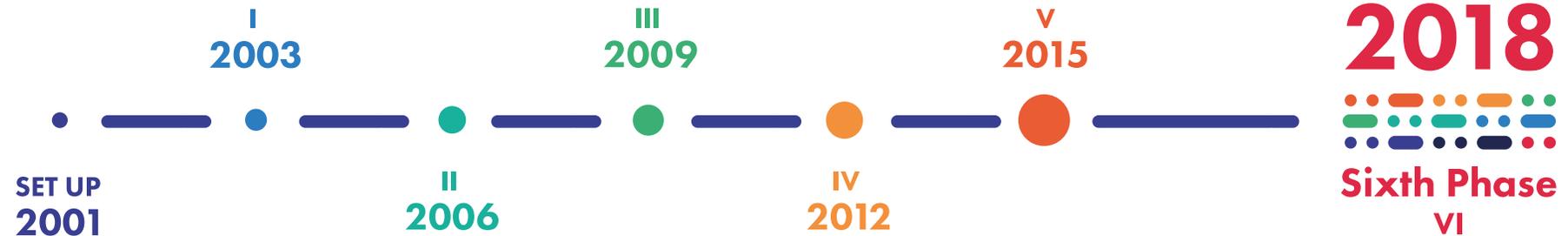
Alberto Di Meglio – CERN openlab Head

2/10/2019

1

DRIVING INNOVATION SINCE 2001

Now in our sixth three-year phase



CERN openlab Collaboration Board 2017

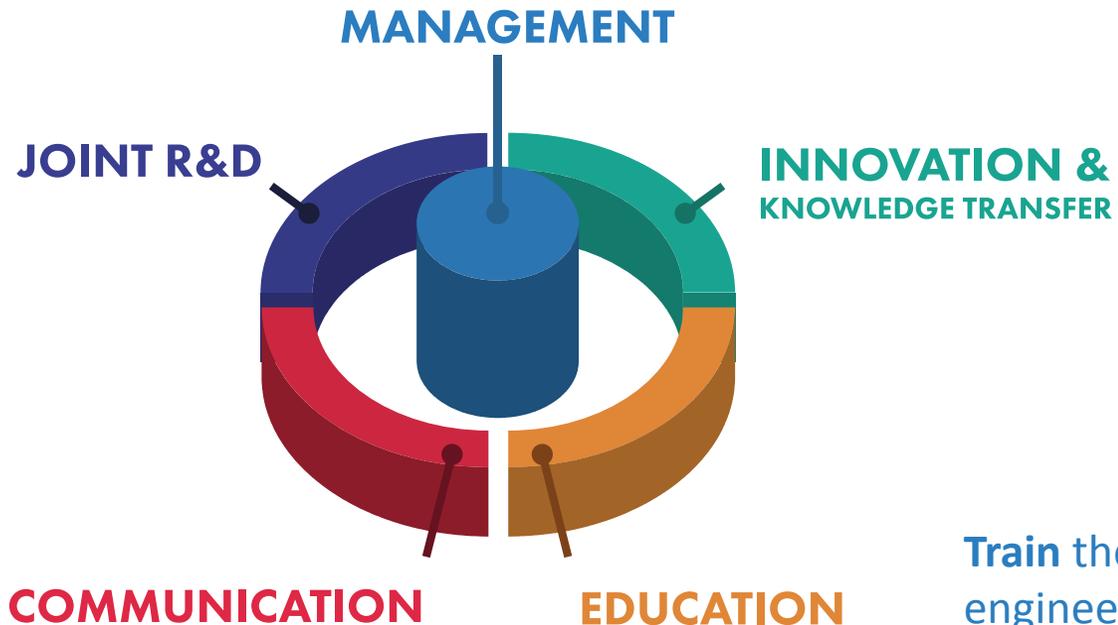


CERN OPENLAB'S MISSION

Our recipe for success

Evaluate and test state-of-the-art technologies in a challenging environment and improve them in collaboration with industry.

Communicate results, demonstrate impact, and reach new audiences.



Collaborate and exchange ideas with other communities to create knowledge and innovation.

Train the next generation of engineers/researchers, **promote** education and cultural exchanges.

JOINT R&D PROJECTS (PHASE VI)

T-Systems



E4
COMPUTER
ENGINEERING

Google

High-bandwidth fabrics,
accelerated platforms for
data acquisition



ORACLE



Fast simulation, Data
quality monitoring,
anomaly detection,
physics data reduction,
benchmarking/scalability,
systems biology and
large-scale multi-
disciplinary platforms

