



Meeting Future Software Challenges in High-Energy Physics

Graeme A Stewart, CERN EP-SFT

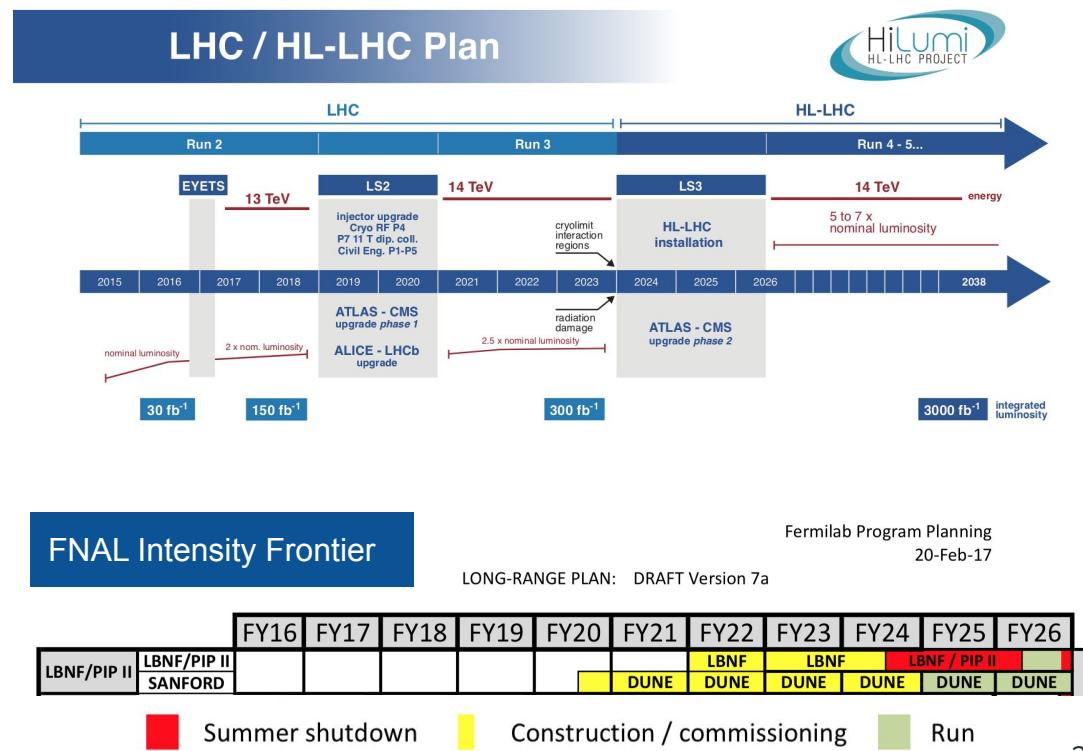


University of Geneva, 22 May 2019

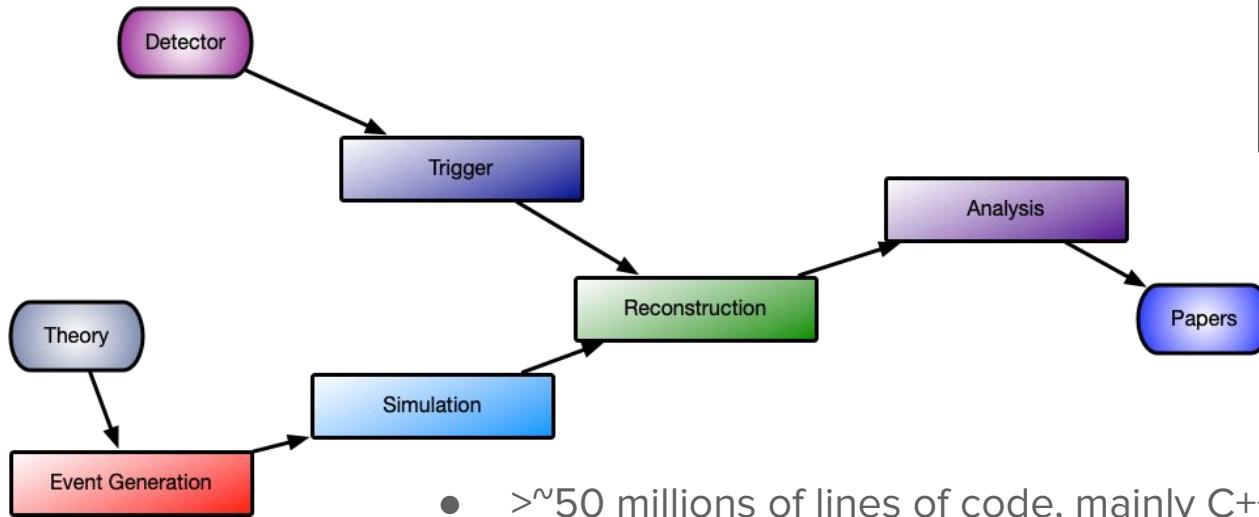
HL-LHC and the Intensity Frontier

Our mission:

- Exploit the Higgs for SM and BSM physics
- b, c, tau physics to study BSM and matter/anti-matter
- Dark matter
- QGP in heavy ion collisions
- Neutrino oscillations and mass
- Explore the unknown



An Overview of HEP Software

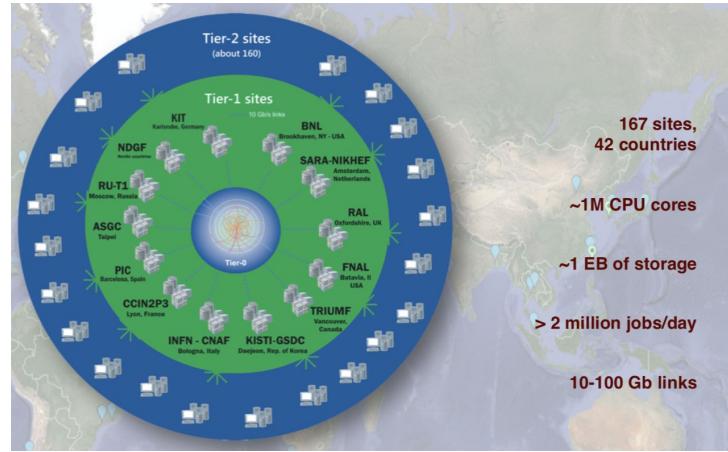


This is the “traditional” view and **how this changes** in the future is an important topic for our discussions

- >[~]50 millions of lines of code, mainly C++, a lot of Python
 - Commercial development cost ~500M CHF
- Critical part of our physics production pipeline, from triggering all the way to analysis and final plots as well as simulation
- Significant pieces of software are already shared by most experiments:
 - Event generators, Geant4, ROOT

