

#### What is scientific computing?

#### Computational science

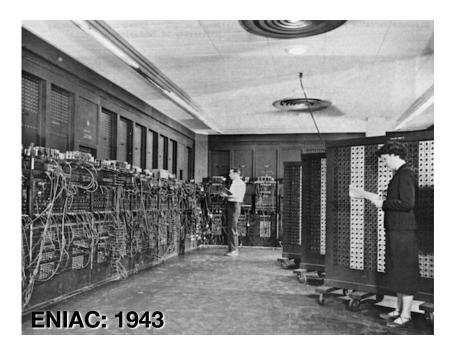
From Wikipedia, the free encyclopedia

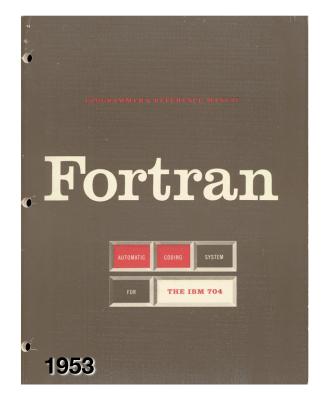
Not to be confused with computer science.

Computational science (also scientific computing or scientific computation (SC)) is a rapidly growing multidisciplinary field that uses advanced computing capabilities to understand and solve complex problems. It is an area of science which spans many disciplines, but at its core it involves the development of models and simulations to understand natural systems.

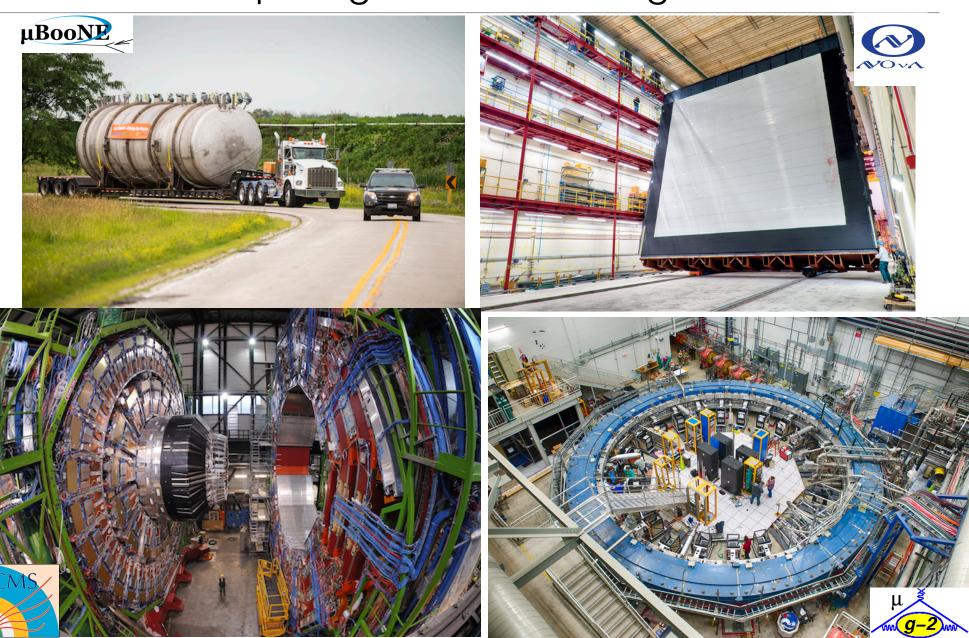
- Algorithms (numerical and non-numerical): mathematical models, computational models, and computer simulations developed to solve science (e.g., biological, physical, and social), engineering, and humanities problems
- Computer and information science that develops and optimizes the advanced system hardware, software, networking, and data management components needed to solve computationally demanding problems
- . The computing infrastructure that supports both the science and engineering problem solving and the developmental computer and information science

In practical use, it is typically the application of computer simulation and other forms of computation from numerical analysis and theoretical computer science to solve problems in various scientific disciplines. The field is different from theory and laboratory experiment which are the traditional forms of science and engineering. The scientific