be|studys
Research and innovation



САМАРСКИЙ УНИВЕРСИТЕТ
SAMARA UNIVERSITY

SIEMENS

Extreme
Connect Beyond the Network



Software Defined
Networks, Security

CERN openlab - Challenges in Computing



Simulation, HPC
on the Cloud,
benchmarking



HUAWEI



Cloud federations,
containers, scalability

TU/e

Storage architectures,
scalability, monitoring

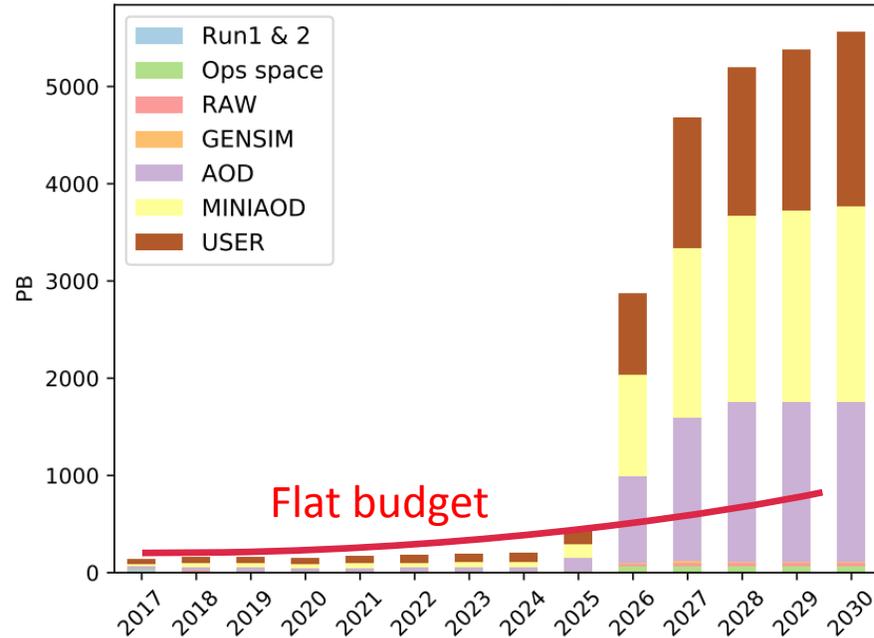
ORACLE



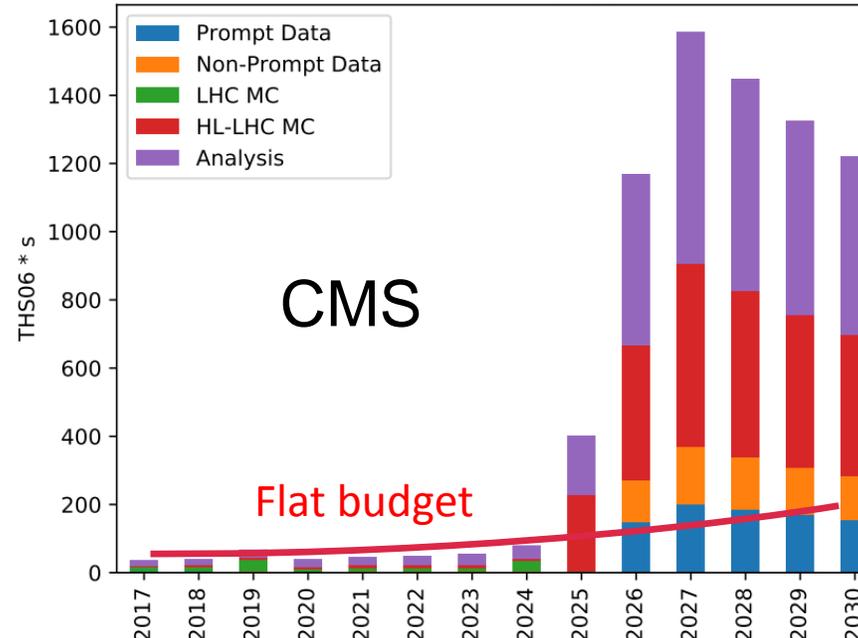
Computing @ HL-LHC



Data on disk by tier



CPU seconds by Type



<https://arxiv.org/pdf/1712.06982.pdf>

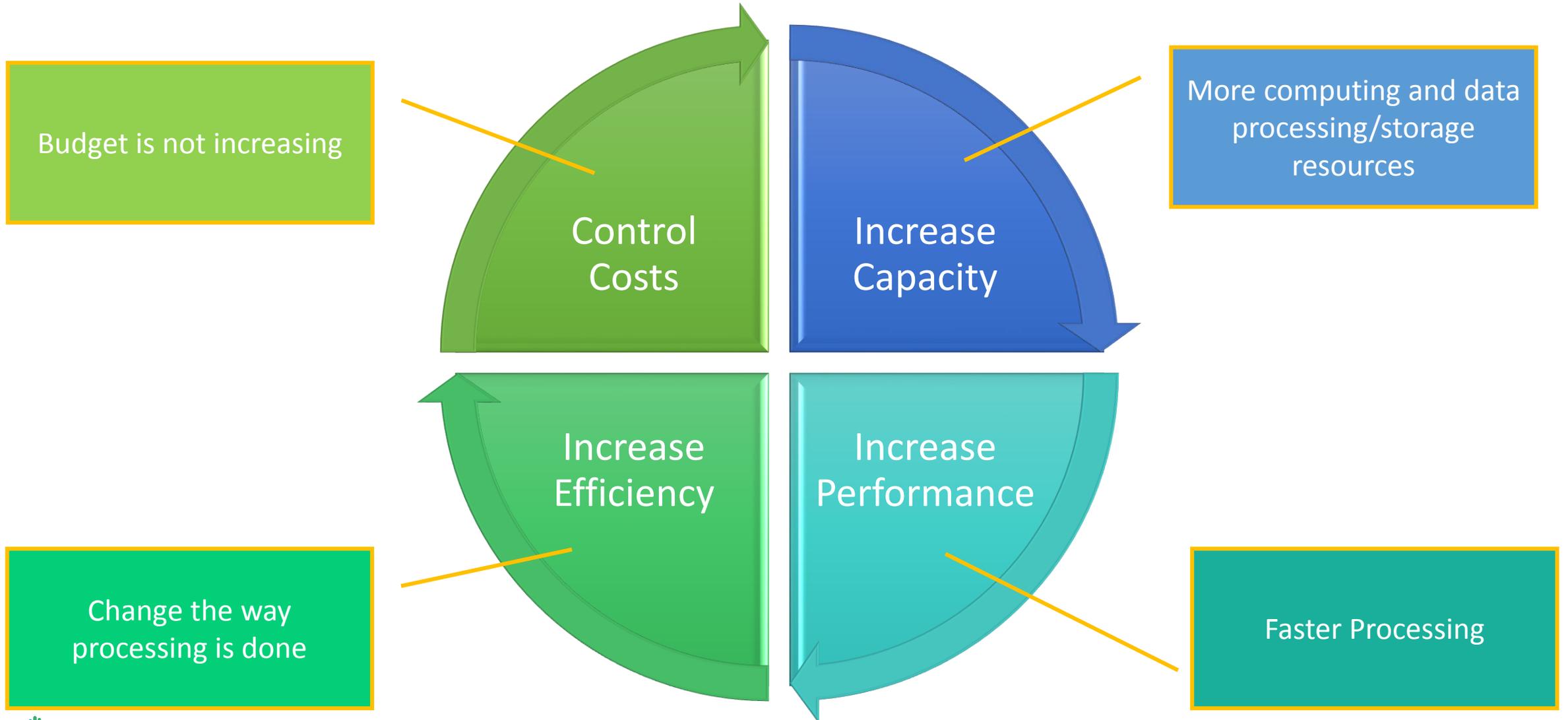
Raw data volume increases exponentially

Processing and analysis load

Technology at ~20%/year will bring x6-10 in ~10 years

Estimates of resource needs x10 above what is realistic to expect

Computing and Data Challenges



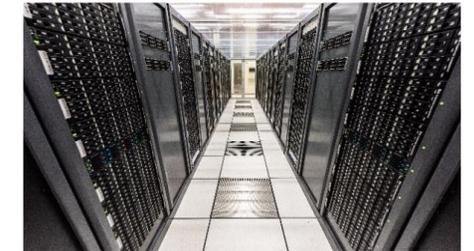
Three Main Areas of R&D



Increase **data centre performance** with hardware accelerators (FPGAs, GPUs, ..) optimized software



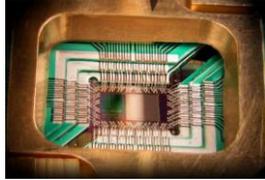
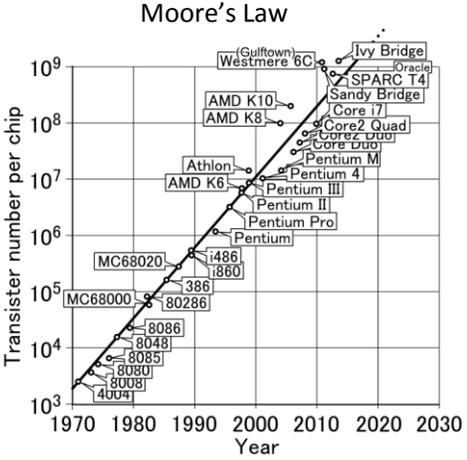
Scale out capacity with public clouds, HPC, new architectures



Change the computing paradigms with new technologies like **Machine Learning, Deep Learning, Advanced Data Analytics, Quantum Computing**



New Computing Platforms



1995

2000

2010

2014

2015

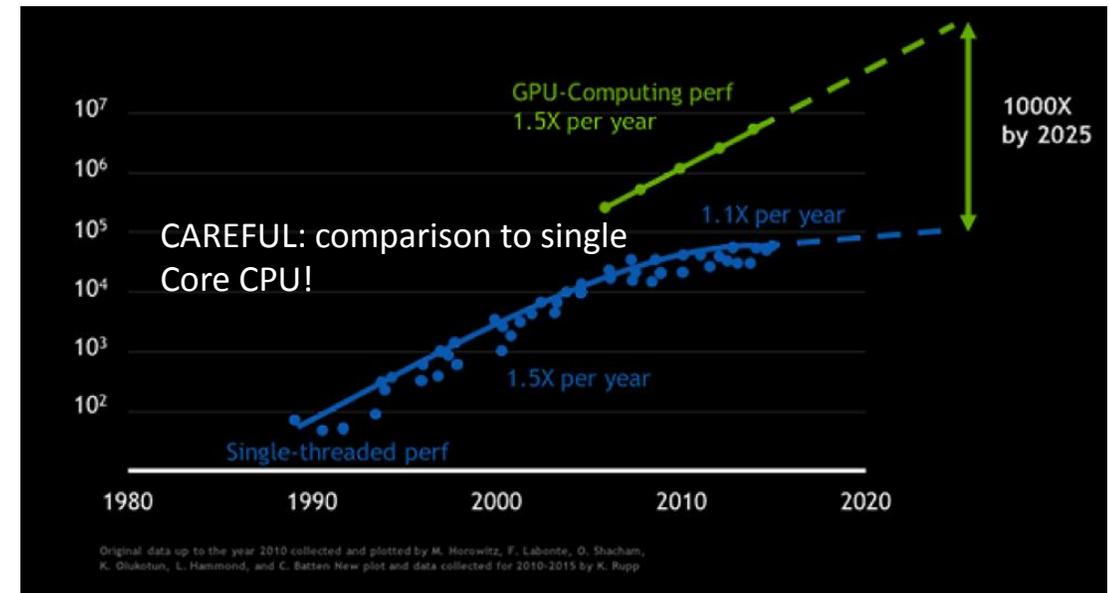
... and beyond

Radically new computing platforms are rapidly moving from pure computer science to realistic devices, e.g. **Neuromorphic Computing** and **Quantum Computing**