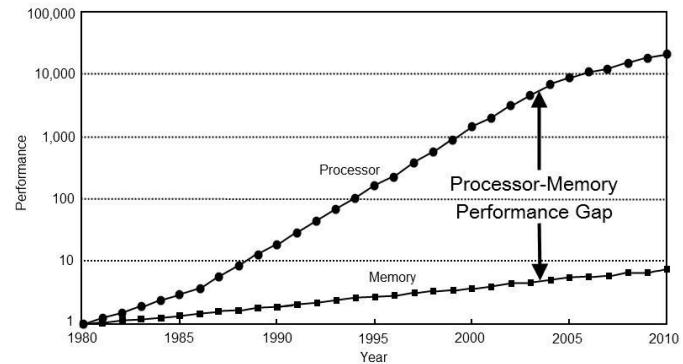
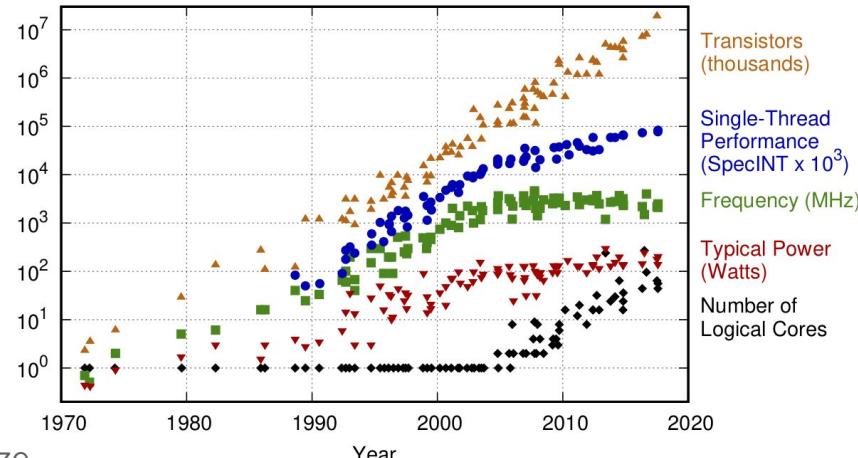
HEP Computing

- Tasks broken into jobs by experiment production systems (levels of parallelism)
 - Tasks → job → events → algorithms
- LHC experiments use
 - 1M CPU cores every hour of every day
 - Store 1000PB of data (600/400PB tape/disk split)
 - We are in the exabyte era already
 - 100PB of data transfers per year (10-100Gb links)
- This is a huge and ongoing cost in hardware and human effort
- With significant challenges ahead of us to support our ongoing physics programme



Technology Evolution

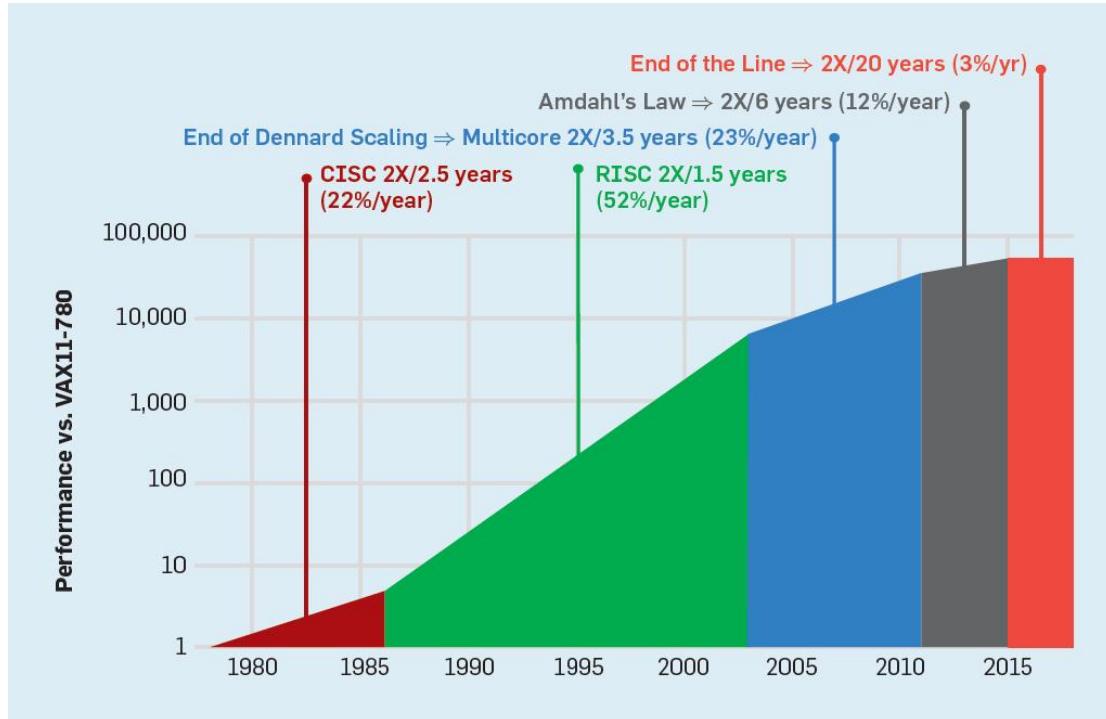
- Moore's Law continues to deliver increases in transistor density
 - But, doubling time is lengthening
- Clock speed scaling failed around 2006
 - No longer possible to ramp the clock speed as process size shrinks
 - Leak currents become important source of power consumption
- So we are basically stuck at $\sim 3\text{GHz}$ clocks from the underlying Wm^{-2} limit
 - This is the *Power Wall*
 - Limits the capabilities of serial processing
- Memory access times are now $\sim 100\text{s}$ of clock cycles
 - Poor data layouts are catastrophic for software performance



Decreasing Returns over Time

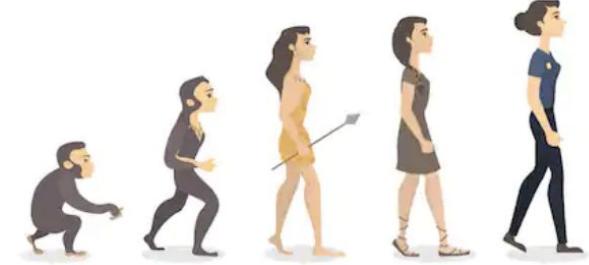
- Conclusion is that diversity of new architectures will only grow
- Best known example is of GPUs

[\[link\]](#)