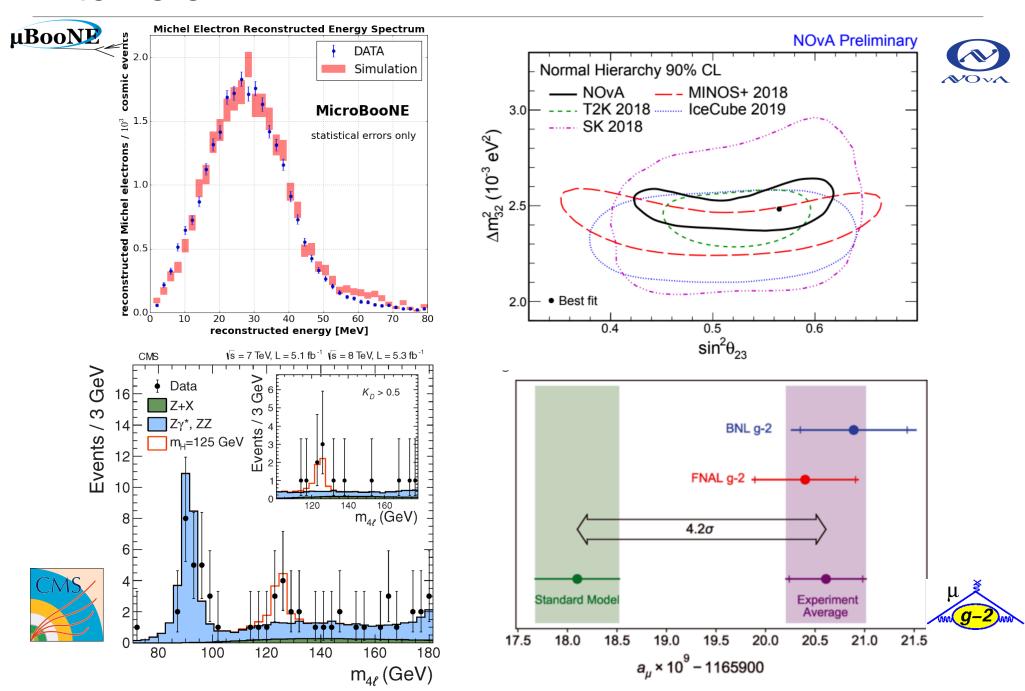




# Scientific Computing in HEP: Getting from here...

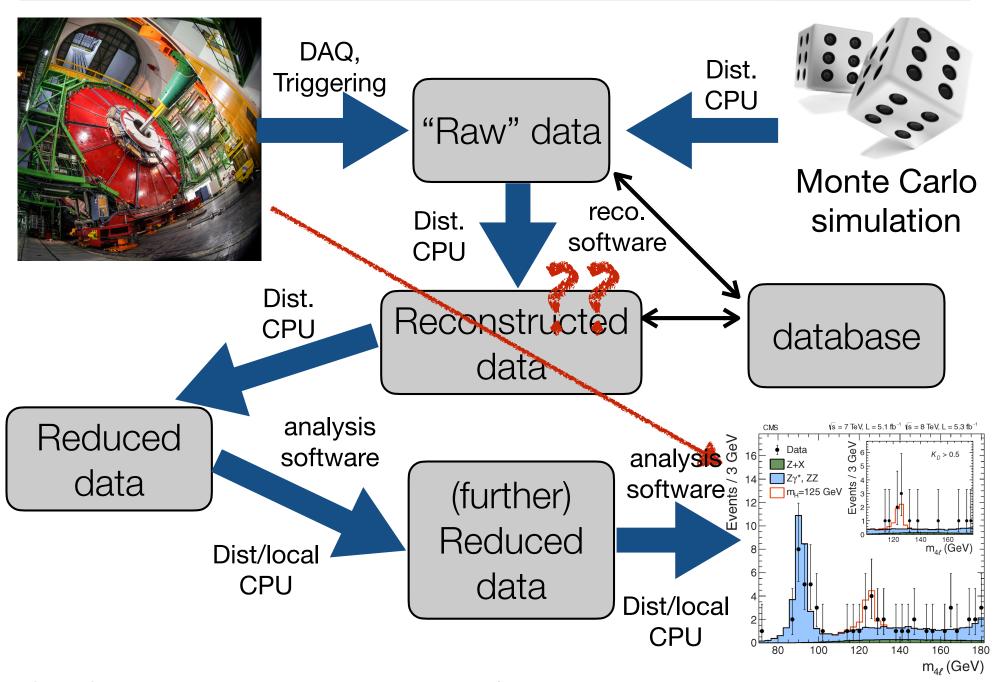


#### ...to here

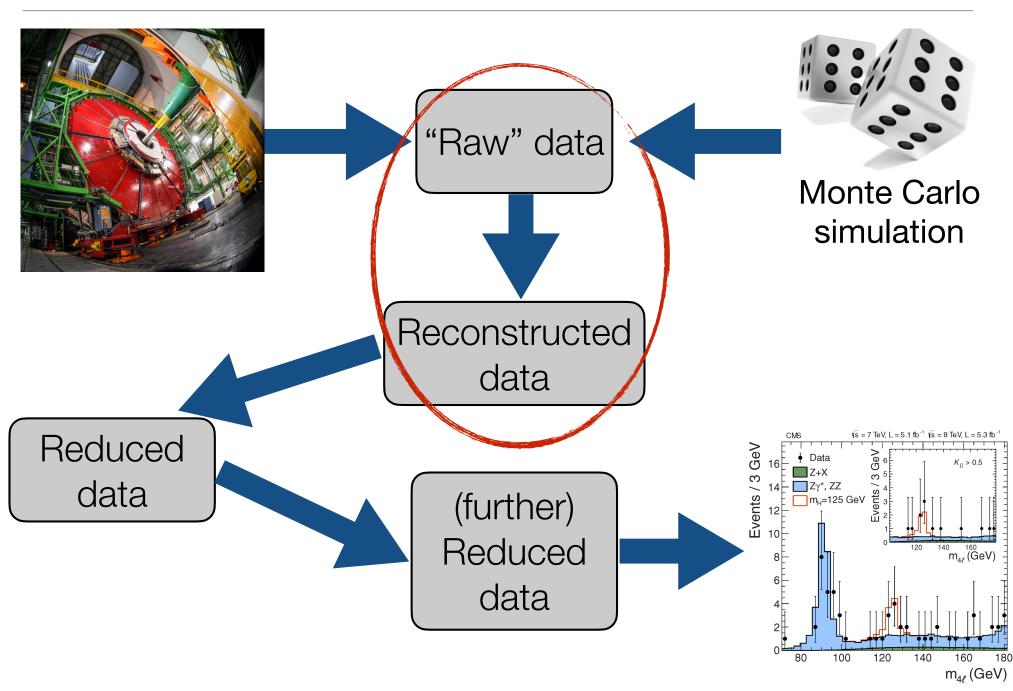


## Software

#### The process

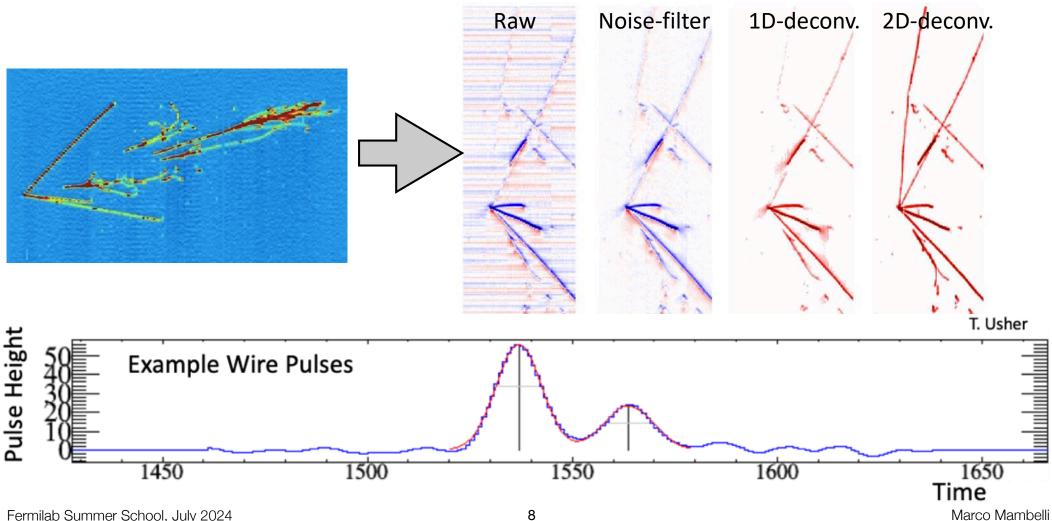


#### Reconstruction

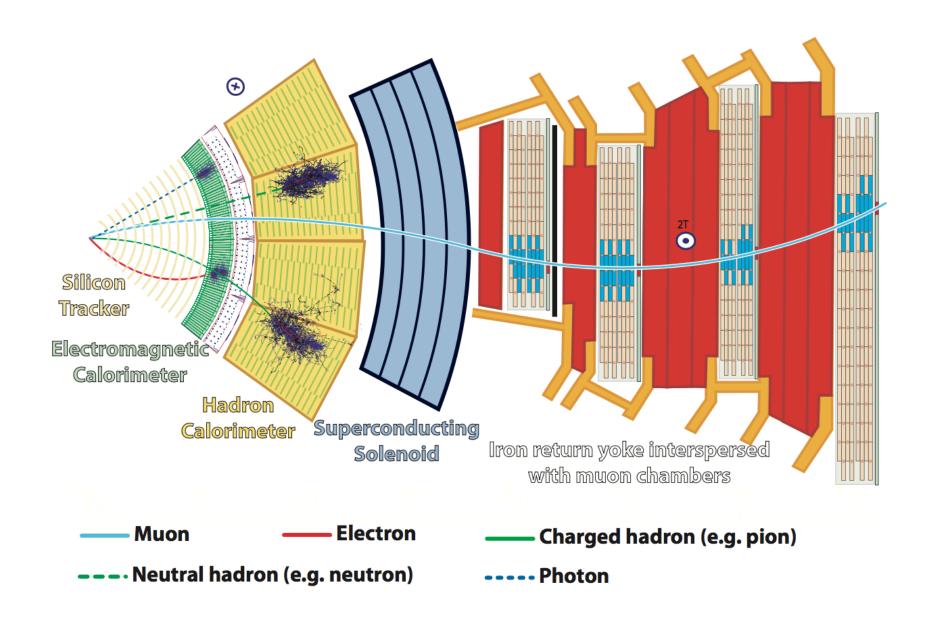


## Reconstruction: LArSoft (MicroBooNE)

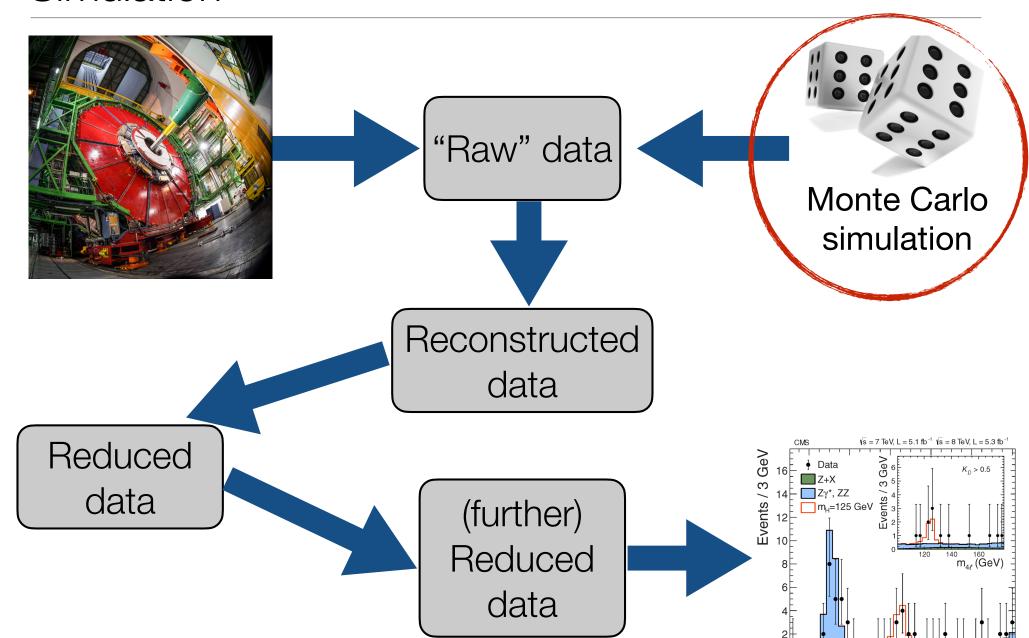
- LArSoft, developed at Fermilab, is a common code base for LAr neutrino detectors
  - Used by MicroBooNE, SBND, ICARUS, DUNE
  - Includes tools for both reconstruction and data analysis



#### Reconstruction: CMS



#### Simulation



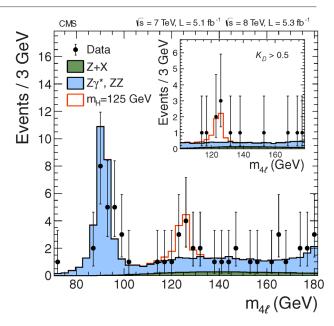
160 180 m<sub>4ℓ</sub> (GeV)

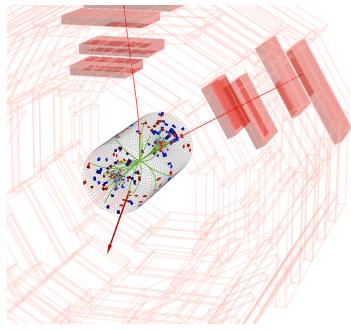
100

120

#### Simulation

- Simulated events are crucial to science at HEP experiments
  - Rely on Monte Carlo Simulation to produce these events
- Event generators simulate the underlying particle interaction of interest
- Resulting interaction event is then fed to a detector simulator
  - Consider the material and geometry of every part of the detector
  - Simulate how particles from interaction and decay would propagate