A **Quantum Computing Initiative** has been launched in December 2018 as a long-term investigation activity

GPU Acceleration

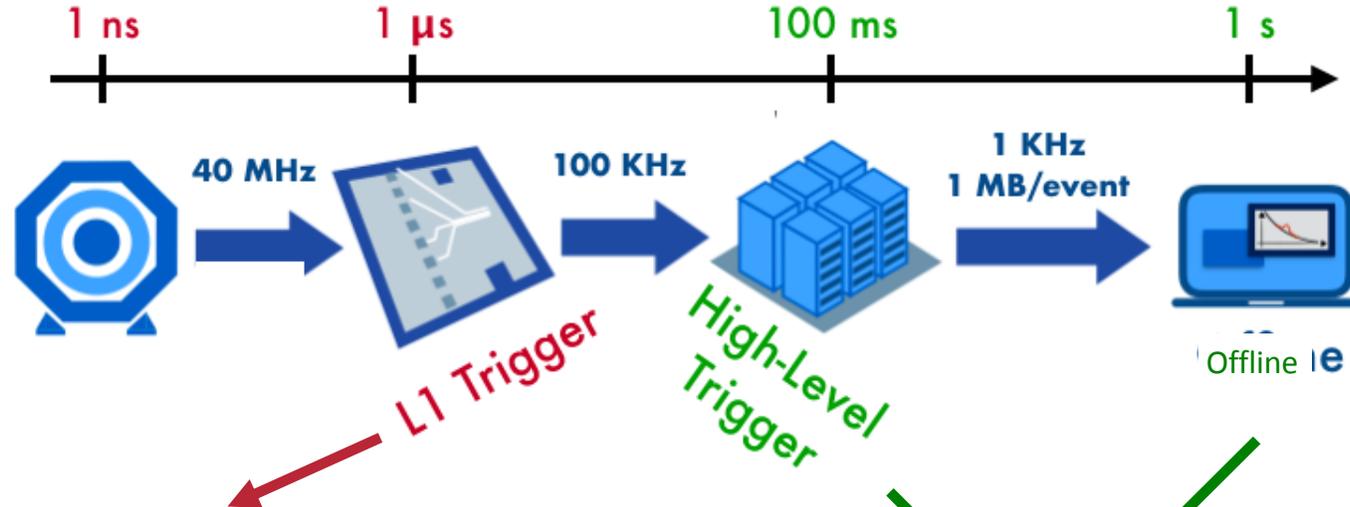
- Extremely parallel architectures are widely used in industry and other sciences
 - Can provide better performance/price and energy efficiency
 - GPU performance is on an 18-month doubling cycle
- R&D effort in the LHC experiments to offload parts of the reconstruction workflows to GPUs
 - For both high level trigger and offline (on CUDA)
- ML/DL frameworks profit from GPU acceleration and substantially reduce turnaround times
 - Industry has invested heavily in optimizing machine learning libraries on GPUs (training and inference)
 - Increasing R&D effort in HEP



Industry is producing low-power GPUs

- Can be used to upgrade existing machines
- Latest **NVIDIA** GPUs (i.e. T4) use 25% of the power at significantly lower cost
- **AMD** GPUs offer very interesting performance/price ratios
 - Software is catching up

FPGA-accelerated Deep Learning

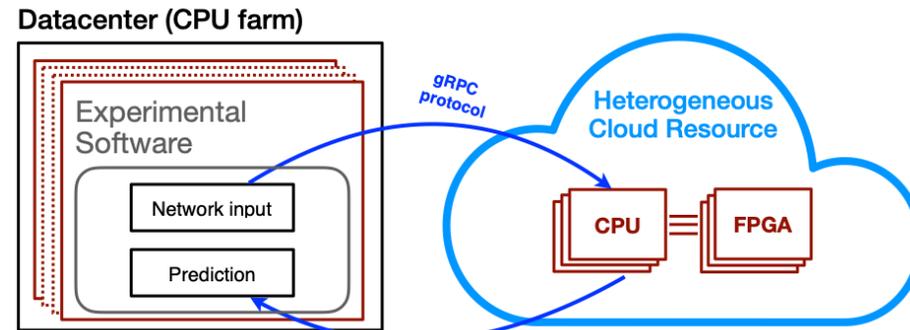
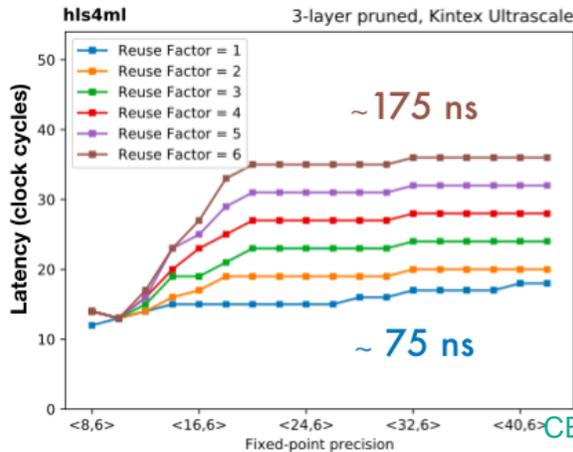


HLS4ML: Fast inference of deep neural networks in FPGAs for particle physics

[\(JINST, 13, 2018\)](#)

SONIC: FPGA-accelerated machine learning inference as a service for particle physics computing

[\(arxiv.1904.08986\)](#)



HEP Quantum Computing Initiative



- Launched in November 2018
 - Event at CERN > 400 participants
- Created a catalog of possible applications in HEP
- Established links with major QC activities in Europe and US

Currently discussing (among others) with



Goal to establish a “hub” or “centre of excellence” in Quantum Computing applications

Research paths in Quantum Computing



- Tools and methodology development on emulators and simulators
 - Proof-of-concept algorithms for HEP workloads
 - Compare results on real devices
- Understand the role that CERN can play as part of broader QC development initiatives
 - APIs and user interfaces to access QC systems
 - Engineering aspects of QC installation (cryogenics, material science, ..

Examples

1. Quantum SVM for Higgs boson searches (University of Wisconsin)
 - Higgs coupling to top quark
 - Higgs decay to muons
 - Double Higgs production
 - Dark Matter
2. Quantum SVM for boosted Higgs (University of Washington)
 - Identify jets originating from Higgs boson decay in order to study Higgs coupling to Standard Model particles
3. “Classical” optimisations
 - Track reconstruction
 - Grid workload optimisation

Storage Evolution

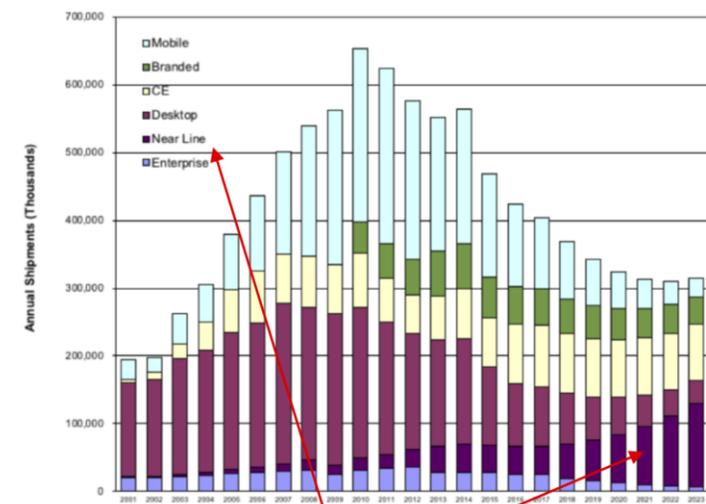
Storage remains a challenge as data volumes increase

