

A photograph of a large-scale computing facility, likely a supercomputer. The image shows rows of server racks filled with circuit boards and components. A robotic arm is visible in the center, positioned over the racks. The lighting is dim, with some red lights visible on the left side.

Computing at Fermilab

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Fermilab Summer Students School

July 20, 2022

What is scientific computing?

Computational science

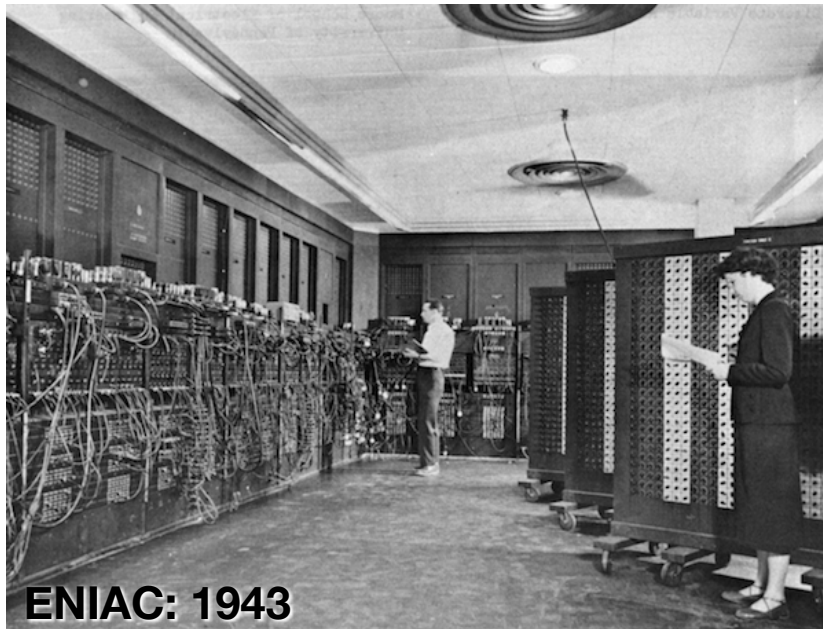
From Wikipedia, the free encyclopedia

Not to be confused with [computer science](#).

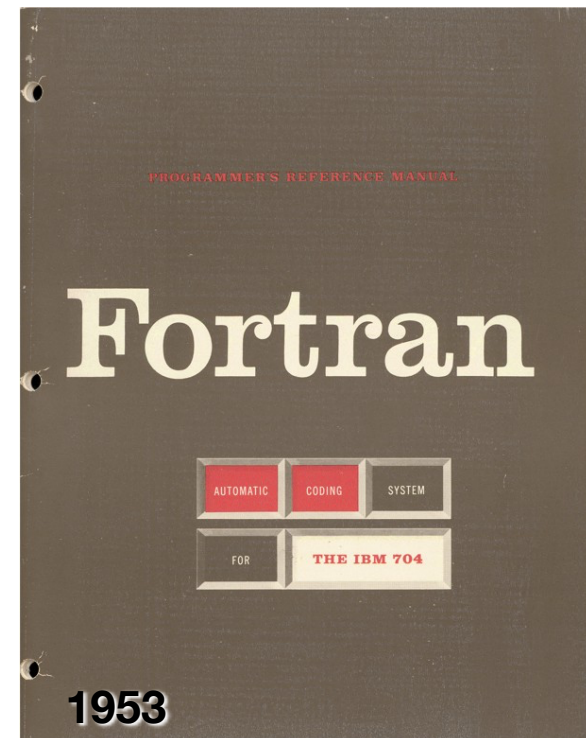
Computational science (also **scientific computing** or **scientific computation (SC)**) is a rapidly growing multidisciplinary field that uses advanced computing capabilities to understand and solve complex problems. It is an area of science which spans many disciplines, but at its core it involves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

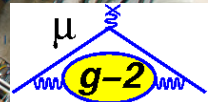
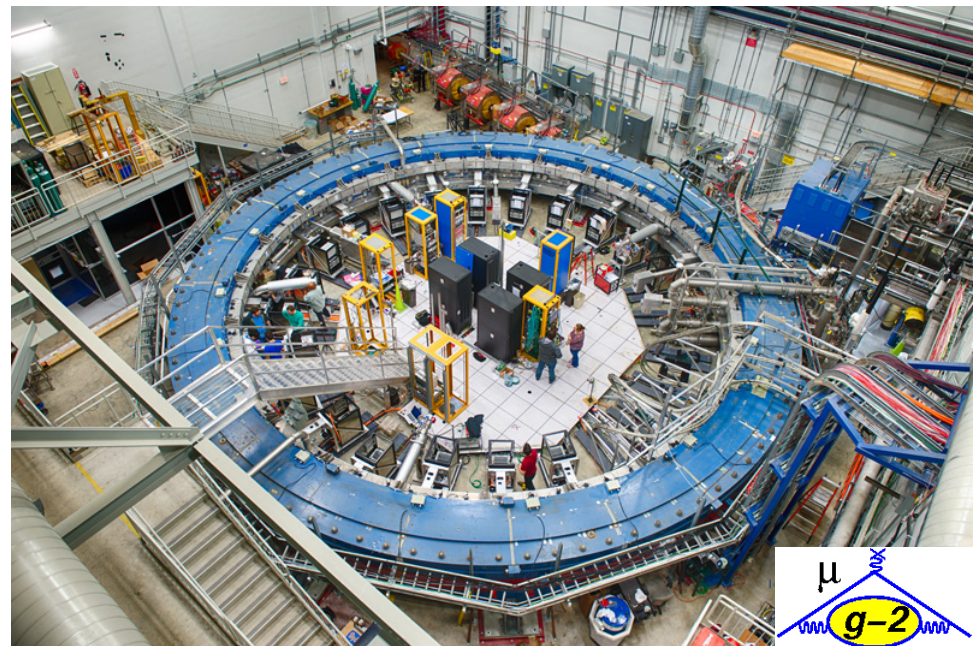
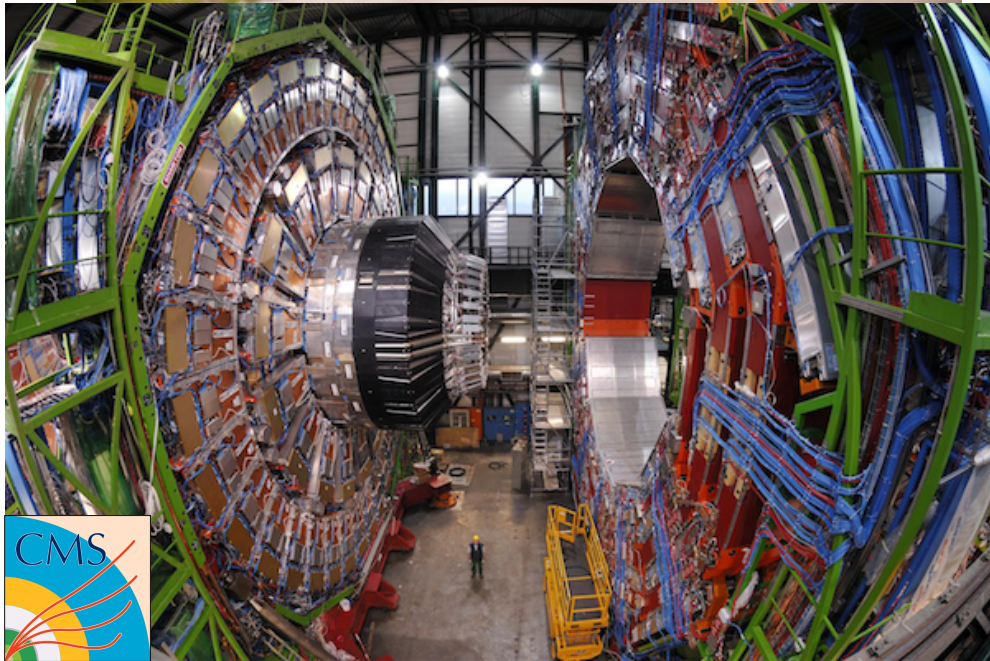
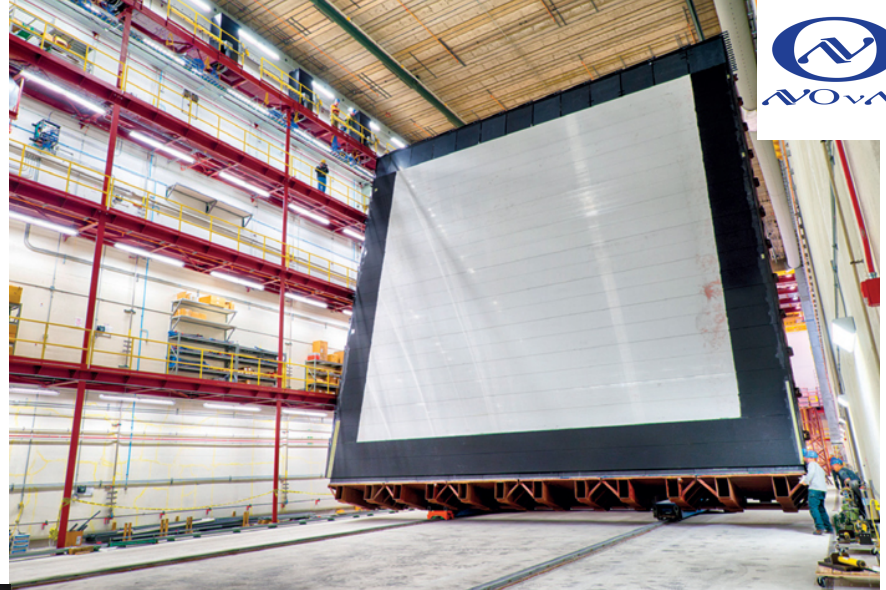
In practical use, it is typically the application of **computer simulation** and other forms of **computation** from **numerical analysis** and **theoretical computer science** to solve problems in various scientific disciplines. The field is different from theory and laboratory experiment which are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