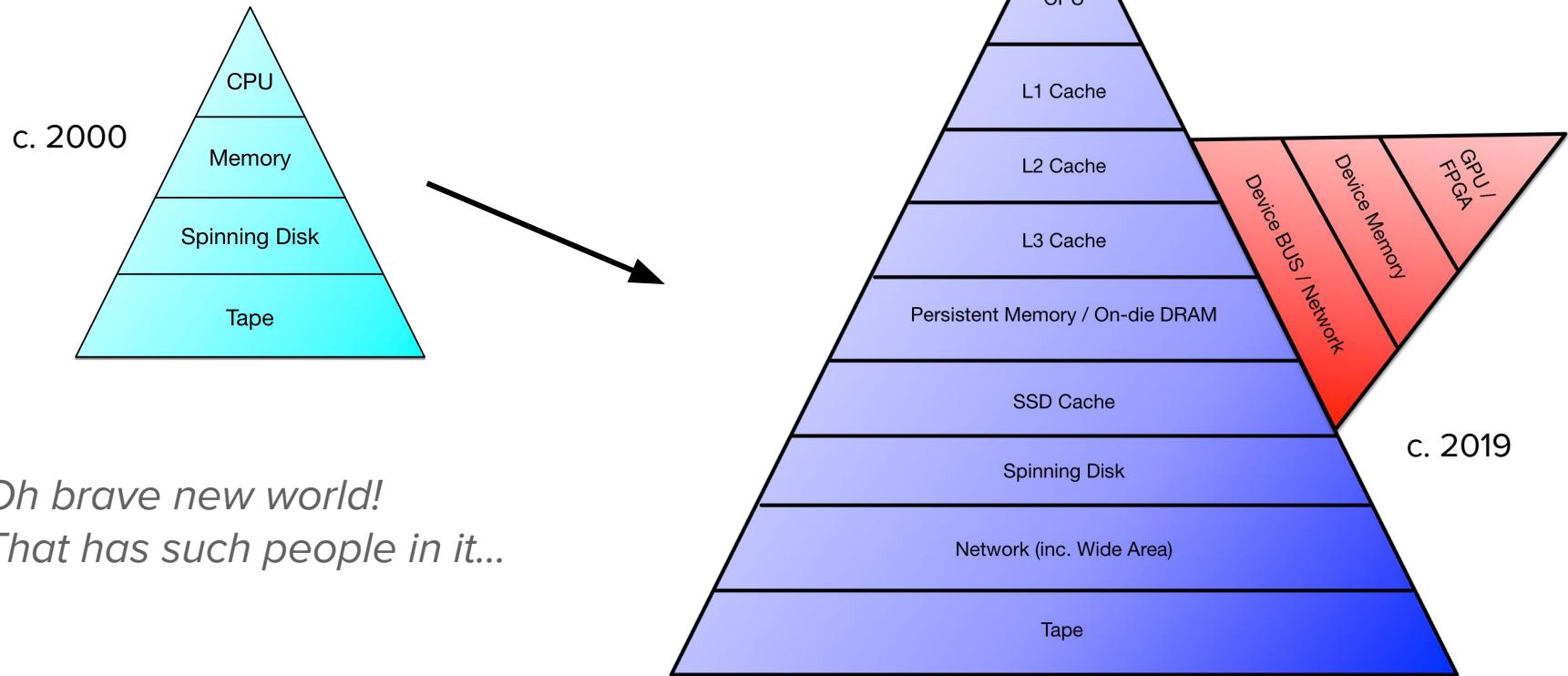


Drivers of Technology Evolution

- Low power devices
 - Driven by mobile technology and Internet of Things
- Data centre processing
 - Extremely large clusters running fairly specialist applications
- Machine learning
 - New silicon devices specialised for training machine learning algorithms, particularly low precision calculations
- Exascale computing
 - Not in itself general purpose, but poses many technical problems whose solutions can be general - HEP pushed to use HPC centres, especially in US
- Energy efficiency is a driver for all of these developments
 - Specialist processors would be designed for very specific tasks
 - Chips would be unable to power all transistors at once: dark silicon is unlit when not used



Hardware Evolution in a Nutshell



Software Challenges and Opportunities

Concurrency

- The one overriding characteristic of modern processor hardware is concurrency
 - SIMD - Single Instruction Multiple Data (a.k.a. vectorisation)
 - Doing exactly the same operation on multiple data objects
 - MIMD - Multiple Instruction Multiple Data (a.k.a. multi-threading or multi-processing)
 - Performing different operations on different data objects, but at the same time
- Because of the inherently parallel nature of HEP processing a lot of concurrency can be exploited at rough granularity
 - Run many jobs from the same task in parallel
 - Run different events from the same job in parallel
- However, the push to highly parallel processing (1000s of GPU cores) requires **parallel algorithms**
 - This often requires completely rethinking problems that had sequential solutions previously, e.g. finding track seeds via cellular automata (TrickTrack library, CMS and FCC)

Heterogeneity

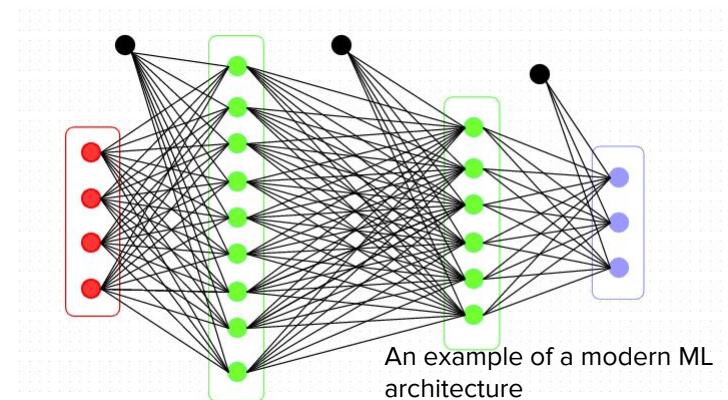
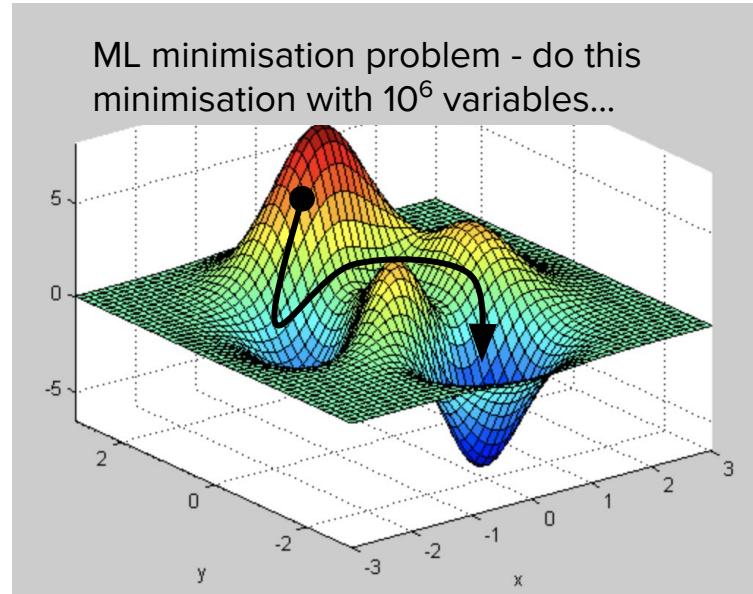
- There are a lot of possible parallel architectures on the market
 - CPUs with multiple cores and wide registers
 - SSE4.2, AVX, AVX2, AVX512, Neon, SVE, Altivec/VMX, VSX
 - GPUs with many cores; FPGAs
 - Nvidia (many generations - often significantly different), AMD, Intel, ...
- In addition there are ‘far out’ architectures proposed, like Intel’s Configurable Spatial Architecture
- Many options for coding, both generic and specific:
 - Cuda, TBB, OpenACC, OpenMP, OpenCL (→ Vulcan), alpaka, Kokkos, ...
- Frustratingly no clear winner, mutually exclusive solutions and many niches
 - One option for now is to isolate the algorithmic code from a ‘wrapper’ that targets a particular device or architecture - approach of ALICE for their GPU/CPU code
 - Hiding details in a lower level library (e.g. VecCore) also helps insulate developers