Similar to computing, industry is specializing

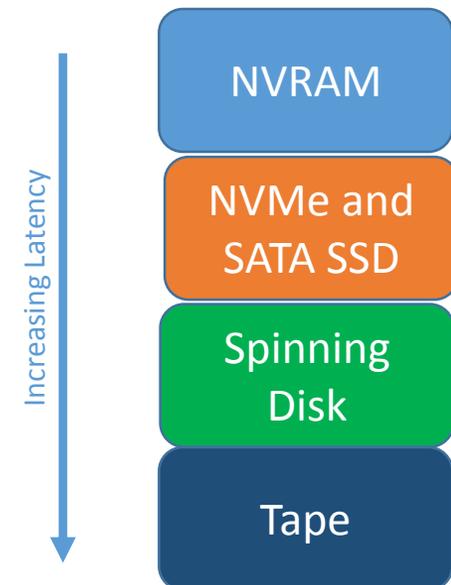
Improvements for HDD continue for capacity but slower for bandwidth and IOPS (Shingled disk, Helium drives, MAMR and HAMR). On-board processing (ARM/RISC V) might provide higher bandwidth, but still early to say

High capacity, fast and affordable SSD-flash based are taking over, thanks to the move from SAS/SATA to NVMe protocols

Smaller size, but extremely fast solutions like 3D XPoint NVRAM blur the boundary between memory and storage, but are still expensive



HDD type used for Cloud + HEP



Archival Storage

Tape remains the preferred (cheap) solution of HEP experiments for long-term storage

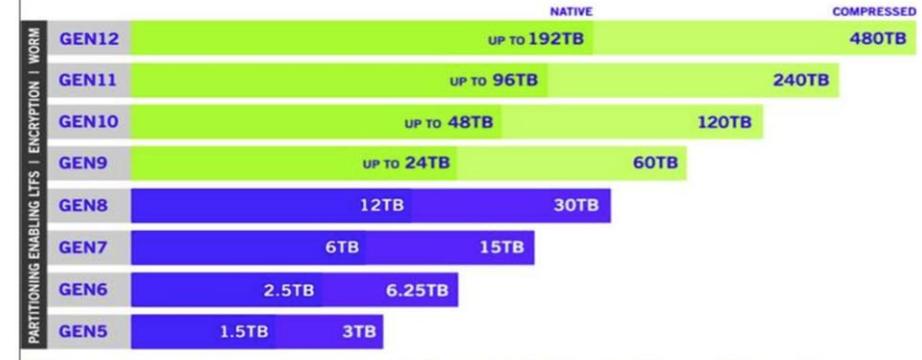
As data volumes increase, LHC experiments are looking at larger roles for tape for data with very structured access

Tape roadmap extends another decade and a factor of 20 in capacity

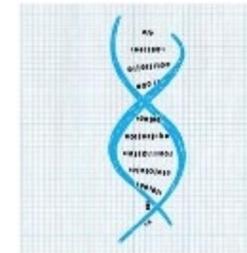
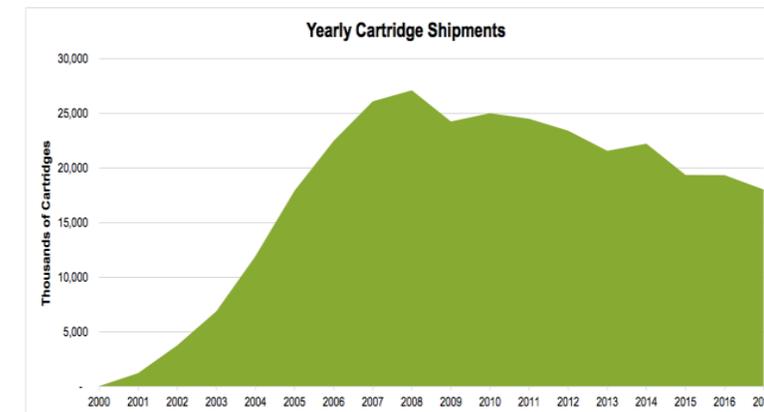
Disruptive technologies on the far horizon for archiving

DNA Storage relying on the density and stability of DNA molecules to encode large volumes of data

LTO ULTRIUM ROADMAP ADDRESSING YOUR STORAGE NEEDS

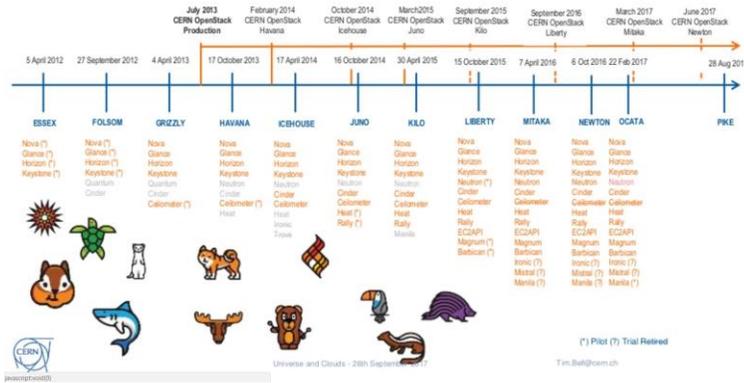


Unit Shipments: Calendar Year



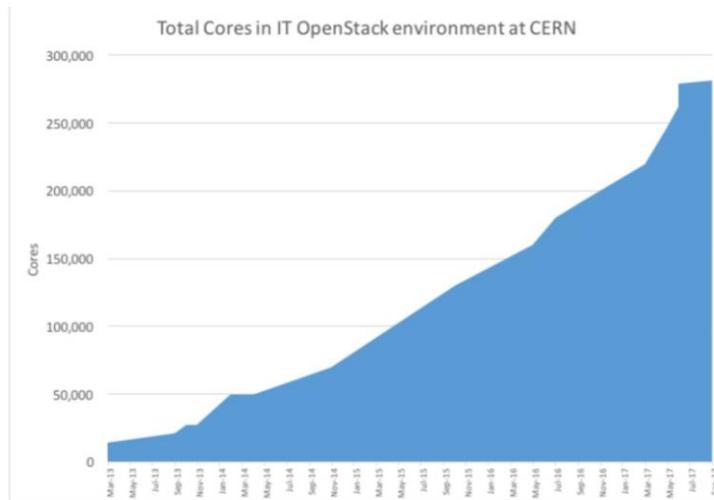
Cloud Computing

CERN OpenStack Service Timeline



CERN runs the **largest OpenStack-based private cloud** in the scientific research world

Use of cloud technology allowed to **increase the resource provisioning in numbers and efficiency** by orders of magnitude



Currently >8000 hypervisors, 281K cores running 33,000 VMs

However, it is still limited by the maximum physical capacity of the DC

Hybrid public-private models are being considered, also to provision types of resources that we do not have (large GPU farms and dedicated data training facilities)



Investments at HPC Sites

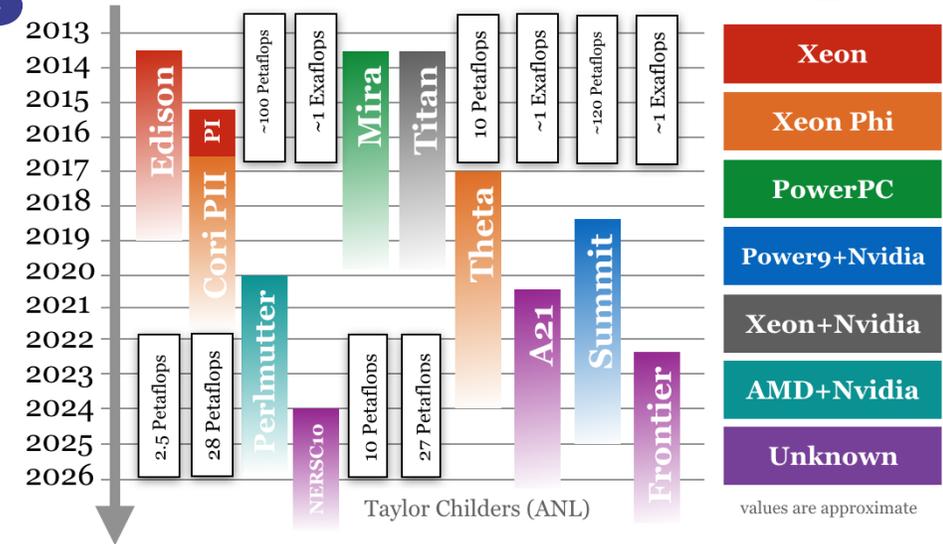
HPC sites will grow by a factor 20 on the time scale of HL-LHC

- Large investments in USA, Europe and Asia
- Mostly available to sciences through allocations

High performance accelerated hardware will provide the bulk of the processing capacity (~10% expected from CPUs) .