ENIAC: 1943

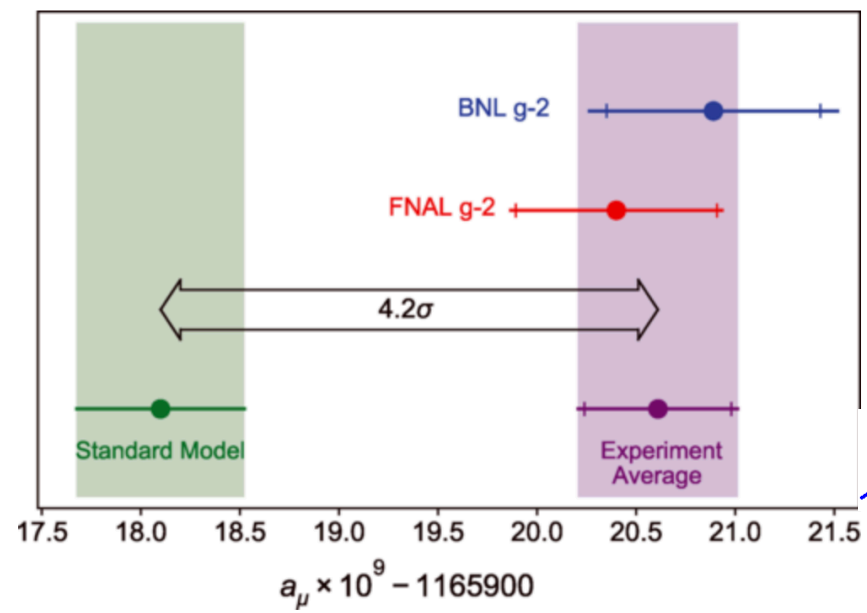
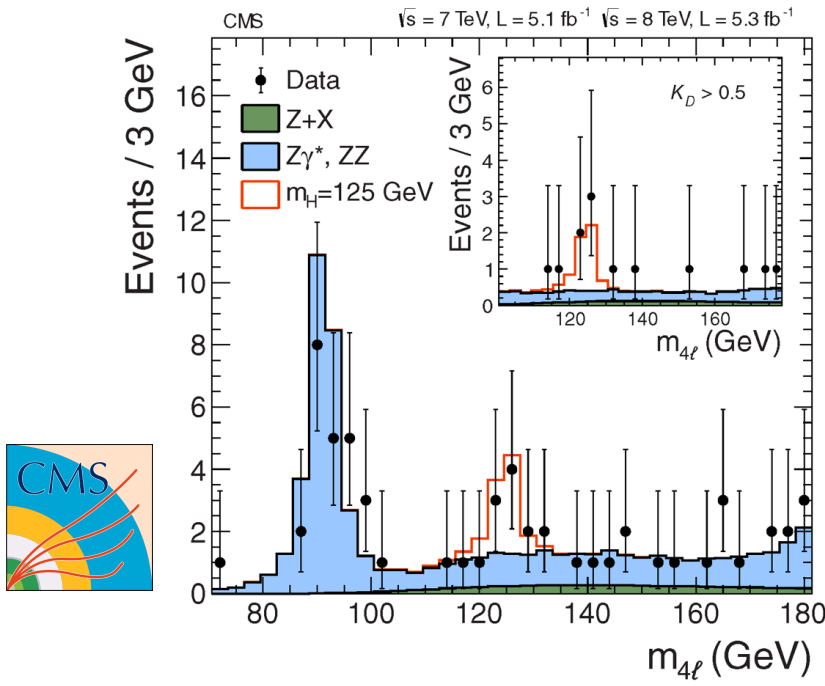
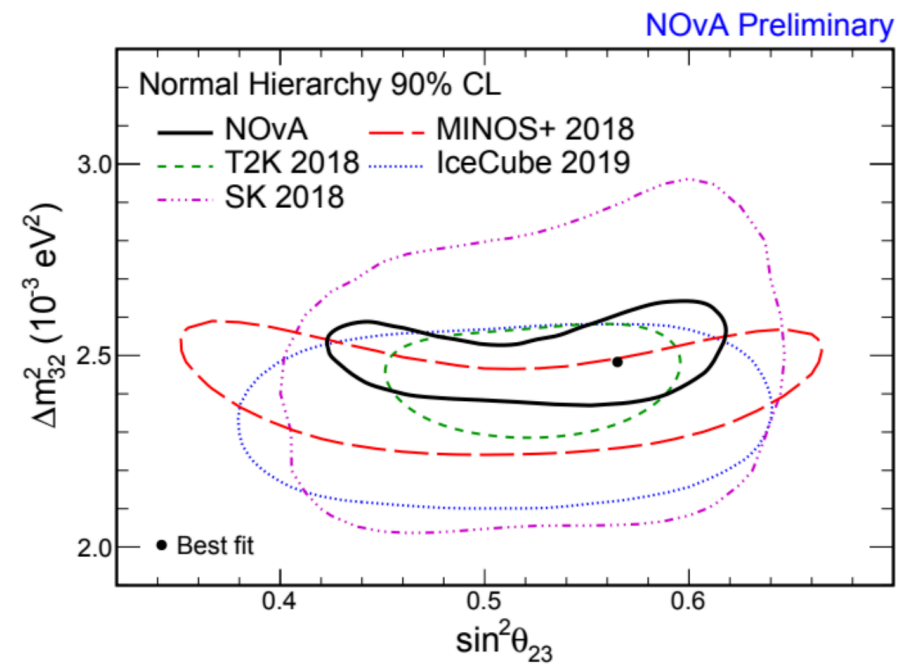
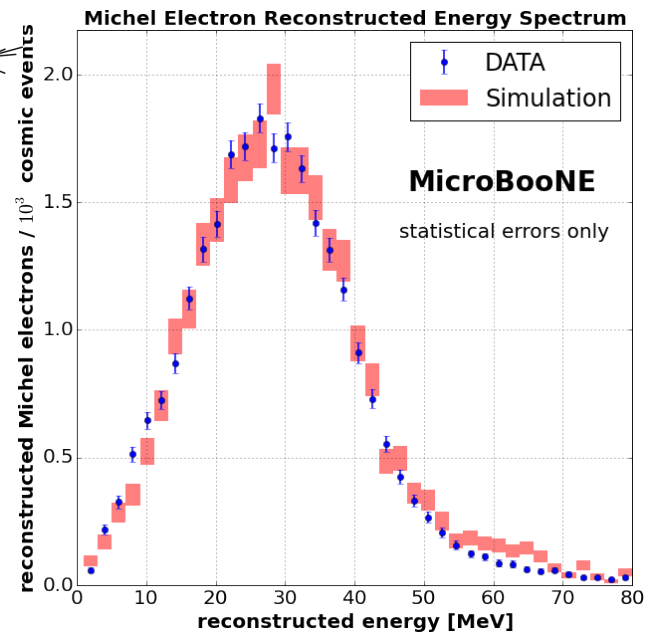