Data Layout and Throughput

- Original HEP C++ Event Data Models were heavily inspired by the Object Oriented paradigm
 - Deep levels of inheritance
 - Access to data through various indirections
 - Scattered objects in memory
- Lacklustre performance was “hidden by the CPU and we survived LHC start
- In-memory data layout has been improved since then (e.g. ATLAS xAOD)
 - But still hard for the compiler to really figure out what’s going on
 - Function calls non-optimal
 - Extensive use of ‘internal’ EDMs in particular areas, e.g. tracking
- iLCSoft / LCIO also proved that common data models help a lot with common software development
- Want to be flexible re. device transfers and offer different persistency options
 - e.g. ALICE Run3 EDM optimised for message passing and the code generation approaches in FCC-hh
PODIO EDM generator

Machine Learning

- Machine learning, or artificial intelligence, used for many years in HEP
 - Algorithms learn by example (training) how to perform tasks instead of being programmed
- Significant advances in the last years in ‘deep learning’
 - Deep means many neural network layers
 - Fast differentiability and use of GPUs
- Rapid development driven by industry
 - Vibrant ecosystem of tools and techniques
 - *Highly optimised for modern, specialised hardware*



Machine Learning in HEP

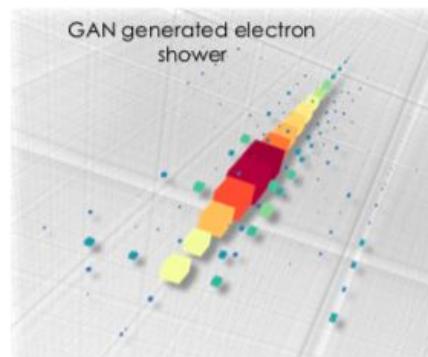
- Better discrimination
 - Important input for analysis (see improvements with Higgs)
 - Also used at HLT as inference can be fast (N.B. training can be slow!)
 - HEP analogies to image recognition or text processing
- Replace expensive calculations with trained output
 - E.g. calorimeter simulations and other complex physical processes
- There are significant opportunities here
 - Need to combine physics and data science knowledge
 - Field evolves rapidly and we need to deepen our expertise
- Integration into our workflows is not at all settled
 - Resource provision, efficient use, heterogeneity and programming models pose problems
 - Training deep models may require *significant* resources

Table 1 | Effect of machine learning on the discovery and study of the Higgs boson

| Analysis | Years of data collection | Sensitivity without machine learning | Sensitivity with machine learning | Ratio of P values | Additional data required |
|---|--------------------------|--------------------------------------|-----------------------------------|---------------------|--------------------------|
| CMS ²⁴ $H \rightarrow \gamma\gamma$ | 2011–2012 | 2.2σ , $P = 0.014$ | 2.7σ , $P = 0.0035$ | 4.0 | 51% |
| ATLAS ⁴³ $H \rightarrow \tau^+\tau^-$ | 2011–2012 | 2.5σ , $P = 0.0062$ | 3.4σ , $P = 0.00034$ | 18 | 85% |
| ATLAS ⁹⁹ $VH \rightarrow bb$ | 2011–2012 | 1.9σ , $P = 0.029$ | 2.5σ , $P = 0.0062$ | 4.7 | 73% |
| ATLAS ⁴¹ $VH \rightarrow bb$ | 2015–2016 | 2.8σ , $P = 0.0026$ | 3.0σ , $P = 0.00135$ | 1.9 | 15% |
| CMS ¹⁰⁰ $VH \rightarrow bb$ | 2011–2012 | 1.4σ , $P = 0.081$ | 2.1σ , $P = 0.018$ | 4.5 | 125% |

Machine learning at the energy and intensity frontiers of particle physics,

<https://doi.org/10.1038/s41586-018-0361-2>



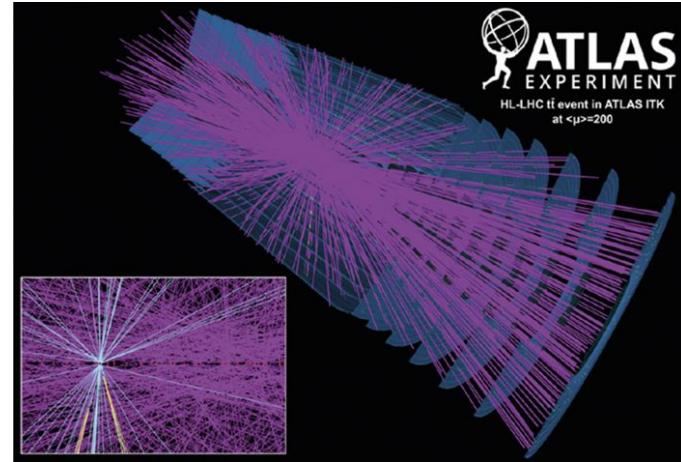
Use of Generative Adversarial Networks to simulate calorimeter showers, trained on G4 events (S. Vallacorsa)

HEP Software and Computing and the HSF Initiative

HEP Software Foundation

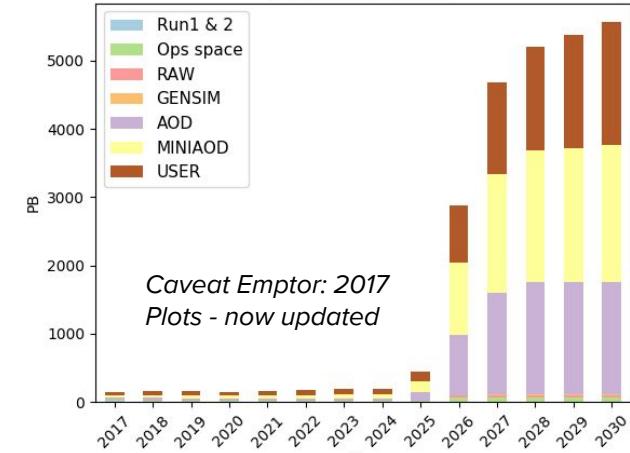
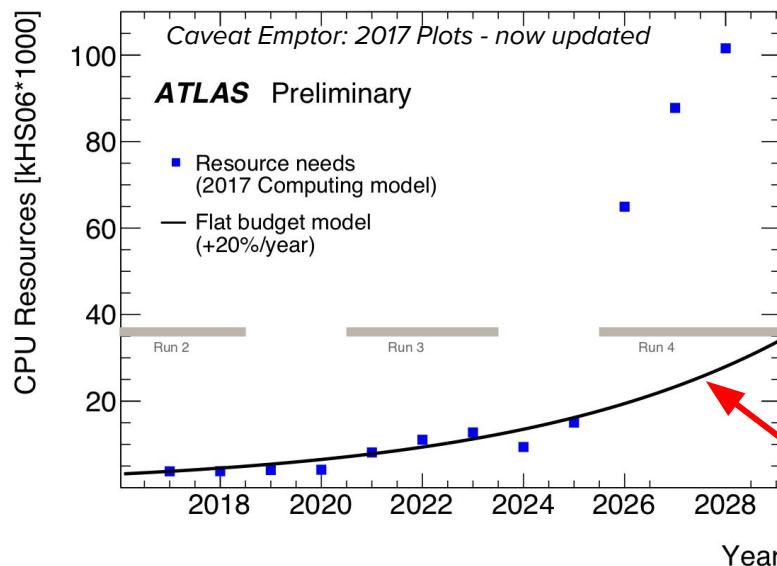
Software at the HL-LHC