Implies:

- R&D on software and techniques to better exploit processing architectures, low latency networking, edge services, ..
- R&D in interoperability across HPC (and Cloud) sites, via consistent services (containers , CVMFS, ..)



Engaging and working together with Super Computing Centres is essential for HEP and opens questions about training needs in software engineering

Name	Supplier & Model	Compute power (Peak) [GFLOP/s]	Energy consumption under Linpack (kW)
Piz Daint gpu	Cray XC50 (Hybrid)	25'326'000	2'272
Piz Daint mc	Cray XC40 (Multicore)	2'193'000	734

Installed 2017

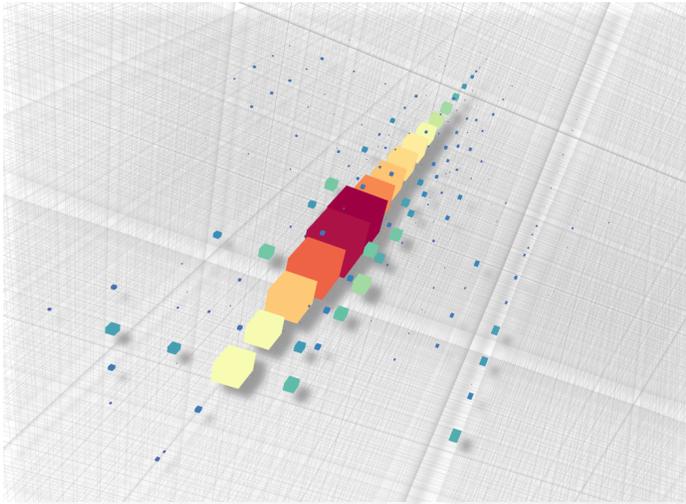
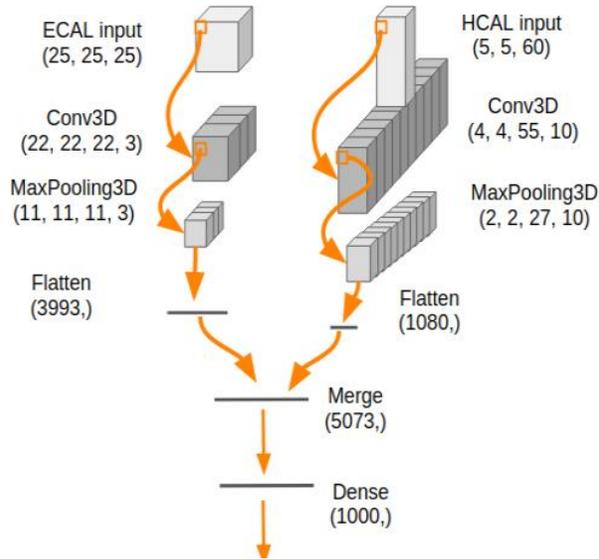


HPC Joint Activities

- **Active effort** in CERN openlab to support WLCG exploration of **common challenges in HPC integration** from individual LHC experiments documents
 - Draft document available at <https://docs.google.com/document/d/1AN1d6NukhBsKnNH1MVvszqWdpcMfFGaYQMEVnS01Tc/>
- Discussions with CSCS, Jülich, Oak Ridge to establish **pilot projects**
- Discussions with PRACE (workshop in October 2018)
 - <https://indico.cern.ch/event/760705/>
 - Proposal of MoU between SKA, CERN and GÉANT under evaluation
- **Leadership of CERN tasks** in EC-funded project **DEEP-EST** (Modular Supercomputing prototype at Julich)
 - 1 FTE for ECAL and HCAL reconstruction on GPU

Courtesy of Maria Girone – CERN openlab

Artificial Intelligence



LHC Experiments have used **Machine Learning** for quite some time now (e.g. BDT), now investigating **Deep Learning** in several areas

New DL algorithms and **fast hardware** can accelerate almost all the typical HEP workloads (data acquisition, event classification, reconstruction, simulation, etc.)

Many ongoing projects in collaboration with Intel, Google, IBM, T-Systems, and others on algorithms (fast simulation, GANs, anomaly detection), tools (Spark, TensorFlow) and hardware (GPU, TPU, FPGA)

Deep Learning for fast detector simulation

Detailed simulation of subatomic particles interactions is essential

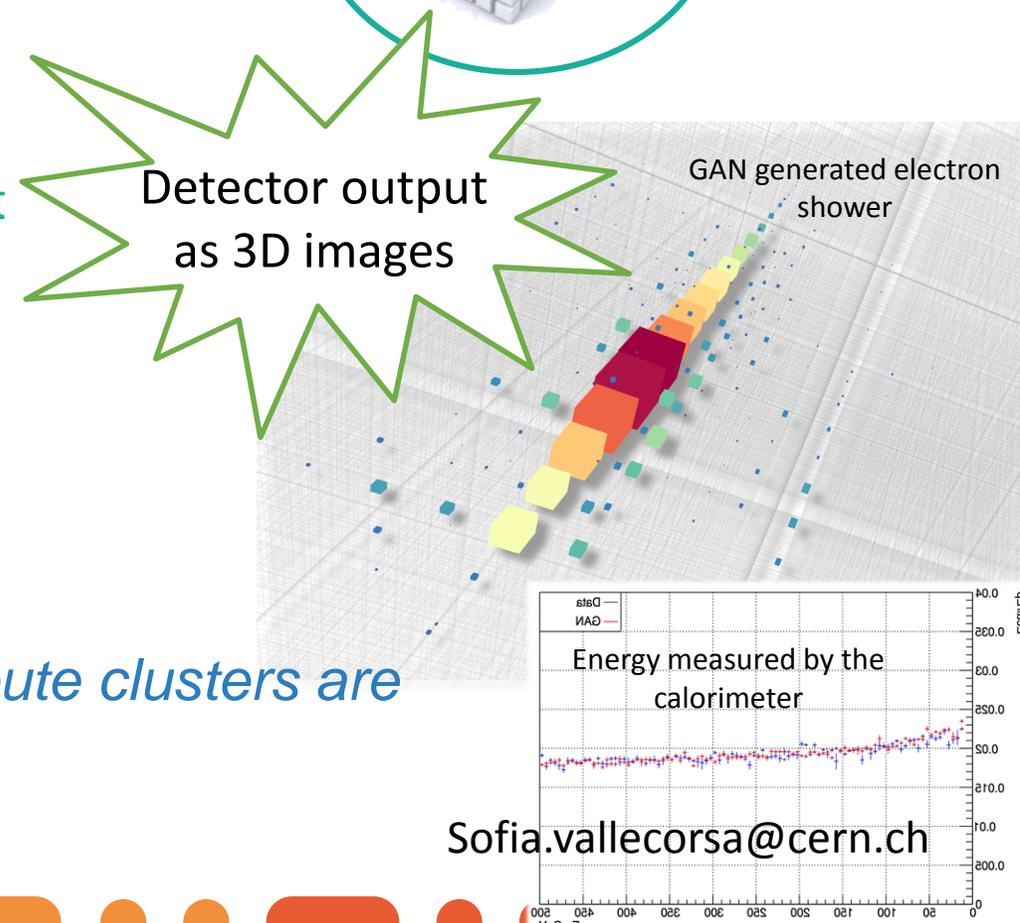
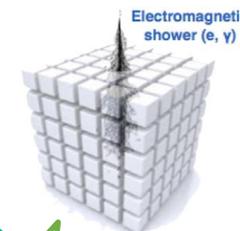
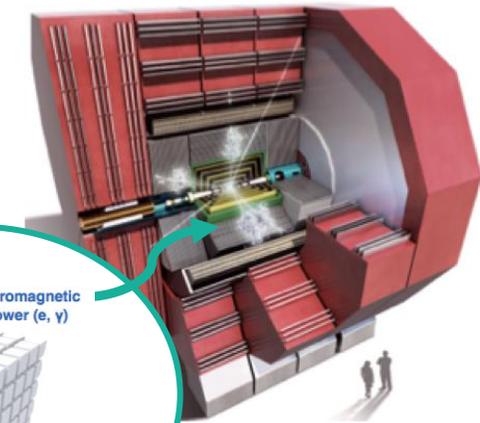
Monte Carlo approach is not fast enough for HL-LHC needs