  - Most detector simulations use GEANT
    - Also used in nuclear and accelerator physics as well as medical and space science





# Analysis



"Raw" data

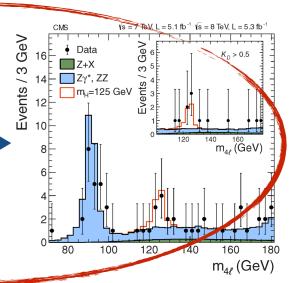


Monte Carlo simulation

Reconstructed data

Reduced data

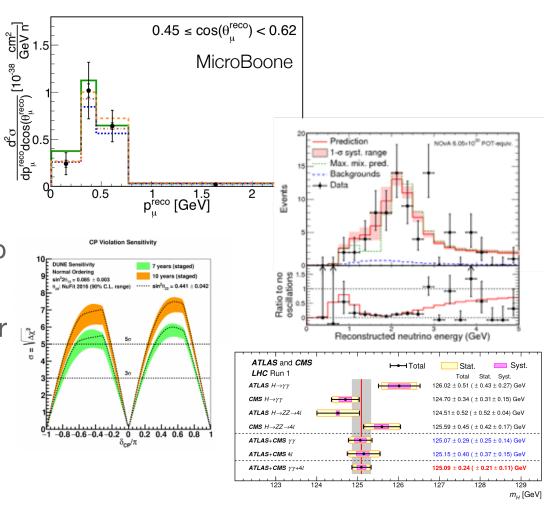
(further) Reduced data



#### Analysis

- Getting from data events of interest to plots, tables, and numbers
  - This is the computing step nearly all HEP experimentalists are familiar with
- Common tools are needed
  - Mathematical functions
  - Statistical analysis
  - Plotting/histogramming
- Nearly all HEP experiments use the ROOT framework
  - Developed by CERN and Fermilab
  - C++ (object oriented)
  - Couples with code written in other languages (e.g. Python)
  - https://pos.sissa.it/093/002/pdf





#### Preservation and Portability

- Experiments and data lifetime
  - Can stretch over OS upgrades and hardware evolution
  - Survives developers or software projects
  - Limited upgrade possibilities during data taking
  - Limited effort at the end of lifetime



- C++, Python, scikit
- CI/CD
- Capturing the whole environment
  - Containers: Docker, Apptainer
  - RFANA