Scientific Computing in HEP: Getting from here...



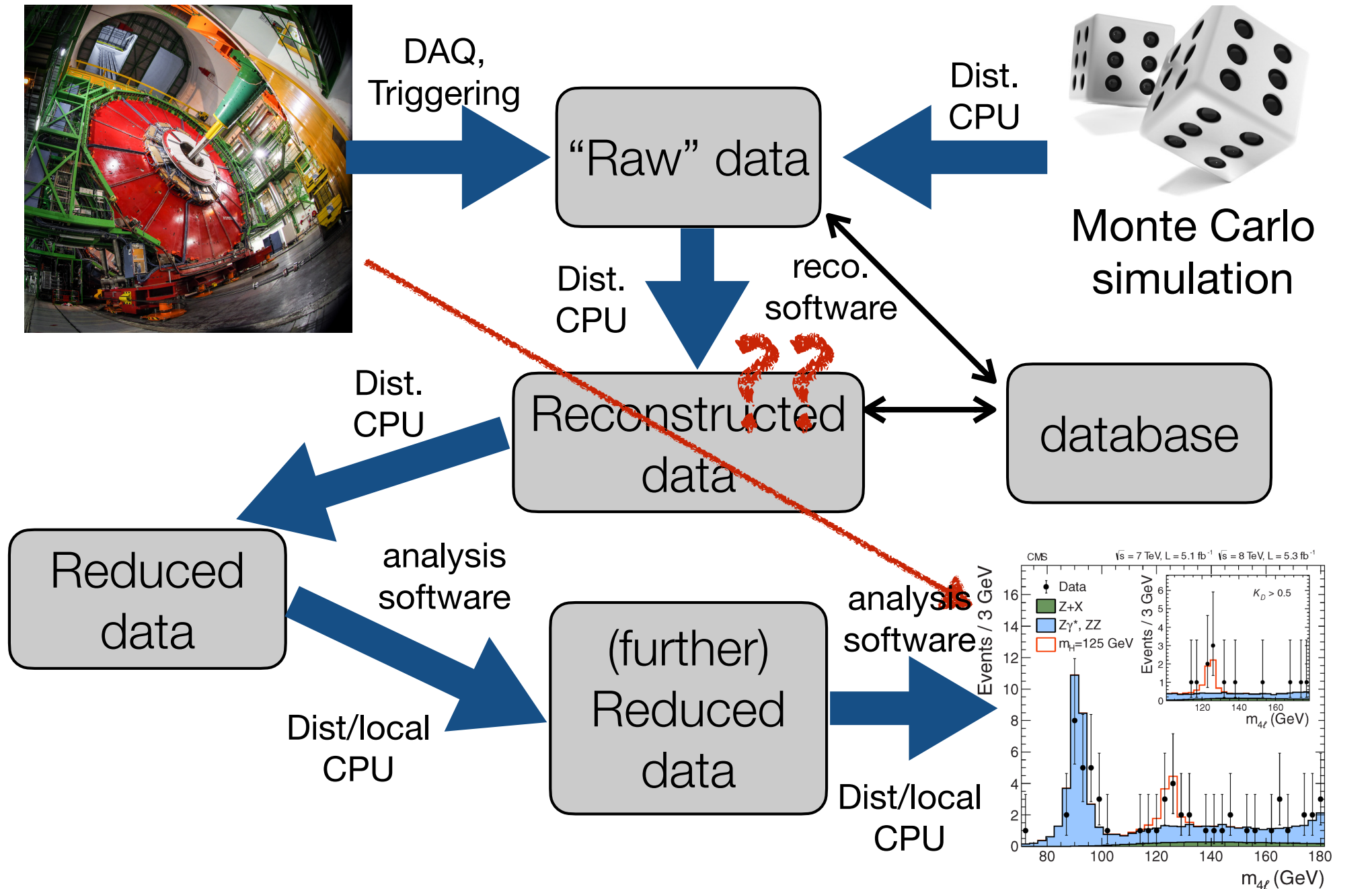
...to here

μBooNE

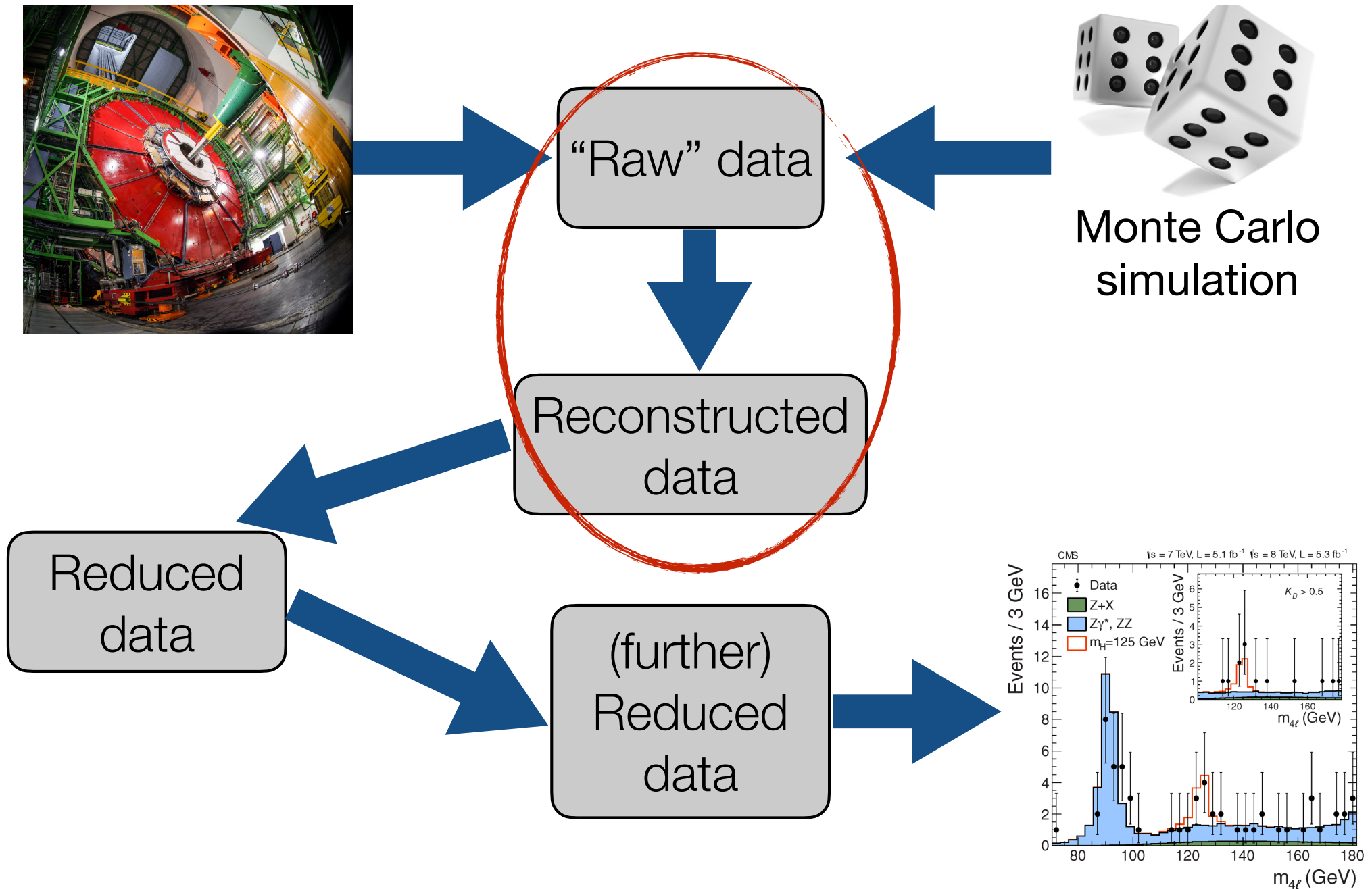


Software

The process

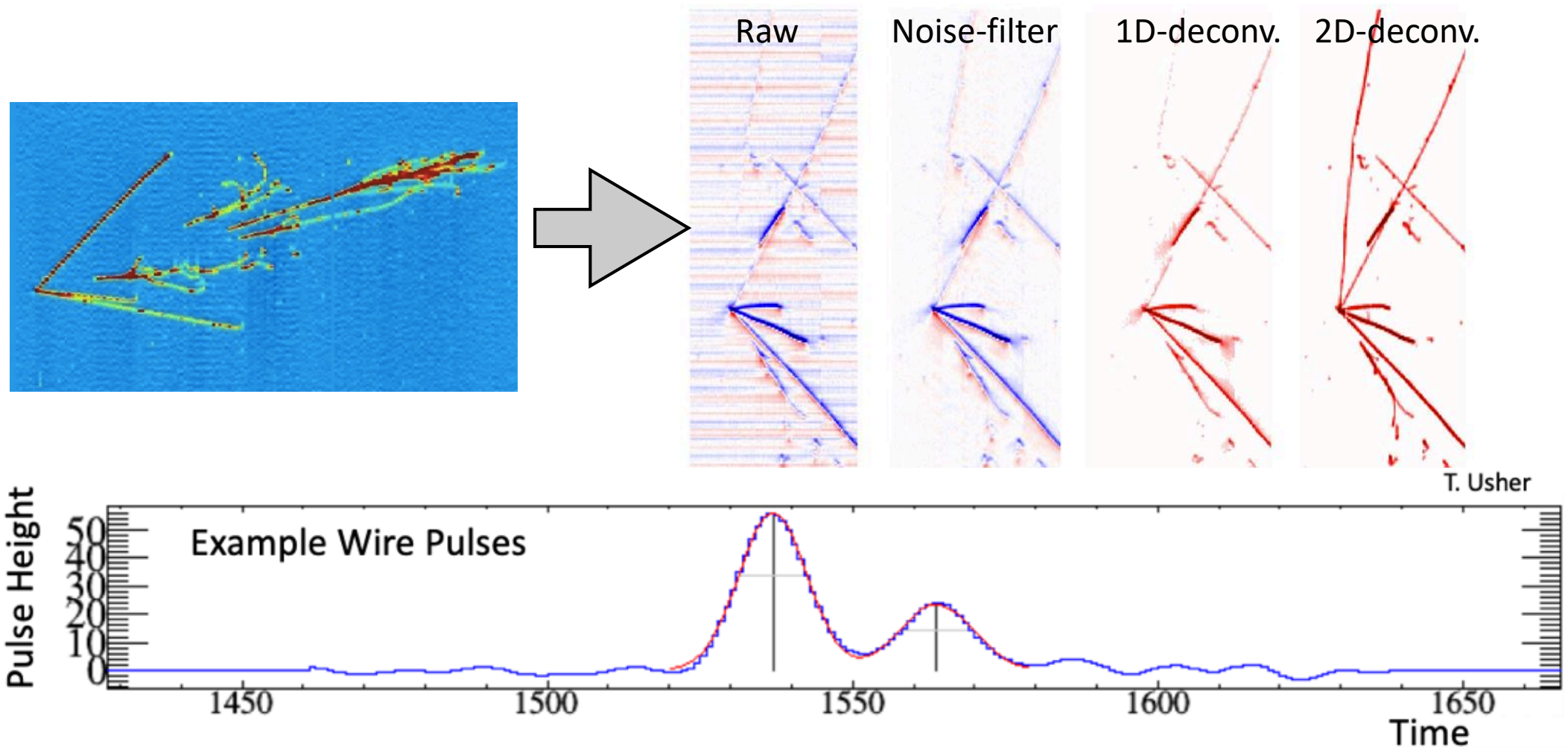