3D convolutional GAN generate realistic detector output
>2000x faster than Monte Carlo on a Intel Xeon processor

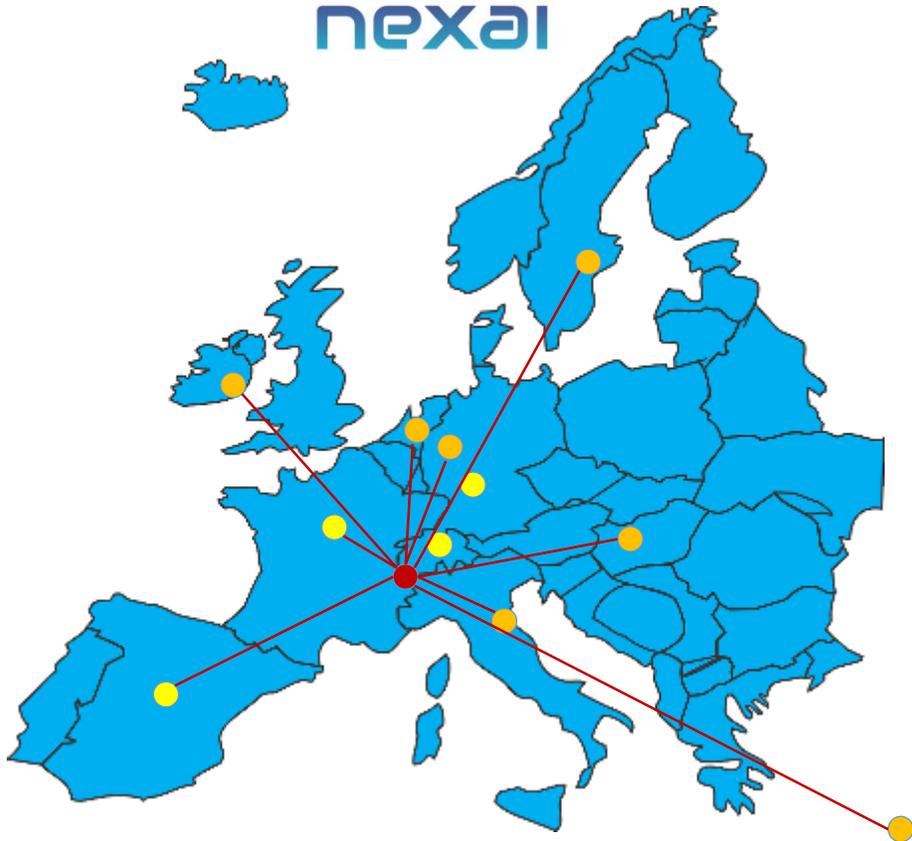
Training takes ~1 day on NVIDIA P100

Use parallel approach to distribute training across multiple nodes

To generalise systems with larger memory and compute clusters are needed → HPC workload



Centre of Excellence in AI



Currently leading the preparation of a proposal for the creation of a Centre of Excellence in Artificial Intelligence

EC call ICT-48-2020

Tier 1-2 HPC Centres

Leading academic Institutes in AI research

Four applications domains:

Fundamental physics research

Personalised Medicine

Sustainable Agriculture/Green Economy

Dematerialized Banking and Security

Medical Research

Drug discovery

Combination of computing, simulations, textual descriptions of symptoms and expected outcomes

Clinical Trials

Optimization, targeted recruitment and enrollment via automated analysis of medical records, pipeline assessment/review

Image Analysis

Machine-learning-based content recognition, offline and (quasi) real-time anatomy localization, tissue segmentation, support to diagnostics, automated/hybrid surgical rooms

Clinical Decision Support Systems

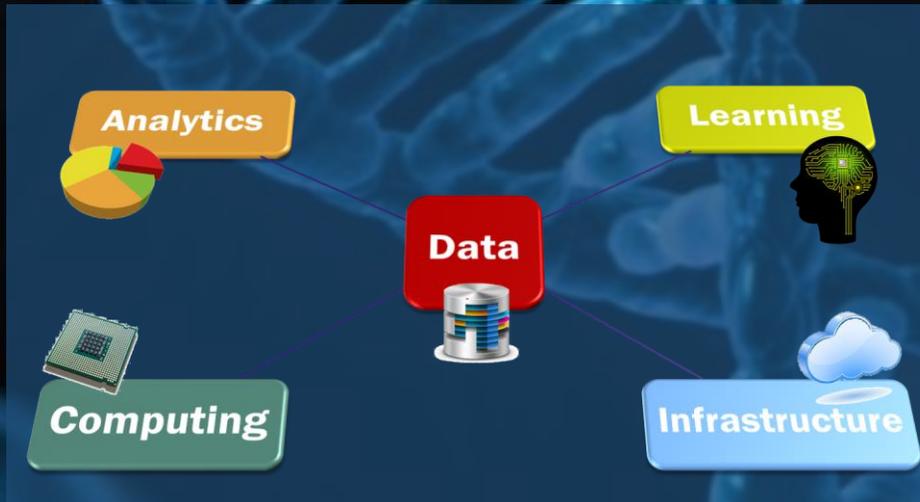
Large-scale data integration/analysis, structured/unstructured data, anonymization, treatment effectiveness and risks predictions

Trends and simulations

Preventive treatments, epidemics assessment and control, optimization of effort/budget allocations

Genomics

NGS data, analysis, storage, transfer at multi-PB scale, integration with other *omics and clinical data, personalized treatments



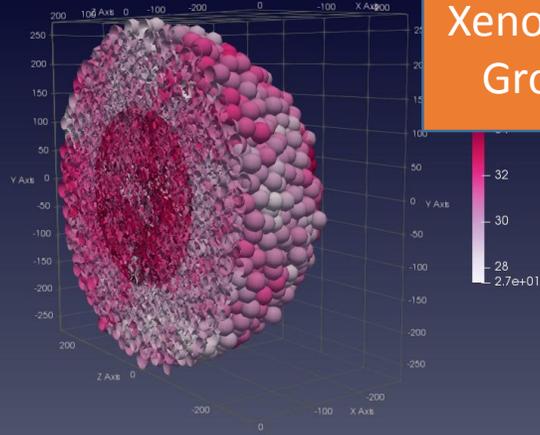
Personalised Medicine

Monitoring/assessment, suggestions, personalized medicine, exploitation of data from wearable devices, social fitness/health platforms

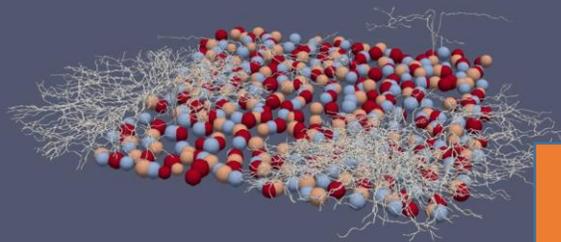
Biological Dynamics:



BioDYNAMO
BIOLOGY DYNAMICS MODELLER



Xenotransplant Tumor
Growth Simulation



Retinal Mosaics
Development

Collaborative platform to support biological simulation experiments from single computers to large-scale cloud-based infrastructures