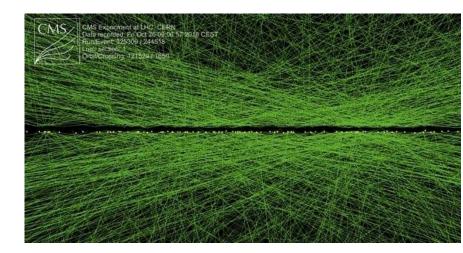


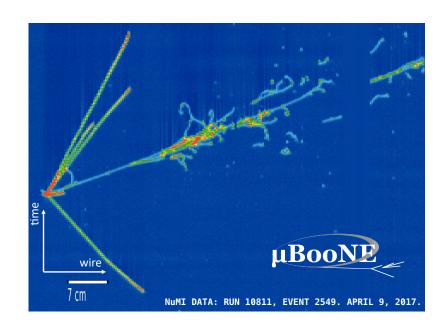


Hardware (CPUs, Storage and Networks)

#### CPU

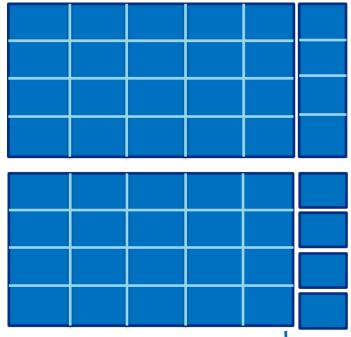
- Reconstruction and simulation of events are biggest CPU drivers
- Such computing is "pleasantly parallel"
  - Processing one event is completely independent of processing any other
  - Relatively short processing times but many events and growing complexity
- CMS typical collider experiment
  - ~30 s/event (~30x more in a few years!)
  - ~billions events (simulated+collision)/ month
- MicroBooNE liquid Argon (LAr) neutrino experiment
  - ~1-2 min/event
  - ~million events (simulated+beam)/month
  - 1 event in DUNE will have ~50x more channels (!)





#### Divide et Impera





- Split the complex problem
- Solve the parts (jobs)
- Get the overall solution

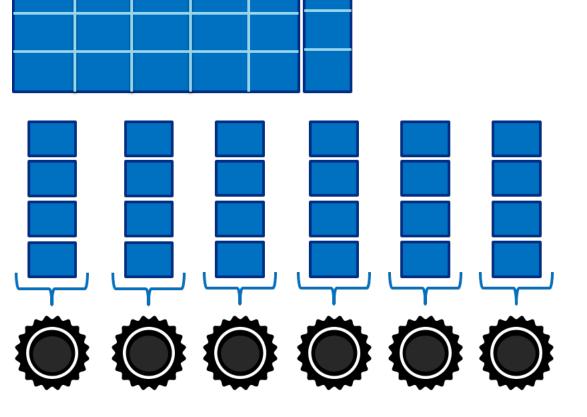


#### Divide et Impera



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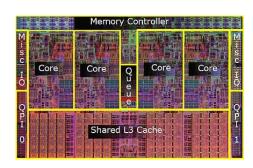


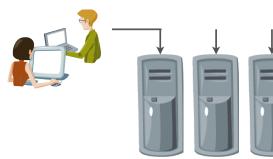
#### Get it fast!

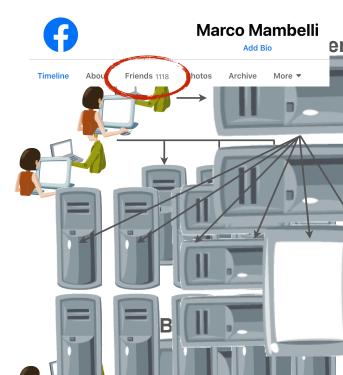
High Throughput Computing (HTC)

#### Divide et Impera: 1 Billion Events

- Use one computer for the CMS events?
  - CPU: Central Processing Unit (typically whole chip)
    - Between 2 and 64 individual cores
    - Each core can process one instruction at a time 30s/8 cores\*18 events ~120 years
- Use your friends' computers as well
  - To get 1B events in one month, we require 1,440
    8-core computers
    - We are almost there with the friends!
  - Your software and data would need to get to each of those computers
  - You'd need to collect output from each
  - And you'd need user accounts on all of them
- Need to find an easier way to get from one computer to many!









- Batch systems
  - Single entry point (queue for many users)
  - Jobs to any available slot
  - Output from each job handled in the same way
  - Batch system can handle user authentication on each individual computer

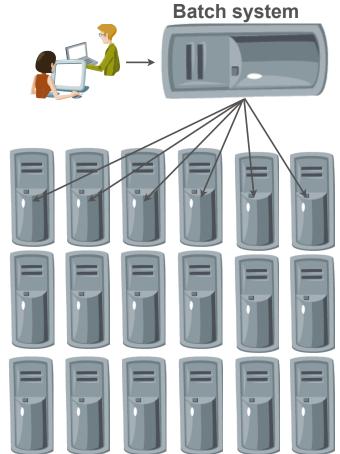




Fermigrid (~20k CPU cores)

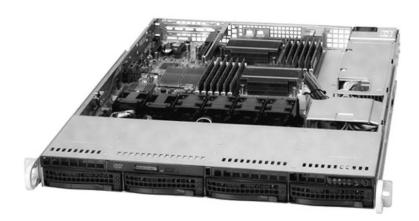
US CMS Tier1 (~20k CPU cores)

CMS LPC (~5k CPU cores)



09. January 2017 30s/20,000 cores\*1B events ~17 days

#### Inside a Data Center: Computing Racks



Rack-mounted units "pizza boxes" dual multi-core processors, large amount of memory, system and small local disk Newest units with 64-core processors, 256 GB of memory, 100 Gb/s network All running Linux!

Hot and Cold rows
Limited by power and cooling
GCC has a power capacity of ~ 2.5
Megawatts





#### The Two Fermilab Data Centers



Feynman Computing Center (FCC)

Two "logical" data centers (FCC2, FCC3)
High Availability, each has own
generator / UPS system
Each a "corner" of a redundant network

#### GRID Computing Center (GCC)

Two "logical" data centers:

- •Room A, Tape Room, Network Room A
- •Rooms B, C, Network Room B

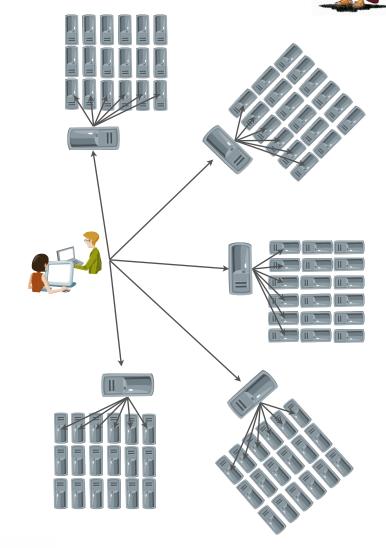
"Lights out" facility, with UPS systems, only good for minutes

Each a "corner" of a redundant network



#### Divide et Impera: the Grid

- Many batch systems accessible from one point: a computing grid
  - grid sites (batch systems) from universities and labs across the world into one grid
  - Distributed high-throughput computing (DHTC)
- Analogy: utility grids
- Delegated and Federated Trust Model
  - Use tokens (JWT) or grid certificates
     (x509) and Virtual Organization (VO)
  - A certificate is an encrypted "signature" that verifies you belong to an organization (e.g., a collaboration like NOvA)
  - Each site decides which VOs to trust