Reconstruction

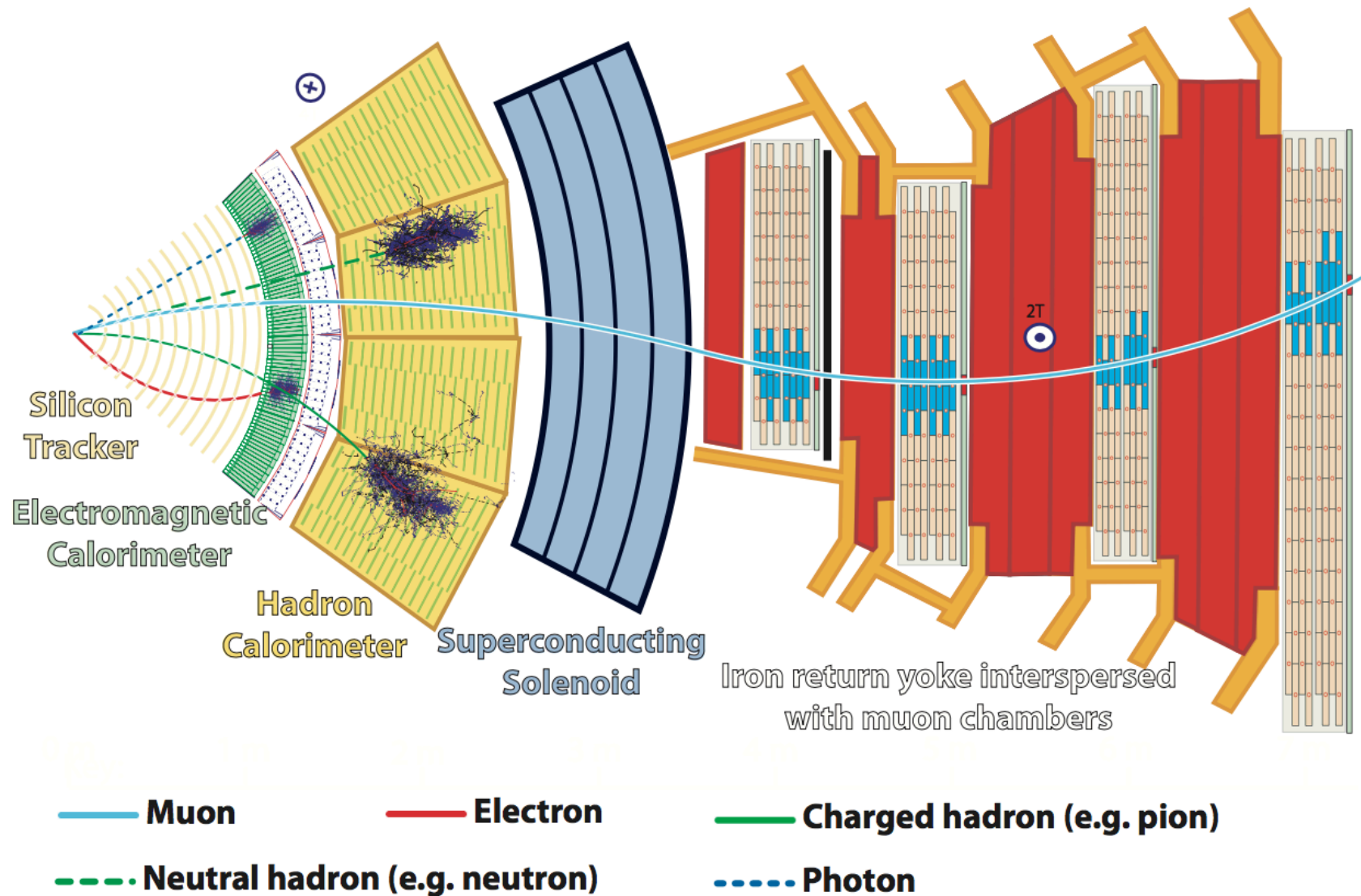


Reconstruction: LArSoft (MicroBooNE)

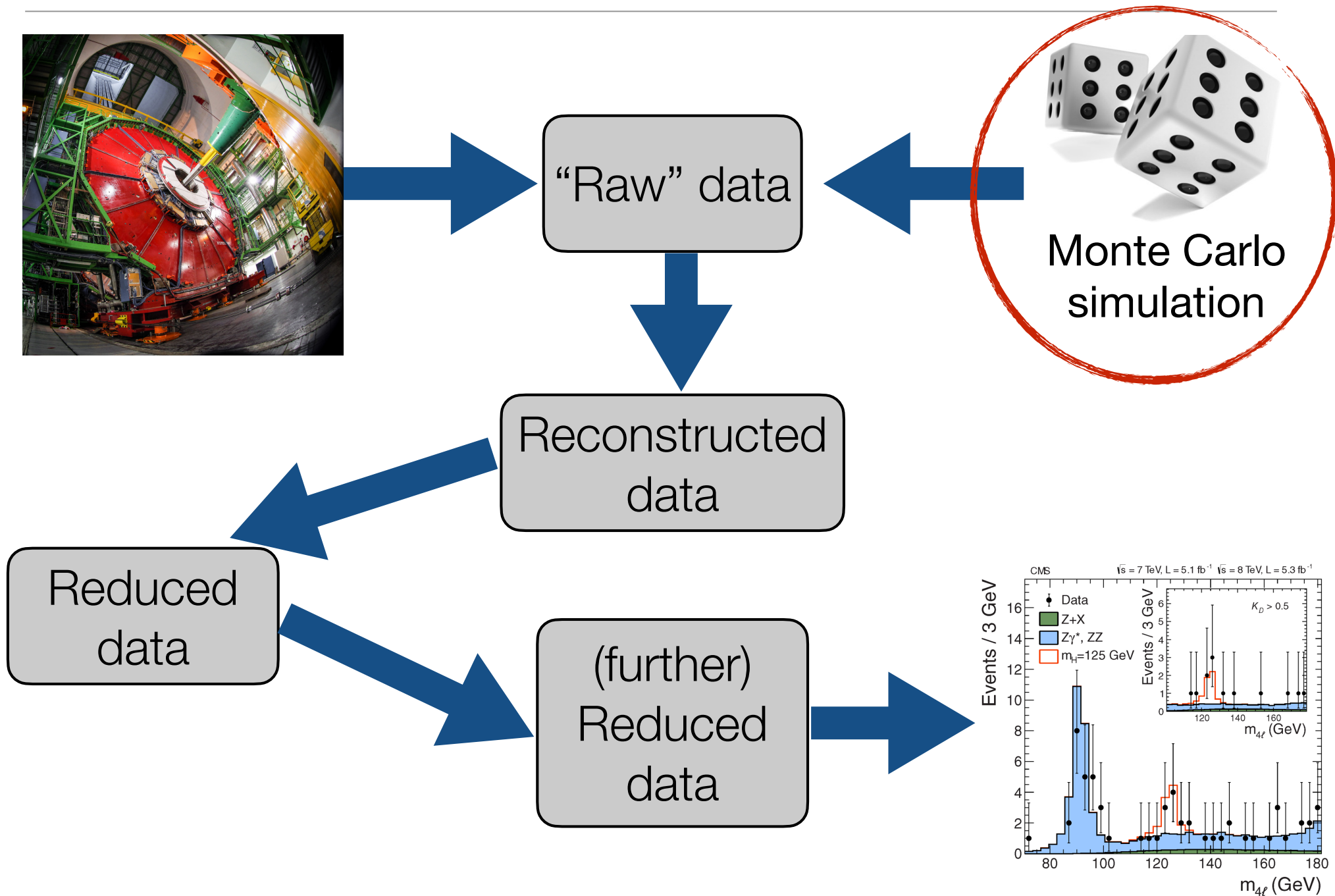
- **LArSoft**, developed at Fermilab, is a common code base for LAr neutrino detectors