- Pile-up of $\sim 200 \Rightarrow$ particularly a challenge for charged particle reconstruction
 - Inner trackers and CMS High Granularity Calorimeter
- HEP software typically executes one instruction at a time (per thread)
 - Since ~ 2013 CPU (core) performance increase is due to more internal parallelism
 - $\times 10$ with the same HW only achievable if using the full potential of processors
 - Major SW re-engineering required (but rewriting everything is not an option)
 - Co-processors like GPUs *require* that this problem is solved
- Increased amount of data requires to revise/evolve our computing and data management approaches
 - We must be able to feed our applications with data efficiently
- *HL-LHC salvation will come from software improvements, not from hardware*



Challenges for the Next Decade

- HL-LHC brings a huge challenge to software and computing
 - Both rate and complexity rise



- Not just a simple extrapolation of Run 2 software and computing
 - Resources needed would hugely exceed those from technology evolution alone

[This is probably too optimistic]

HEP Software Foundation (HSF)



- The LHC experiments, Belle II and DUNE face the same challenges
 - HEP software must evolve to meet these challenges
 - Need to exploit all the expertise available, inside and outside our community, for parallelisation
 - New approaches needed to overcome limitations in today's code
- Cannot afford any more duplicated efforts
 - Each experiment has its own solution for almost everything (framework, reconstruction algorithms, ...)
- HSF started with a number of workshops and working groups on common topics (packaging, licensing)
- The goal of the HSF is to facilitate coordination and common efforts in software and computing across HEP in general
 - Our philosophy is bottom up, a.k.a. *do-ocracy*

Community White Paper Inception

- We wanted to describe a **global vision for software and computing** for the HL-LHC era and HEP in the 2020s
- Formal charge from the WLCG in July 2016
 - Anticipate a "software upgrade" in preparation for HL-LHC
 - Identify and prioritize the software research and development investments
 - i. to achieve improvements in software efficiency, scalability and performance and to make use of the advances in CPU, storage and network technologies
 - ii. to enable new approaches to computing and software that could radically extend the physics reach of the detectors
 - iii. to ensure the long term sustainability of the software through the lifetime of the HL-LHC
- Long process of 1 year, with many working groups and 2 major workshops

A Roadmap for HEP Software and Computing R&D for the 2020s

- 70 page document
- 13 sections summarising R&D in a variety of technical areas for HEP Software and Computing
 - Almost all major domains of HEP Software and Computing are covered
- 1 section on Training and Careers
- 310 authors from 124 institutions
- <https://doi.org/10.1007/s41781-018-0018-8>;
[arXiv:1712.06982](https://arxiv.org/abs/1712.06982)