09. January 2017

## Overlapping grids



## Overlapping grids

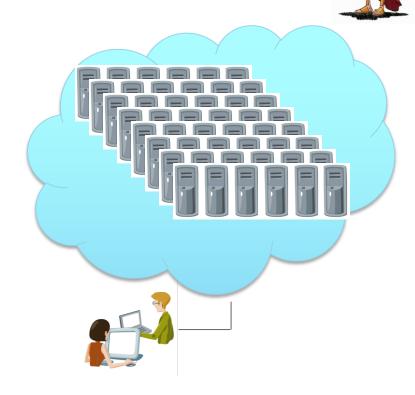
#### World LHC Computing Grid



200+ sites, 2M+ CPU cores

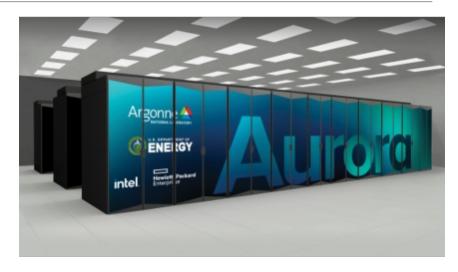
#### Divide et Impera: the Cloud

- And when you still don't have enough, then you can rent it
  - Commercial Clouds like AWS (Amazon), GCE (Google) and Azure (Microsoft) can rent you a seemly endless amount of computing power
  - Elastic computing: expands at will
- Problem: irregular use pattern
- Solution: burst-out by renting resources on the Cloud for peak usage
- More expensive than local resources
- Difficult to justify non-capital expenditures

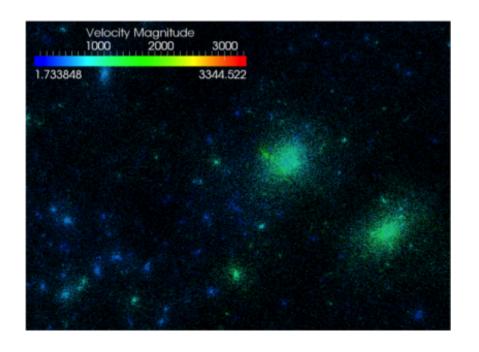


## High Performance Computing

- Much scientific computing outside of experimental HEP is not "pleasantly parallel"
- Better platform: High Performance
   Computing (HPC)
  - Batch systems where individual computers are interconnected via high-speed links
  - Large HPC systems are often called "supercomputers"
- Fermilab has 5 HPC clusters
  - Total of 18.5k CPU cores
  - Used for Lattice QCD calculations, accelerator modeling, large-scale astrophysical simulations and testing
- Used also for event processing by splitting it into pieces

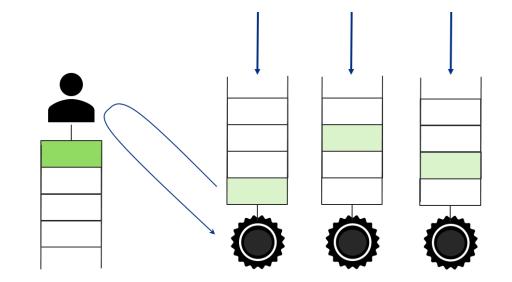


Aurora@Argonne: 20k CPUs 60k GPUs, 10PB memory, 2 DP exaflops

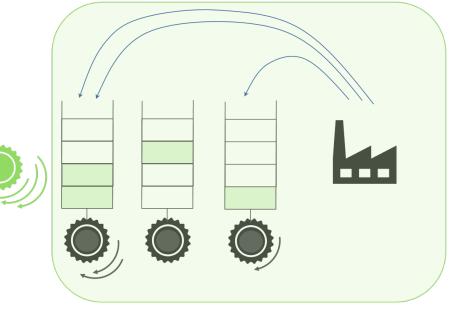


## Simplify access: Pilot jobs

- Late binding
- Separation of tasks
  - Pilot jobs
    - Test
    - Set up
    - Wait in the queue
  - User jobs
    - Science



- A Factory will submit pilot jobs for you
  - Use resources as available (pressure)
  - Separation of knowledge



## Putting it all together: GlideinWMS

**Factory** 

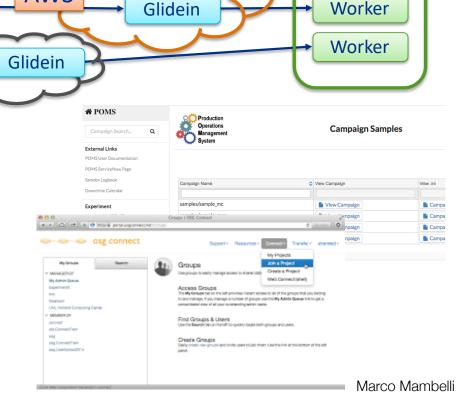
 The Glidein Workload Management System is a pilot based resource provisioning tool for Distributed High Throughput Computing

 Provides reliable and uniform virtual clusters, the global pool

 Submits Glideins to unreliable heterogeneous resources

 Knows "how to talk" to all the different systems

- Multiple Frontends and Factories work together to provide High Availability
- Used by: CMS, DUNE, FIFE, OSG, (POMS, Jobsub, OSG-Connect)



Job Queue

Worker

Worker

**Frontend** 

Cluster

**AWS** 

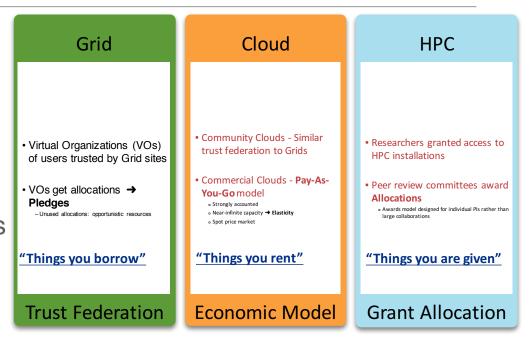
Glidein

Glidein

CE

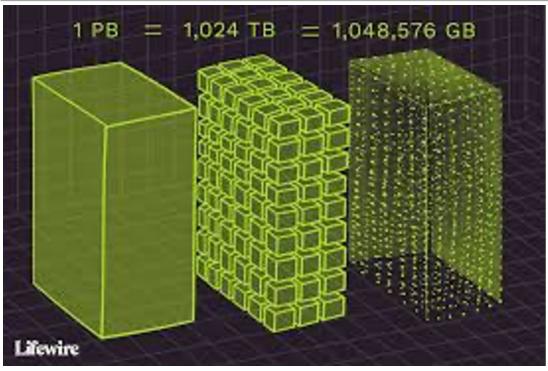
#### **HEPCloud**

- Making it all possible from one place:
   HEPCloud
- Unified interface to Grid, Cloud, and HPC resources
- Currently used mainly to run CMS workflows on NERSC supercomputers
- Optimized bidding for AWS spot pricing
- Better handling of heterogeneous resources (GPUs, QPUs)
- In 2021 doubled CMS Tier 1 capacity using NERSC and other facilities, 160M CoreHrs
- Simulated 1 billion events in 48 hours