 - Used by MicroBooNE, SBND, ICARUS, DUNE
 - Includes tools for both reconstruction and data analysis



Reconstruction: CMS

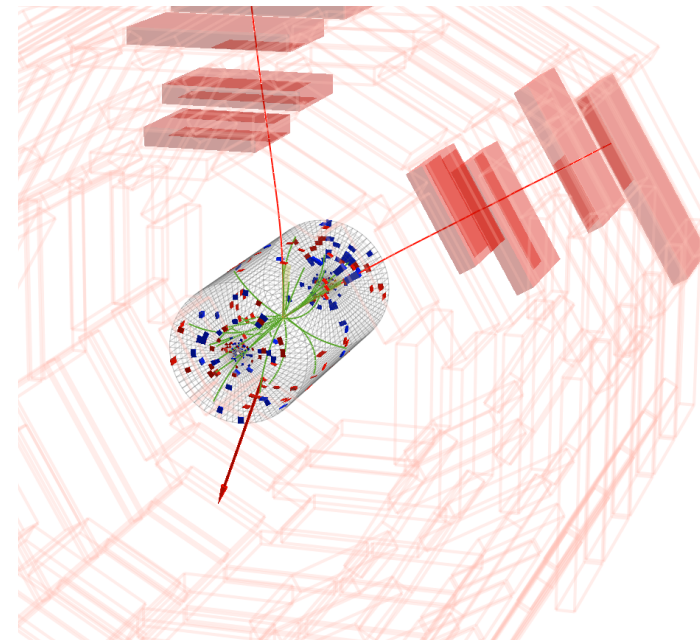
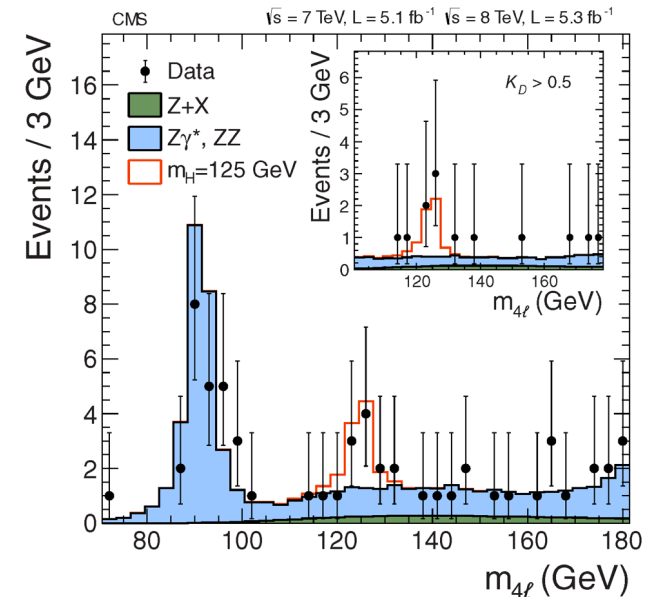