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```
int main {  
    cout << "write software" << endl;  
    return 0;  
}
```

HSF Working Groups

- The Roadmap established what challenges the community faced
 - But it did not spell out *how* to face them in detail
- HSF had adopted a model of working groups from its earliest days
 - These were open groups of people in the community, motivated enough to organise around a common topic, usually at their own initiative
- This model seemed a good one for moving forwards on the key topics
 - We were a little more formal this time around
 - Call for nominations from the whole community, then search committee
 - Significant engagement from LHC experiments and beyond, e.g. Belle II
- The HSF's role is one of an information conduit and meeting point
 - Report on interesting and common work being done
 - Forum for technical comments and discussion
 - Encourage cooperation across experiments and regions

[[HSF-TN-2016-01](#)]

Some important practical matters!

Copyright and Licensing

M. Jouvin¹ J. Harvey² A. McNab³ E. Sexton-Kennedy⁴ T. Wenaus⁵

¹Laboratoire de l'Accélérateur Linéaire (CNRS) ²CERN ³University of Manchester ⁴Fermi National Accelerator Laboratory ⁵Brookhaven National Laboratory

- Long neglected inside collaborations
 - Code was arbitrarily licensed or unlicensed, copyright assigned to random authors and institutes
 - Yet this is essential to be able to
 - Open source our software properly
 - Combine with other open source projects and collaborate
- Copyright
 - Advice to keep this as low a number as practicable as copyright holders decide the licence
 - LHC experiments: © CERN for the benefit of collaboration X
- License
 - Favour liberal licenses for industry collaboration: LGPL, Apache, MIT
 - Definitely avoid GPL for libraries you want other people to use

Software Nuts and Bolts

[[HSF-TN-2016-03](#)]

L. Sexton-Kennedy¹, B. Hegner², B. Viren³

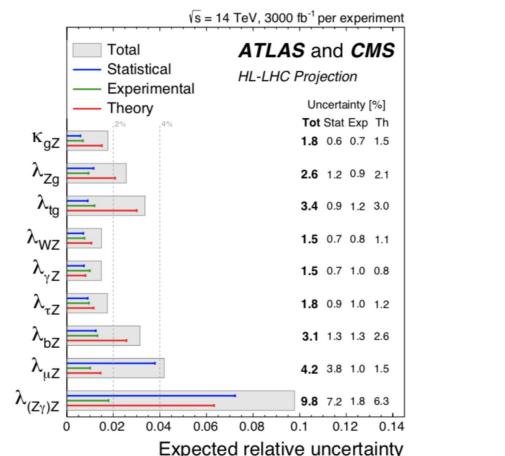
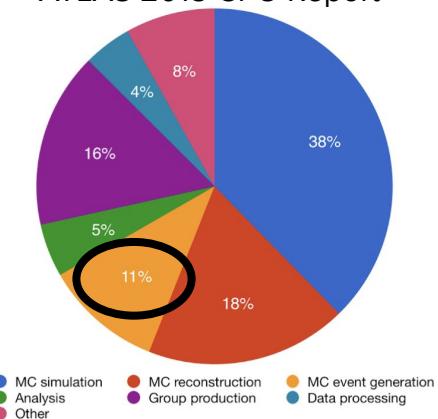
¹FNAL, ²CERN, ³BNL

- Software Tools WG
 - Active group promoting best practice for correctness and performance
 - There has been a revolution in adopting best open source practice in recent years
 - git, GitHub, GitLab, CMake, merge requests, code review, ...
 - HSF has an active group promoting best practice for correctness and performance
 - Profiling, static analysis
- Packaging WG
 - We don't build our experiment software in isolation
 - Need a software stack, incorporating many components from the open source world and HEP community
 - This touches deeply on license and license combinations
 - Preference for tools that are not home grown and have a wider support base
 - Spack (LBNL) and Conda actively being prototyped



Event Generators

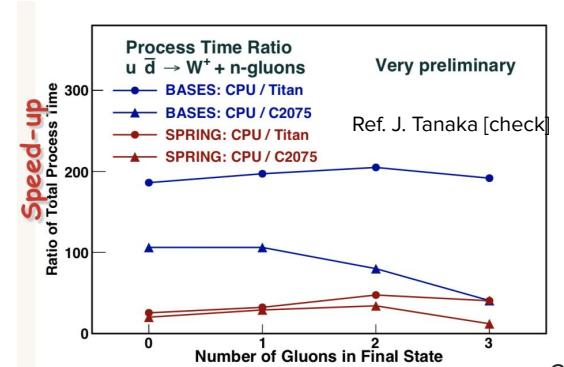
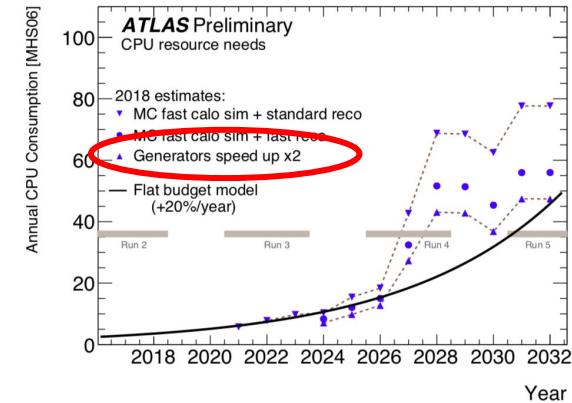
- Event generators are the start of the simulation chain
 - At the LHC Run1 only leading order generators were used
 - Negligible CPU consumption compared with detector simulation - no pressure to optimise
- However, with LHC upgrades coming higher order generators become much more important
 - These are inherently much more costly to run
 - Problems of negative weights can increase hugely the samples needed for weighted event samples
- In addition, the theory community, who develop these codes usually work in small teams
 - Recognition for technical improvements is limited/missing



Many electroweak measurement errors dominated by theory (red). [B. Hinemann](#)

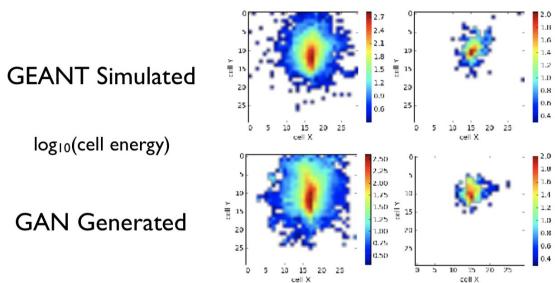
Event Generators - Technical Improvements

- HSF/LPCC workshop in November brought theory and experiment together to look at computing challenges of event generation
 - This was the first workshop of its kind
- Working group tackling technical challenges
 - Setting a baseline for further comparisons
 - Understanding how to run generators for best efficiency
 - Support for technical improvements (e.g. thread safety)
 - Porting to other architectures
 - Could be very suitable code to do this with (smaller, self contained code bases, numerically intensive)
 - e.g. building on the work done so far in MadGraph with GPUs

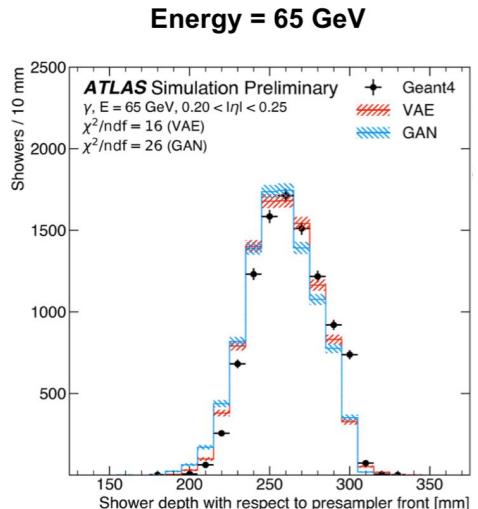


Detector Simulation

- A major consumer of LHC grid resources today
 - Experiments with higher data rates will need to more simulation
- Faster simulation, with no or minimal loss of accuracy, is the goal
 - Range of techniques have been used for a long time (frozen showers, paramtric response)
 - Key point is deciding when it's good enough for physics
- Machine learning lends itself to problems like this
 - Calorimeter simulations usually targeted
 - Variational Auto Encoders (VAEs) attempt to compress the data down to a ‘latent space’ - can be randomly sampled to generate new events