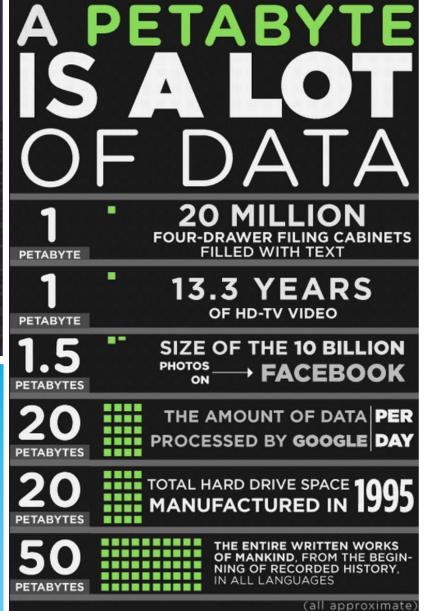




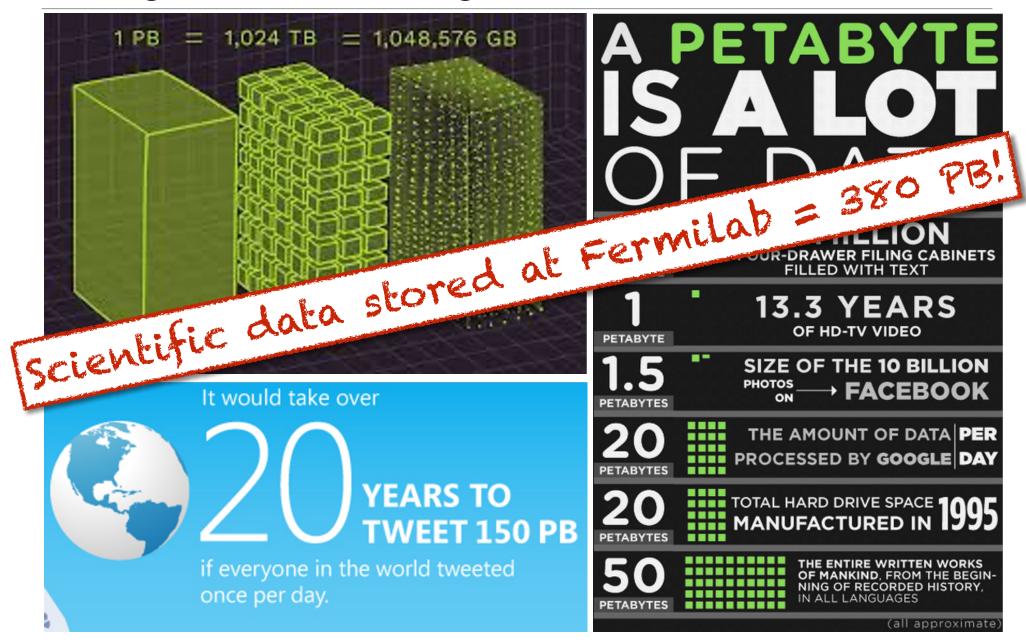
#### Storage: understanding units



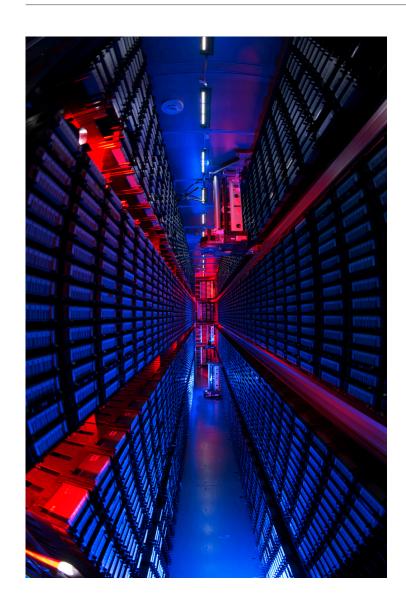




## Storage: understanding units



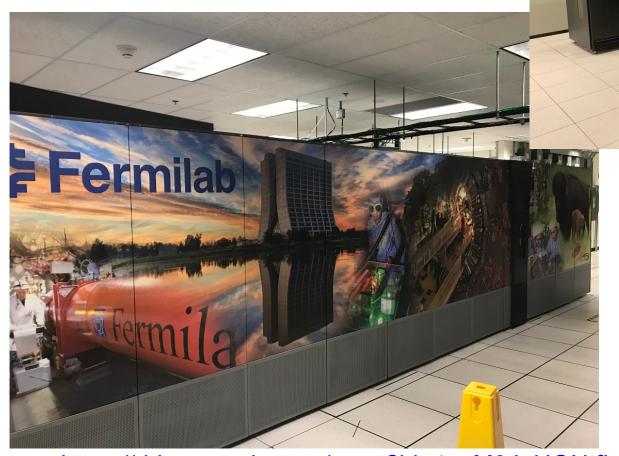
#### Storage: tape



- Primary storage medium for scientific data at Fermilab: magnetic tape
- Still the most efficient way to store petabytes of data if:
  - Not all of it is accessed at the same time
  - Access patterns are fairly linear
  - Sufficient disk for staging
- Fermilab has seven robotic tape libraries
  - Each library can hold up to 10,000 tapes
- Current tapes hold ~12TB of data each (1km of tape!)
- Total active on tape: 300 PB
  - CMS, 89.69
  - NOVA, 53.65
  - uBoone, 28.48
  - gm2, 17.51
  - DUNE, 15.06

## Storage: tape libraries

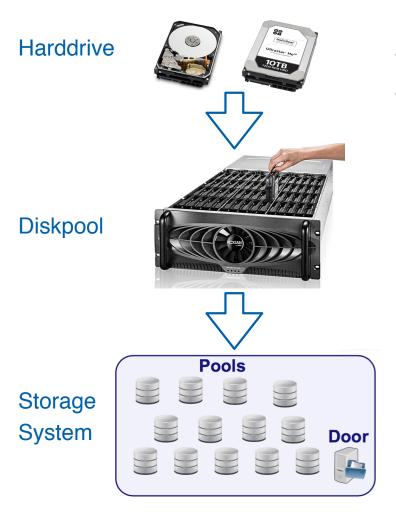
- 9000 to 10000 tapes per library
- A few dozen drives
- Fire suppression system