Simulation

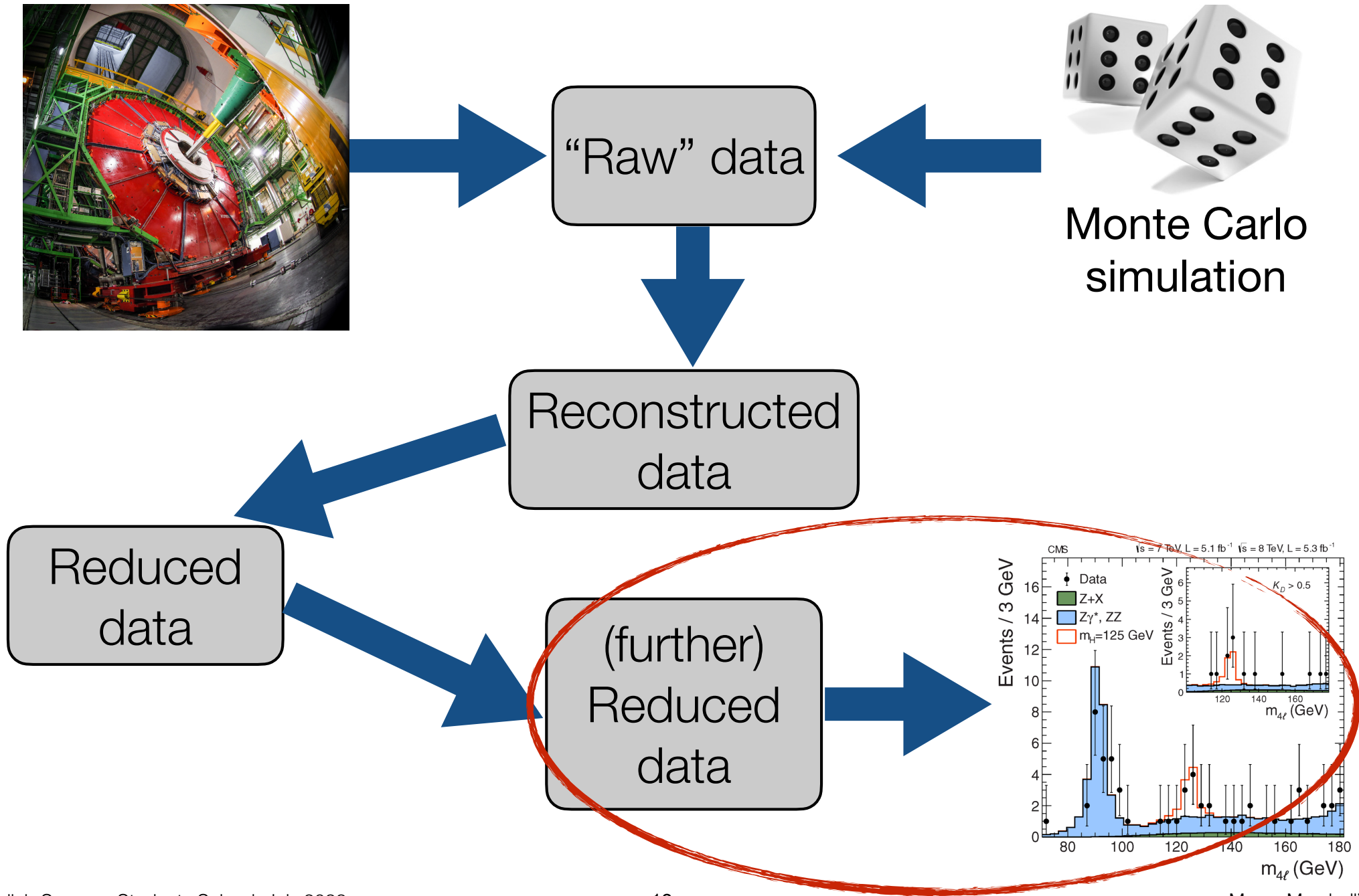


Simulation

- **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

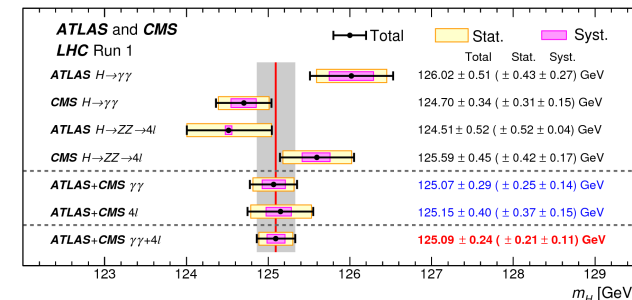
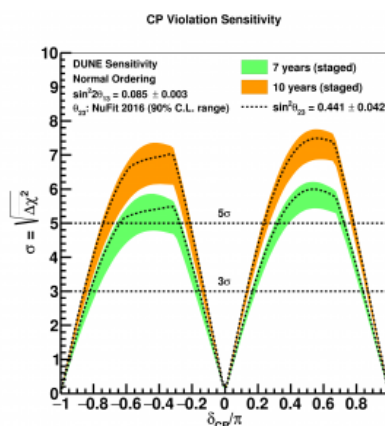
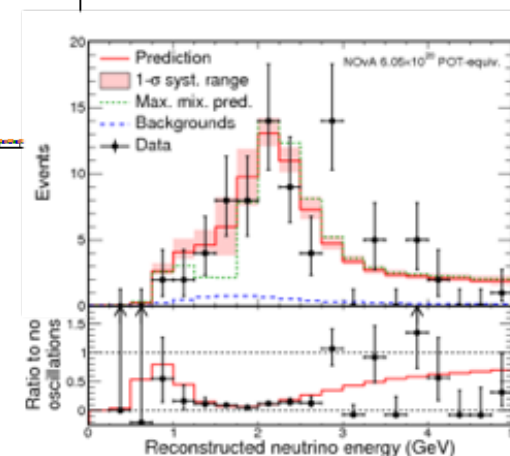
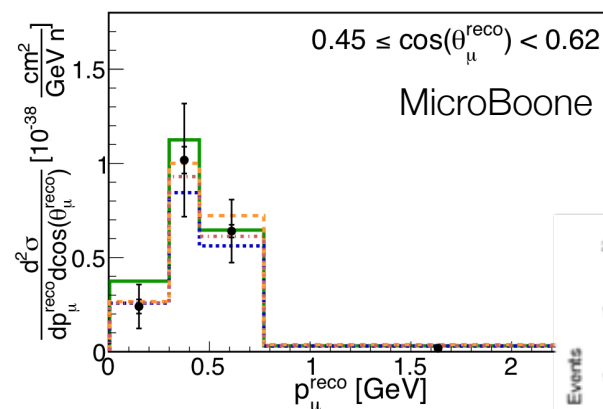