 - Generative Adverserial Networks (GANs) train two networks, one to generate events, the other to try to classify as real/fake
 - R&D on lifecycle integration into Geant4 is starting...



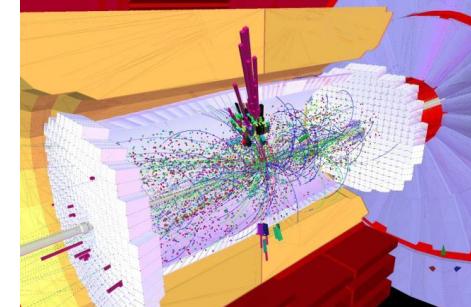
LHCb ECal simulated with G4, generated with GAN [F. Ratnikov]



ATLAS VAE and GAN cf. Geant4 simulation
[ATL-SOFT-PUB-2018-001.]

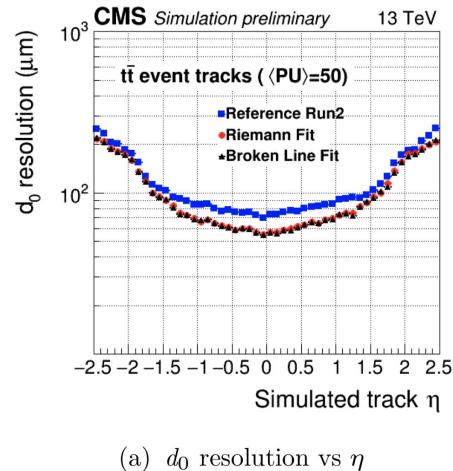
Detector Simulation

- Technical improvement programme helps (and helps everyone)
- GeantV R&D modernises code and introduces vectorisation
 - Speed-ups observed
 - Vectorisation introduces small gains
 - Code modernisation seems to help a lot
- Geant4 now have a new R&D working group that will take studies forward
- Some studies of running Geant4 on GPUs have begun
 - US Exascale Computing Project is funding this
 - Motivated by the next generation of US supercomputers that target exaflop
 - 90-95% of FLOP capacity in GPUs
 - However, migration of physics code is an incredibly tricky business
 - This would be a long haul, but a huge achievement for all of HEP

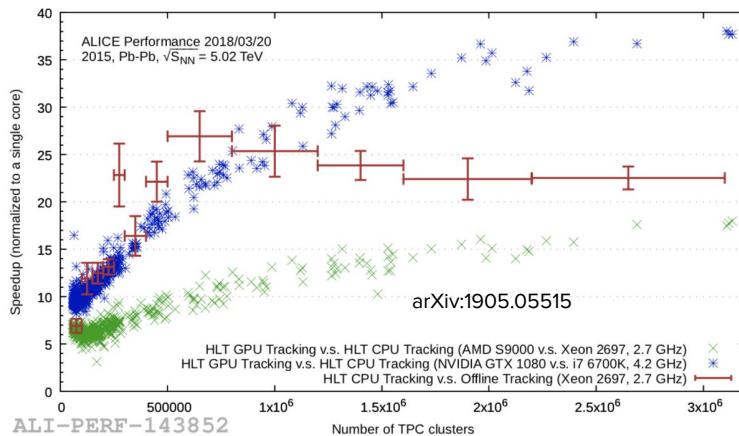


Reconstruction and Software Triggers

- Hardware triggers no longer sufficient for modern experiments
 - More and more initial reconstruction needs to happen in software
- Close to the machine, need to deal with tremendous rates and get sufficient discrimination
 - Pressure to break with legacy code is high
 - Lots of experimentation with rewriting code for GPUs
 - E.g. LHCb's Allen project (HLT1 on GPU)
 - ALICE have ported a lot of reconstruction to GPUs and also improved the algorithms a lot
 - CMS Patatrack project has improved physics performance as well
 - Revisiting old code helps!
- Lessons learned keep data model simple, bulk data, be asynchronous, minimise data transfers



(a) d_0 resolution vs η



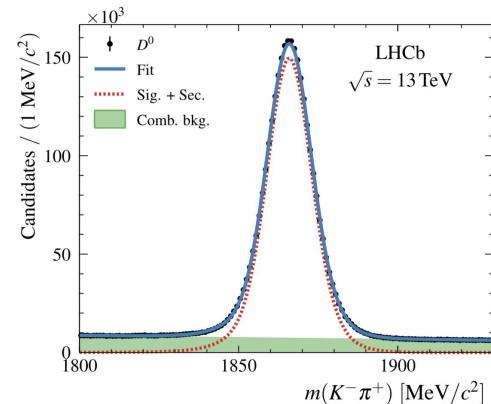
arXiv:1905.05515

Reconstruction and Software Triggers

- Real Time Analysis (HEP Version)
 - Design a system that can produce analysis useful outputs as part of the trigger decision
 - If this captures the most useful information from the event, can dispense with raw information
 - *This is a way to fit more physics into the budget*
- LHCb Turbo Stream has been introduced in Run2 and will be dominant in Run3
- Whole ALICE data reduction scheme is based around keeping ‘useful’ parts of events (no more binary trigger)
 - O2 → Online/Offline Data Reduction Farm
- ATLAS and CMS have schemes under development for special handling of samples for which full raw data is unaffordable (aka. data scouting)

| Persistence method | Average event size (kB) |
|-----------------------|-------------------------|
| Turbo | 7 |
| Selective persistence | 16 |
| Complete persistence | 48 |
| Raw event | 69 |

LHCb Run2 Turbo took 25% of events for only 10% of bandwidth



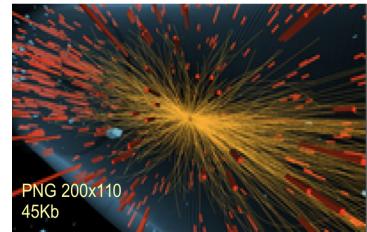
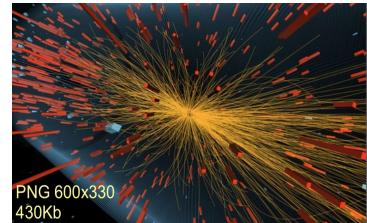
LHCb charm physics analysis using Turbo Stream (arXiv:1510.01707)



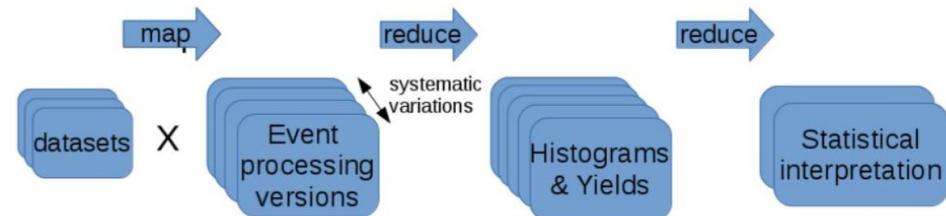
ANALYSIS FACILITIES
Dedicated and dense,
do more with less: aim
at > 95% efficiency

Analysis

- Scaling for analysis level data also a huge challenge for all LHC experiments
- Efficient use of analysis data can come with combining many analyses as carriages in a train like model (pioneered by PHENIX and then ALICE)
 - Also goes well with techniques like tape carousels (ATLAS scheme for rotating primary AOD data from tape systems into a disk buffer)
 - Interest in *analysis clusters*, specialised for analysis operations over the generic grid resources (WLCG/HSF pre-CHEP workshop 2-3 November)
- Reducing volume of data needed helps hugely
 - CMS ~1kB nanoAOD makes a vast difference to analysis efficiency and “papers per petabyte”
 - Smaller EDM is easier to make efficient
 - Requires analyst agreement on corrections, scale factors, etc.
 - However the alternative is perhaps that your analysis never gets done



Analysis



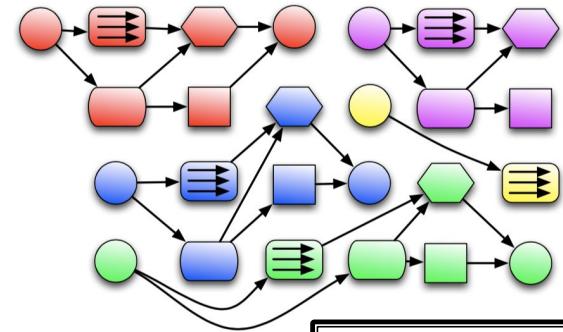
- Improve analysis ergonomics - how the user interacts with the system to express their analysis
 - Streamline common tasks
 - Handle all input datasets; Corrections and systematics
 - Compute per event and accumulate; Statistical interpretations
 - Declarative models, building on ROOT's RDataFrame
 - Say *what*, not *how* and let the backend optimise
 - E.g. split and merge, GPU execution
- Notebook like interfaces gain ground, as do containers - lots of high level Python
- Interest in data science tools and machine learning is significant for this community - inspiring new approaches (e.g. uproot, awkward array, scikit-hep, Coffea)
 - This is an ecosystem into which HEP can contribute