https://drive.google.com/open?id=1caM9JoUONtflgqxrcJBBN5OImGrpnYy5

## Storage: disk

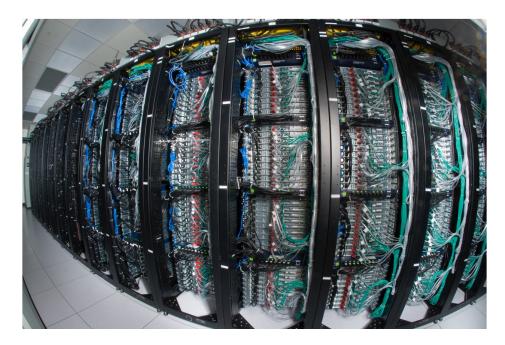


Oliver Gutsche I Software, Computing & Analysis - CMS DAS January 2017

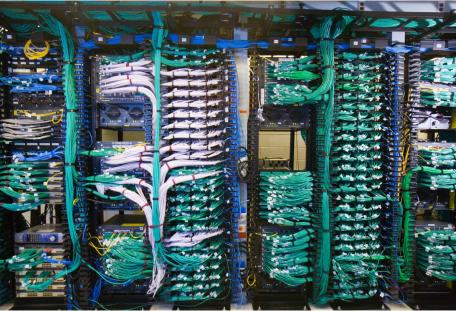
- ~80PB of disk (hard drive) storage
  - Most used as staging area from tape
- Disks are organized into pools
- Software allows collections of pools to appear to a user as a single storage device
  - Fermilab uses a system called "dCache"
  - In a typical week, data throughput in the Fermilab dCache pools average 30GB/s



# Networking: Local Networks







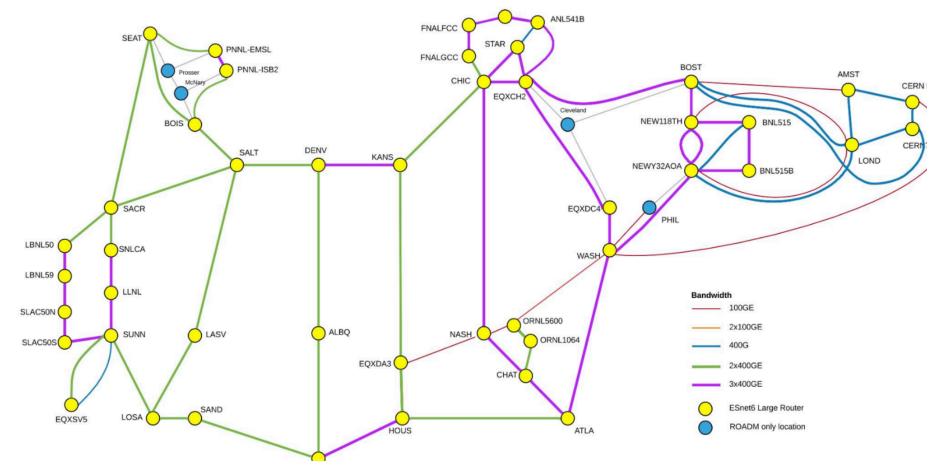


#### Networking: Wide Area Networks

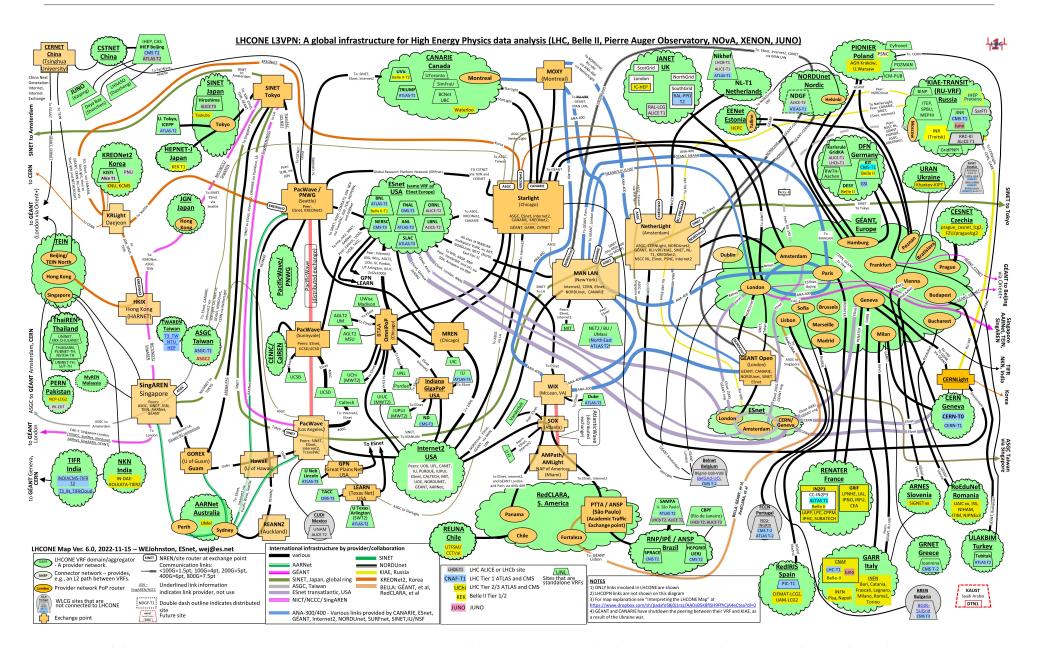
- Distributed computing requires
   fast and reliable networks
- Dedicated fiber optic links
   (typically 400 Gbps) connect

   experiments and major labs

- e.g. ESNet at national labs, GÉANT in Europe
- Dedicated LHC links connect these together (LHCONE, LHCOPN)



#### Networking

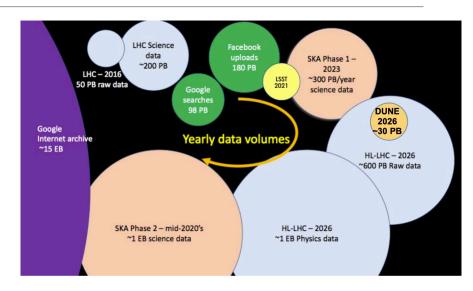


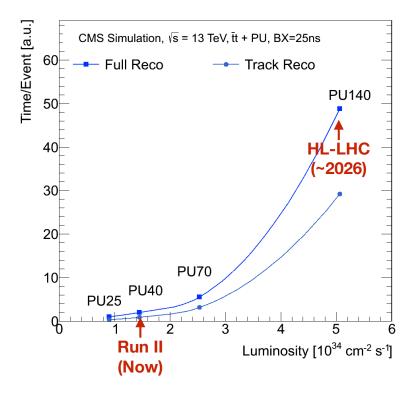
 $LHCONE-intermediate\ detail\ from\ William\ Johnston-https://www.dropbox.com/sh/padxfo58j0j1raz/AADsB5K8flSH9FfhCjA4eCtea?dl=0.$ 

## The Future

## Explosion of data and ever-growing need for CPU

- HL-LHC, DUNE, LSST, SKA will produce up to exabytes of data per year
- More than one order of magnitude above current dataset sizes
- And more complex data to allow precision measurements
- Growing dataset size and event complexity = more computing!
  - If we scale current algorithms, the CPU needs of LHC experiments will grow by a factor of 30 in a decade
  - Similar issues faced by newer, bigger LAr experiments such as DUNE