Analysis



Analysis

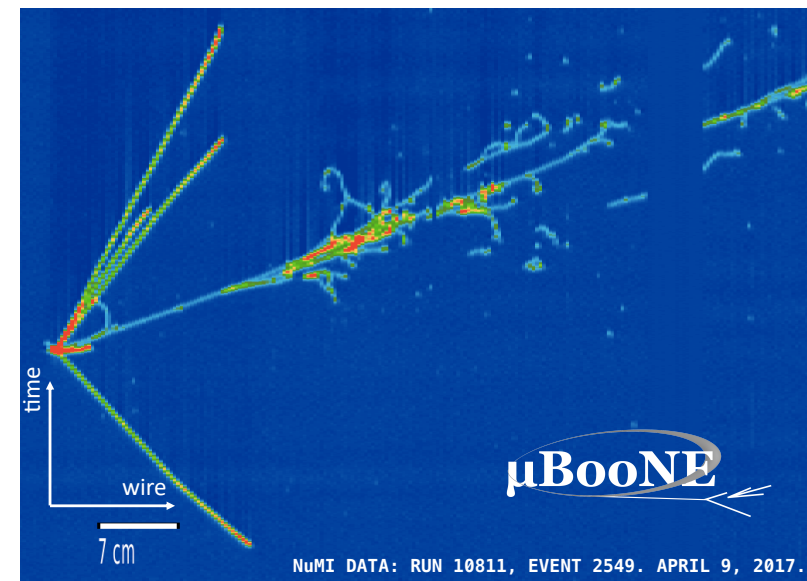
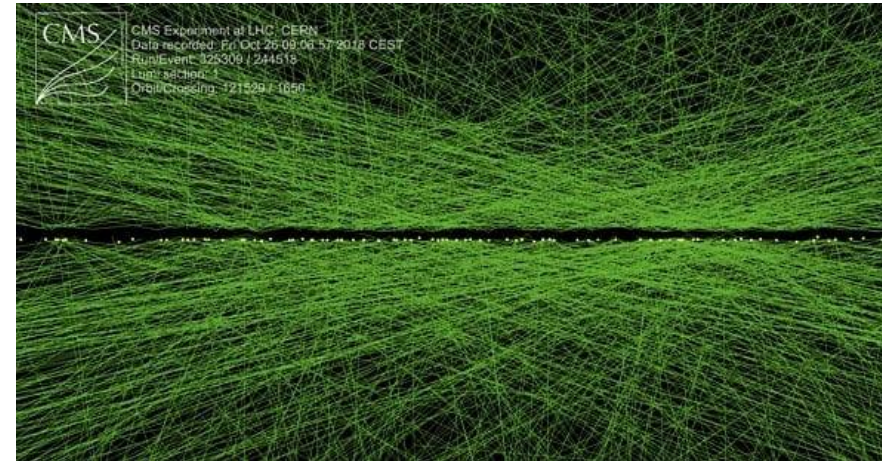
- Getting from data events of interest to plots, tables, and numbers
 - This is the computing step nearly all HEP experimentalists are familiar with
- Common tools are needed
 - Mathematical functions
 - Statistical analysis
 - Plotting/histogramming
- Nearly all HEP experiments use the **ROOT** framework
 - 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>



Hardware (CPUs, Storage and Networks)

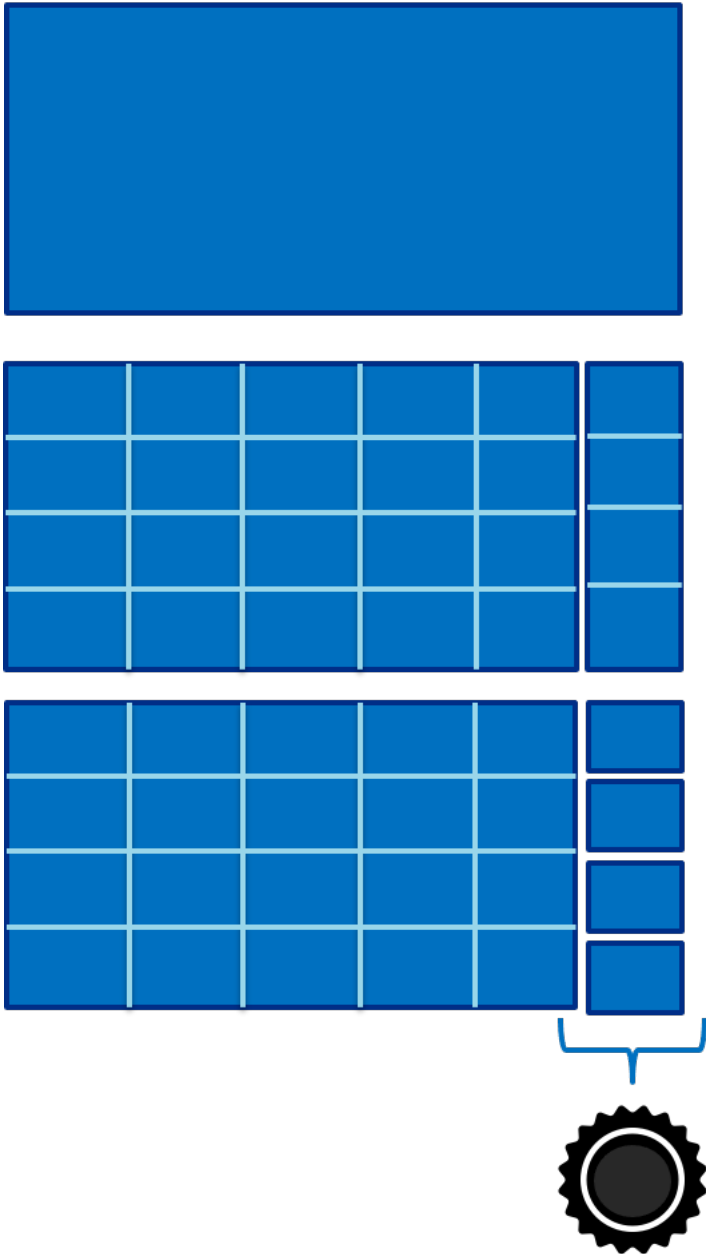
CPU

- **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 decade!)
 - ~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 (!)