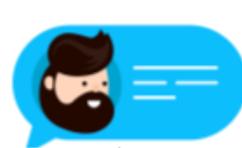
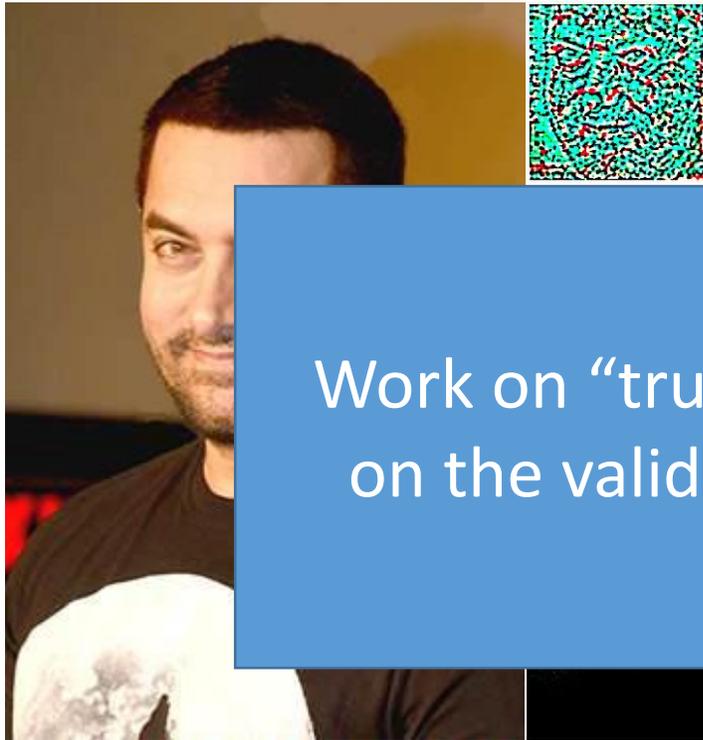
Extensible modular architecture able to support different cell types definitions, behaviours and interactions

Growing library of tools, cell definitions, simulation workflows and models

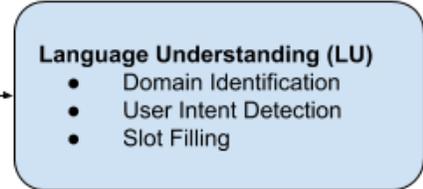
Potentially applicable to many fields, from developmental studies to in-silico treatments investigations to environmental application (sewage, waste treatment, etc.)

Open source license and contribution model

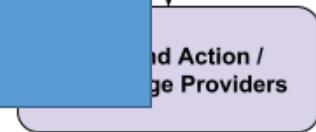
Machine Assisted Diagnostics



Text input: "What should I do if I think I have the Flu?"



Semantic Frame
Request_treatment
Diagnosis = Flu
Diagnosis_type = Self



Work on "trustable AI" is important for its impact on the validation of output of Neural Networks

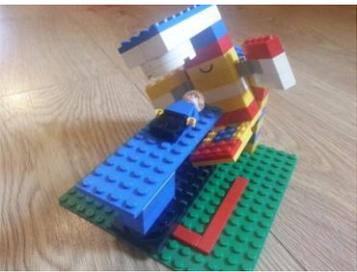
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Young, 2000 - <https://doi.org/10.1098/rsta.2000.0593>

Phenotype Modelling and Genetic Evidence

Conversational medical chatbots and Natural Language Processing

LINAC Personalised Preventive Maintenance



Models trained on existing data from LINAC2/4

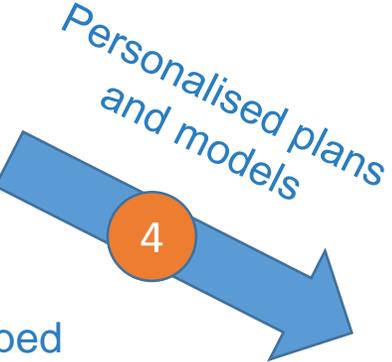


Initial maintenance plans



EQUIPMENT MAINTENANCE PLAN										
COMBENT TYPE	DESCRIPTION	LOCATION	EQUIPMENT PLAN FILE		NAME FILE VALUATED					
ITEM #	MAINTENANCE DESC/DESCRIPTION	REPLACEMENT DATE	EQUIPMENT ID	TYPE	PROCEDURE CODE #	EST. TIME (HRS)	SPECIAL INSTRUCTIONS		APPROVAL (HRS)	

Initial models based on MTTF and know fault patterns



EQUIPMENT MAINTENANCE PLAN									
COMBENT TYPE	DESCRIPTION	LOCATION	EQUIPMENT PLAN FILE		NAME FILE VALUATED				
ITEM #	MAINTENANCE DESC/DESCRIPTION	REPLACEMENT DATE	EQUIPMENT ID	TYPE	PROCEDURE CODE #	EST. TIME (HRS)	SPECIAL INSTRUCTIONS		APPROVAL (HRS)

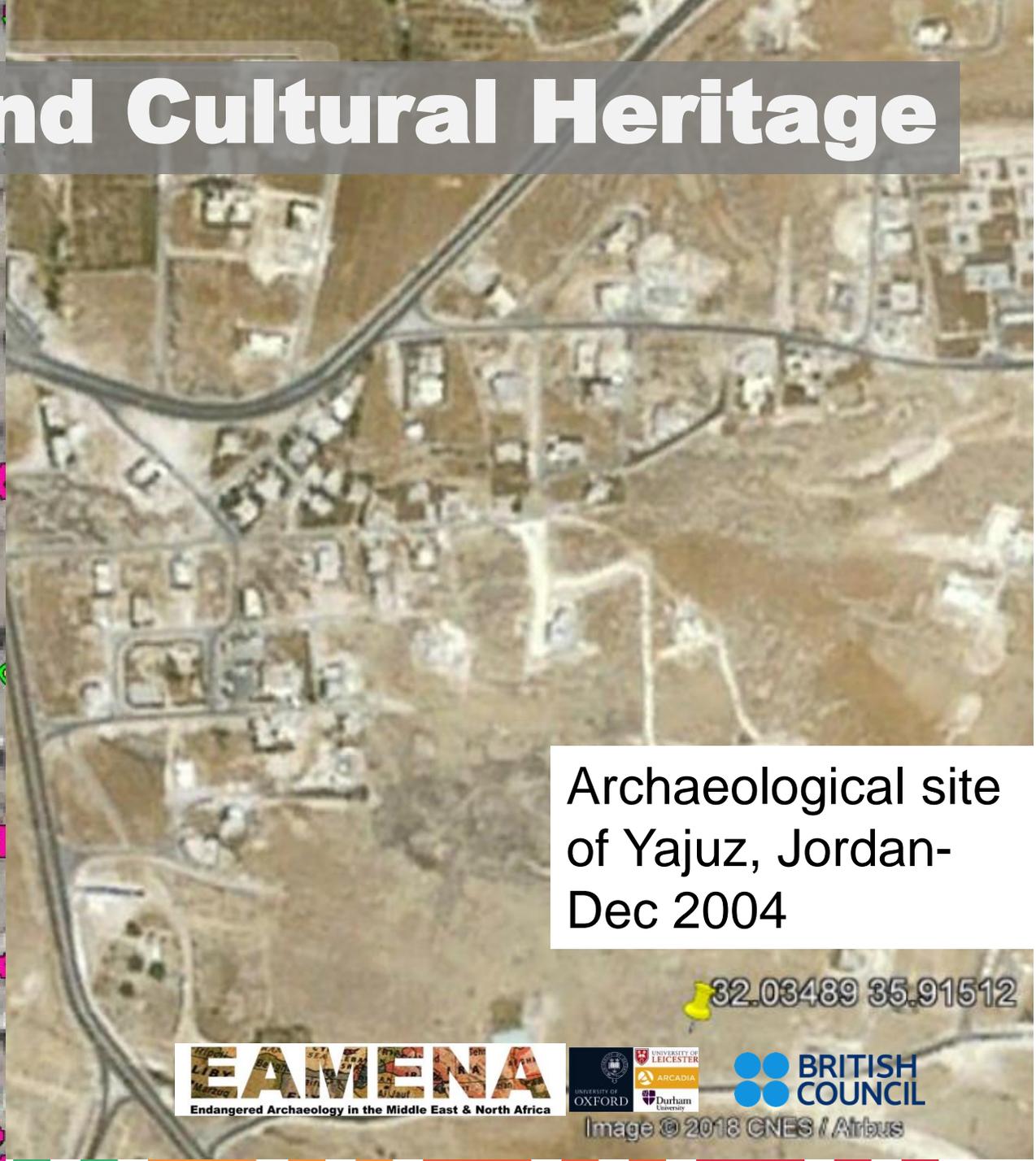
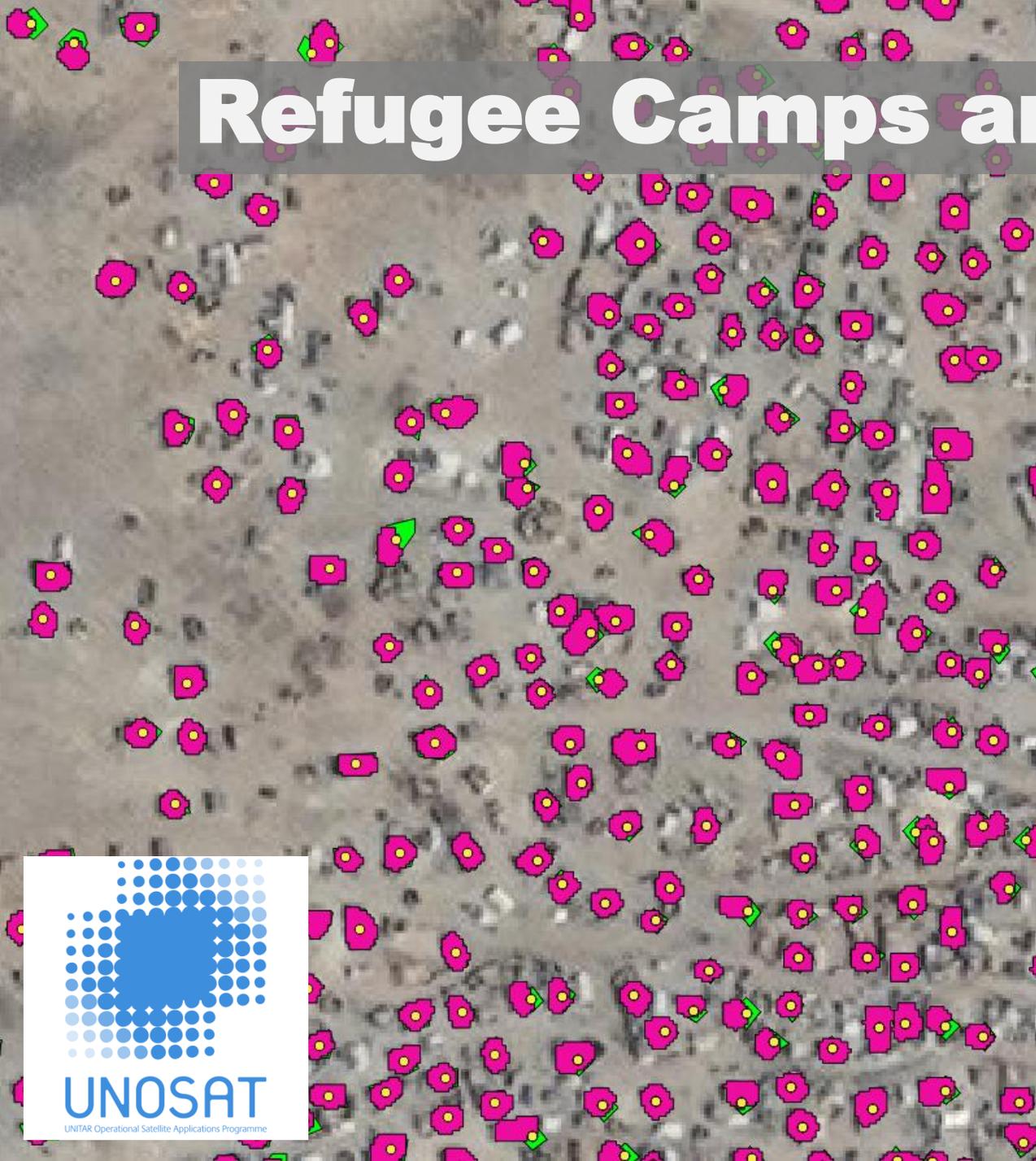