```
# * Jet select/cleaning against loose leptons , jet pt > 25 , jet id
flow.DefaultConfig(jetPtCut=25,jetIdCut=0,jetPUIdCut=0)
flow.SubCollection("CleanJet","Jet","""
    Jet_pt > jetPtCut &&
    Jet_jetId > jetIdCut &&
    Jet_puId > jetPUIdCut &&
    (Jet_LeptonIdx== -1 || Jet_LeptonDr > 0.3)
""")
```

A. Rizzi, NAIL prototype

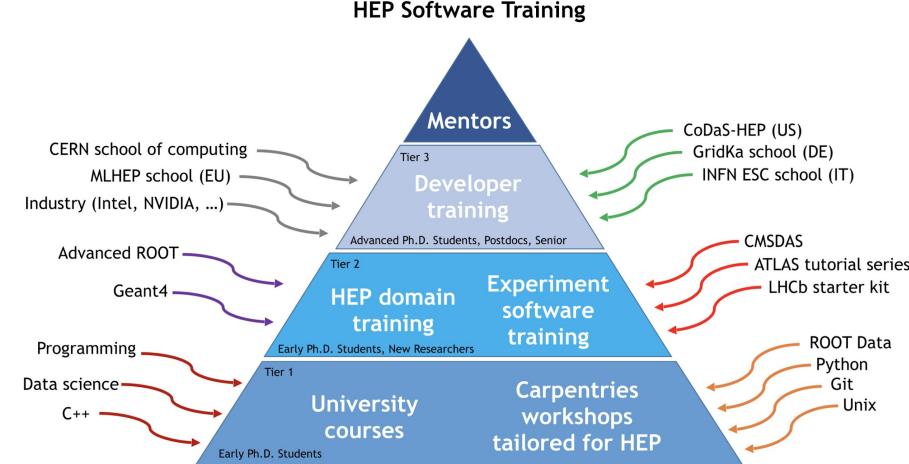
Frameworks and Integration

- Increasingly heterogeneous world requires advanced software support infrastructure
 - Software frameworks support use of different devices as well as insulate developers from many of the details of concurrency and threading models
 - Adapt to the new heterogeneous landscape
 - Latency hiding is critical to maintaining throughout
 - Framework development has traditionally been quite fragmented, but new experiments should offer a chance to increase convergence
 - Better to start off together than try to re-converge later (iLCSoft, LArSoft examples of success, albeit without concurrency; Gaudi for LHCb, ATLAS)
 - ALFA for ALICE and FAIR experiments
- New HSF working group being established now ([draft mandate](#))



Training and Careers

- Many new skills are needed for today's software developers and users
- Base has relatively uniform demands
 - Any common components help us
- LHCb StarterKit initiative taken up by several experiments, sharing training material
 - Links to 'Carpentries' being remade (US training projects) - up the level!
- New areas of challenge
 - Concurrency, accelerators, data science
 - Need to foster new C++ expertise (unlikely to be replaced soon as our core language, but needs to be modernised)
- Careers area for HEP software experts is an area of great concern
 - Need a functioning career path that retains skills and rewards passing them on
 - Recognition that software is a key part of HEP now

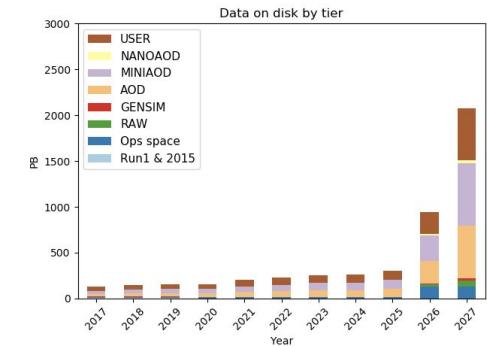
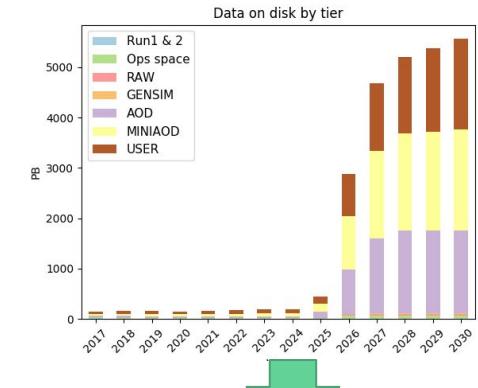
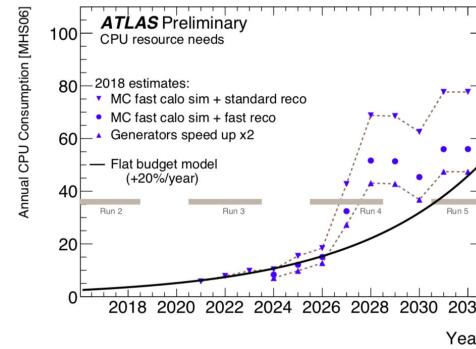
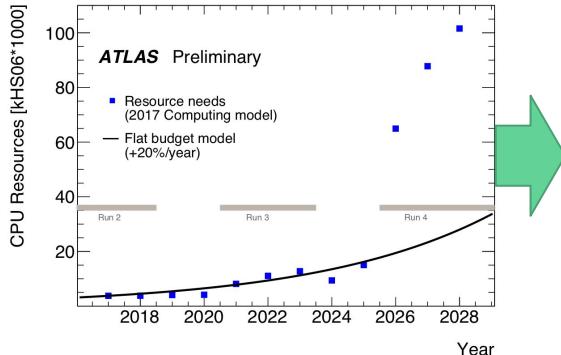


Instrumentation and Computing

Xinchou Lou, Brigitte Vachon
Scientific Secretaries: Emilia Leogrande, Rogers Jones

Meeting the HL-LHC Challenge!

- Already since the Roadmap was written experiments have made great progress in meeting the HL-LHC challenge
 - Bad software, is extremely expensive
 - Good and clever software allows much more physics to fit in the budget



Conclusions

- We have a wide ranging and ambitious physics programme in HEP and in associated disciplines
 - Our experiments are highly data intensive and require high quality software and computing
- The landscape for software is becoming ever more challenging
 - Working together on common problems is not only the best use of our resources, our funding agencies will mandate it
- HSF is now established to help HEP achieve that goal and marshalls effort around the community
 - Roadmap delivered and active working groups in key areas



HL-LHC is a challenge and also a great opportunity to improve HEP software

HSF Getting Involved...

- Join the HSF Forum, hsf-forum@gmail.com
 - Few messages a week with updates, jobs, items of interest
 - Owned by the community - please just post items of relevance
 - Join a working group, https://hepsoftwarefoundation.org/what_are_WGs.html
 - Follow the group's meetings and discussions
 - Suggest a meeting topic
 - Annual meetings and Workshops
 - Establishing a tradition of a joint meeting with WLCG each Year (next short meeting pre-CHEP, November)
 - Propose a new activity area
 - The HSF is there to help gather interest
- Data Analysis
 - Detector Simulation
 - Frameworks
 - Physics Generators
 - Packaging
 - PyHEP - Python in HEP
 - Quantum Computing
 - Reconstruction and Software Triggers
 - Software/Developer Tools
 - Training
 - Visualization