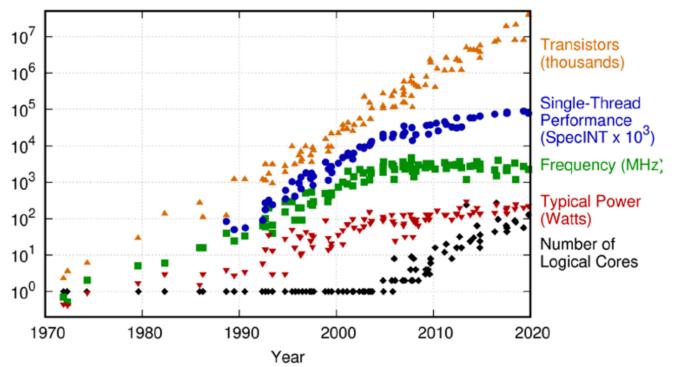




#### Moore's Law probably won't help

- Moore's Law: Number of transistors on a chip doubles every 2 years
- That's still true, but single-thread performance has stopped increasing
- Instead, number of cores is now dramatically increasing

#### 48 Years of Microprocessor Trend Data



Original data up to the year 2010 collected and plotted by M. Horowitz, F. Labonte, O. Shacham, K. Olukotun, L. Hammond, and C. Batten New plot and data collected for 2010-2019 by K. Rupp

#### To take advantage of these:



Intel Many Integrate Core (MIC) CPU



**NVIDIA GPU** 

We need to rewrite a lot of software!

## What about supercomputers?

#### ASCR\* Computing Upgrades At a Glance

System attributes	NERSC Now	OLCF Now	ALCF Now	NERSC Upgrade	ALCF Upgrade	OLCF Upgrade	ALCF Upgrade
Name	Cori	Summit	Theta	Perlmutter	Polaris	Frontier	Aurora
Installation (planned or actual)	2016	2018	2017	2021	2021	2021-2	2022-3
System peak (PF)	30	200	12	>120	35-45	1500	> 1000
Peak Power (MW)	3.7	10	<2.1	6	<2	29	60

This is the hard part - getting the computing power without melting the building.



\*Advanced Scientific Computing Research (Dept. of Energy)

PF = petaflops, floating point operations per second

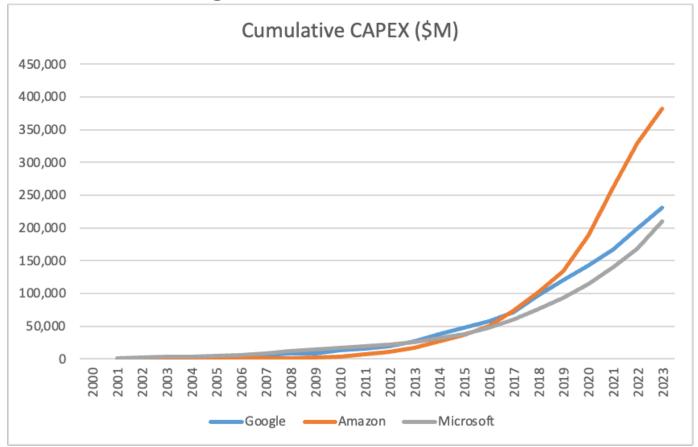
1,000 PF = 1 exaflops



http://exascaleproject.org/

# Commercial cloud computing (>> HEP computing)

- Total spending on cloud computing is now close to \$700 billion per year
- Many huge companies (Netflix, for example) don't buy their own clusters but rely entirely on cloud computing
- HEP experiments are using these resources as well



https://platformonomics.com/2024/02/follow-the-capex-cloud-table-stakes-2023-retrospective

#### Analysis techniques: tools from industry

- HEP experiments were some of the first cases where people had to deal with analyzing really big datasets
  - Had to develop our own tools to get the science done (ROOT, for example)
- Not true anymore. Basically every big company you can think of has huge amounts of data at their fingertips
  - Many tools of been developed outside labs and universities to help store, process, and analyze all this data
- Fermilab's approach for CMS analysis is COFFEA (the COmpact Framework for Elaborate Algorithms)
  - Instead of a for loop over events, use array programming expressions to process many events simultaneously
  - Uses Apache Spark and tools from the scientific python "ecosystem" based on numpy

#### Al at Fermilab

#### **Algorithms for HEP science**

Physics-inspired data & models; Robust & generalizable learning; Fast and efficient algorithms

Operations and control systems

Computing hardware and infrastructure

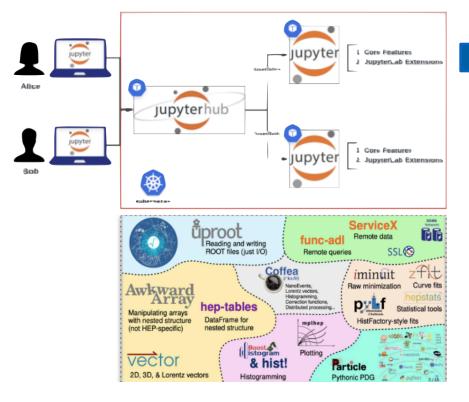
Real-time Al systems at edge

Tia Miceli - Al at Fermilab - https://indico.fnal.gov/event/59656/contributions/269042/attachments/168498/225755/20230629\_Al\_at\_Fermilab\_TiaMiceli.pdf

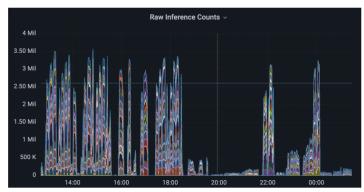
## Computing hardware infrastructure

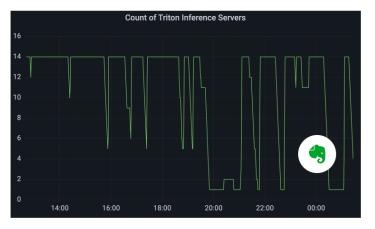
 Elastic Analysis Facility @ Fermilab provides resources and data-science standard industry tools for Al training and inference

- Additional GPU resources available on CMS LPC, Wilson Cluster
- Capable of **bursting** to O(100k) batch computing CPU cores



Flechas et al., <u>arXiv:2203.10161</u> Benjamin et al., <u>arXiv:2203.08010</u>





Tia Miceli - Al at Fermilab - https://indico.fnal.gov/event/59656/contributions/269042/attachments/168498/225755/20230629\_Al\_at\_Fermilab\_TiaMiceli.pdf

#### Conclusions

- The complexity of HEP experiments doesn't stop with the detectors
- Scientific computing permeates every aspect of how we do physics at Fermilab
- There are challenges ahead
  - Many that you could help solve!

- Thanks to all those who helped with content
  - Especially: Bo Jayatilaka, Allison Hall, Kevin Pedro, Lorena Lobato, Dmitry Litvintsev, Sophie Berkman, Oliver Gutsche, Ken Herner, Burt Holzman, Michael Kirby, Anne Schukraft, Erica Snider, Alexander Radovic, Stuart Fuess

Questions?

http://computing.fnal.gov