Models are shipped to installations



Real LINACs data In different environments and operation constraints

Refugee Camps and Cultural Heritage



Archaeological site of Yajuz, Jordan-
Dec 2004

32.03489 35.91512

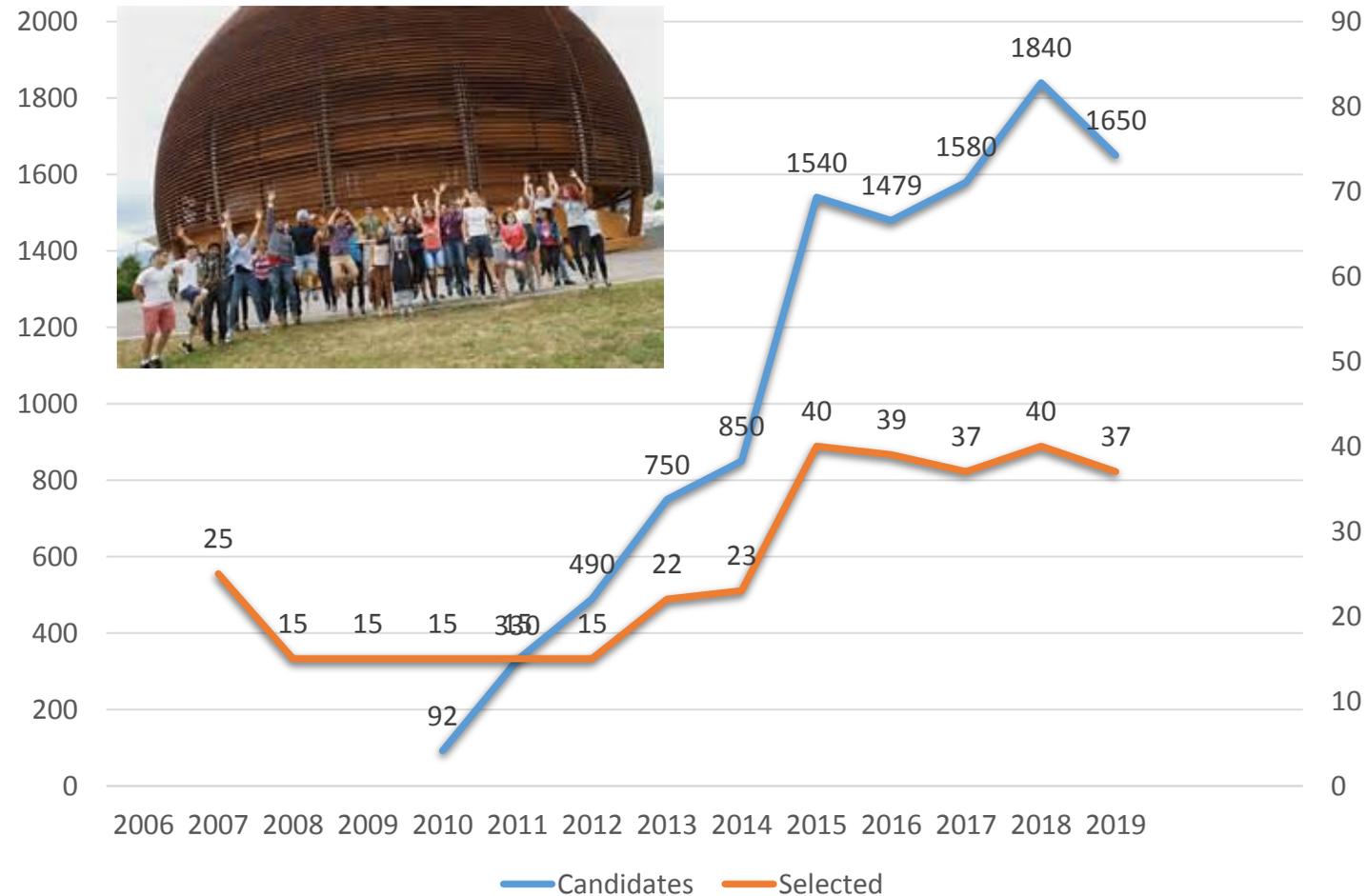


Image © 2018 CNES / Airbus



Education and Training

SUMMER STUDENT PROGRAMME



In 2019

- 1650 applicants
- 37 selected students
- 14 lectures
- Visits to external labs and companies
- Lightning talks session
- Technical reports



Thanks!

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