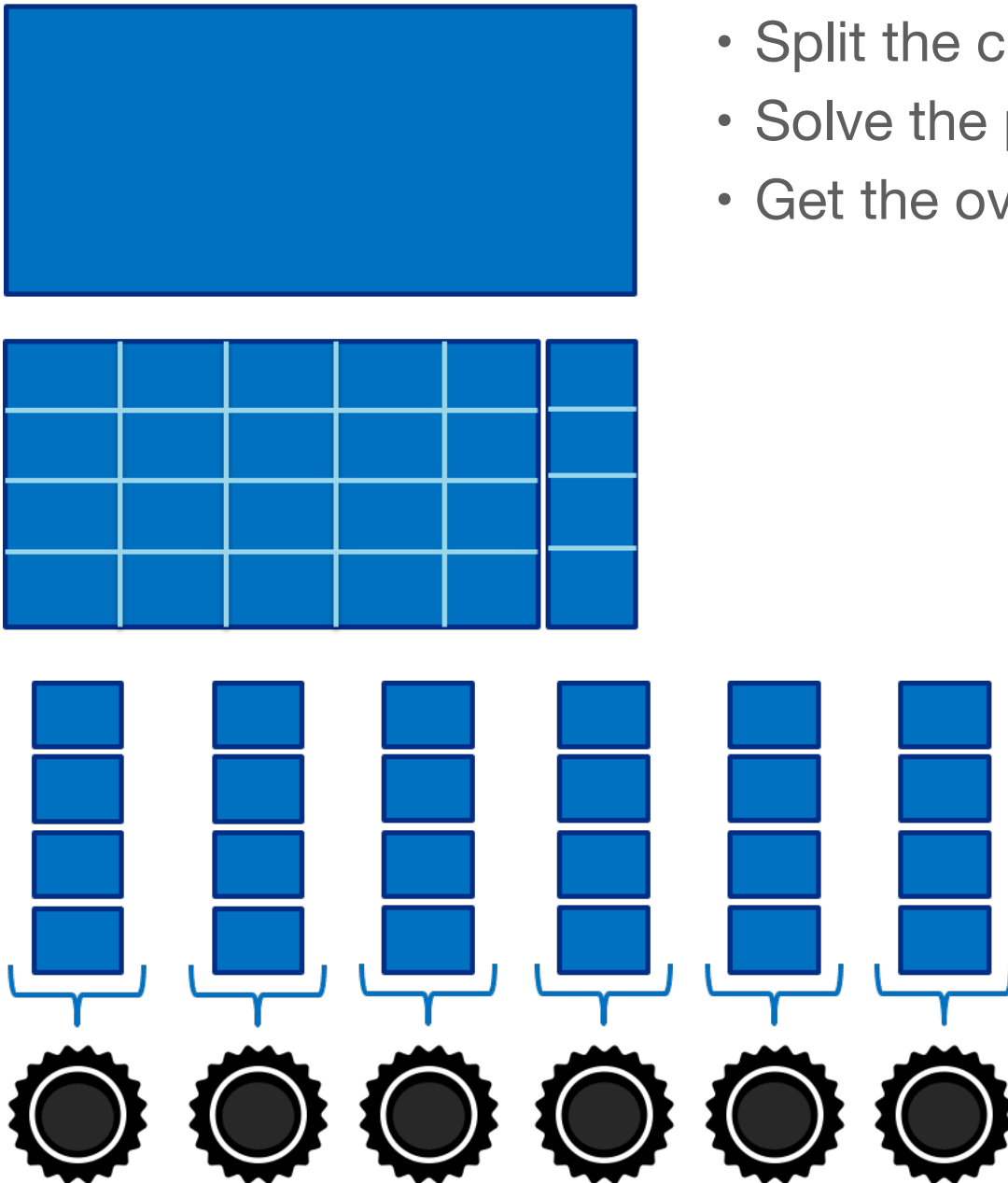
Divide et Impera

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



Divide et Impera

- Split the complex problem
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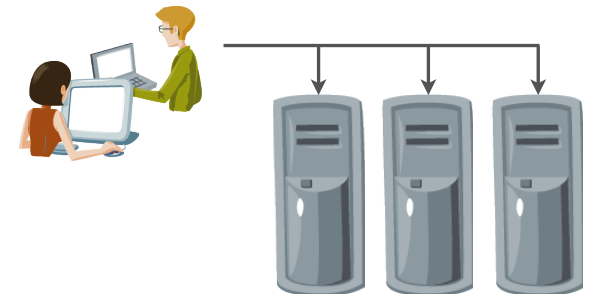
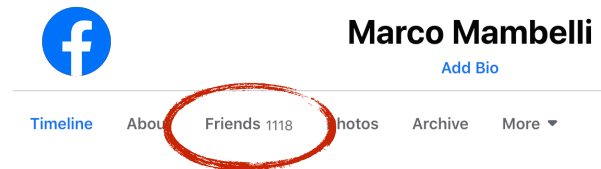
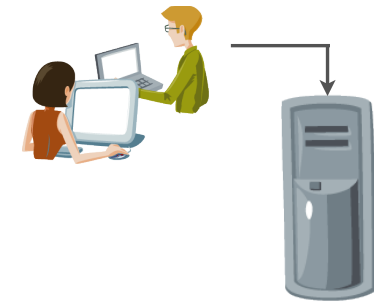
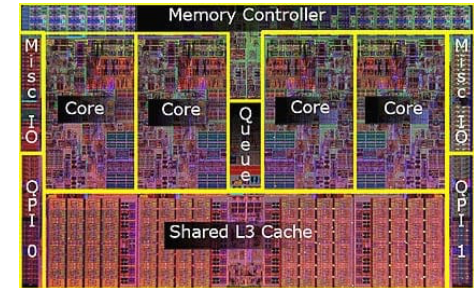


Get it fast!

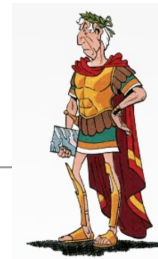
High Throughput Computing
(HTC)

Divide et Impera: 1 Billion Events

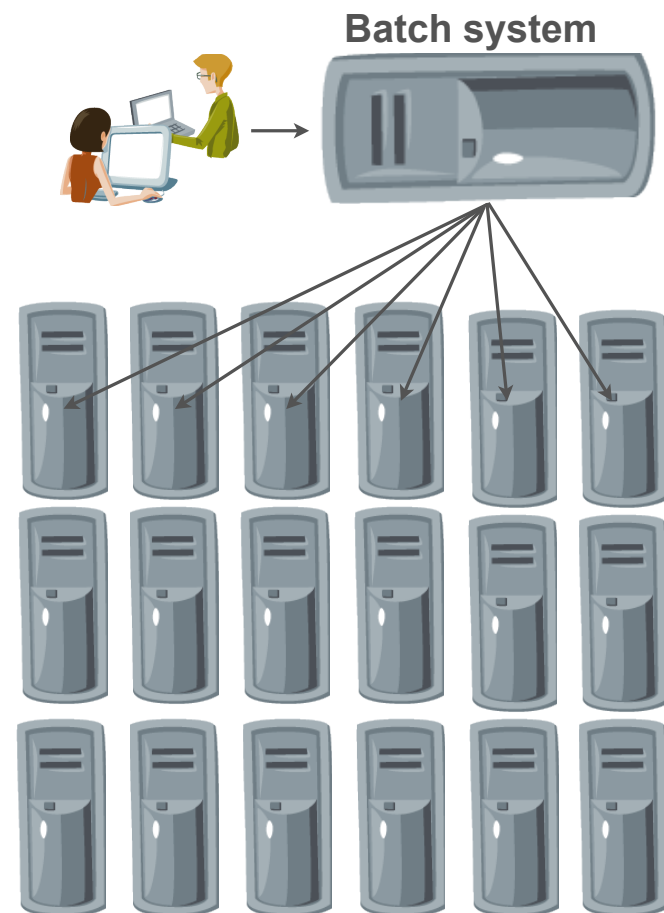
- Use one computer for the CMS events?
 - CPU: Central Processing Unit (typically whole chip)
 - Between 2 and 64 individual cores
 - Each core can process one instruction at a time
- 30s/8 cores*1B events ~120 years*
- Use your friends' computers as well
 - To get 1B events in one month, we require **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!



Divide et Impera: batch systems



- 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



Some batch systems at Fermilab:

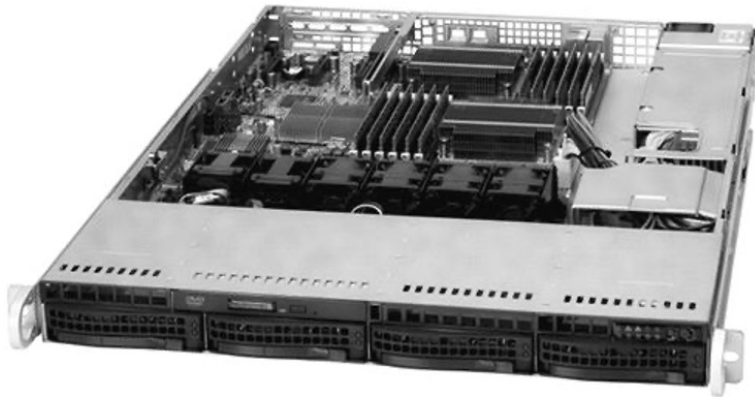
Fermigrid (~20k CPU cores)

US CMS Tier1 (~20k CPU cores)

CMS LPC (~5k CPU cores)

$30s/20,000 \text{ cores} * 1B \text{ events} \sim 17 \text{ days}$

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:

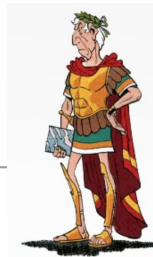
- Room A, Tape Room, Network Room A
- Rooms B, C, Network Room B

“Lights out” facility, with UPS systems,
only good for minutes

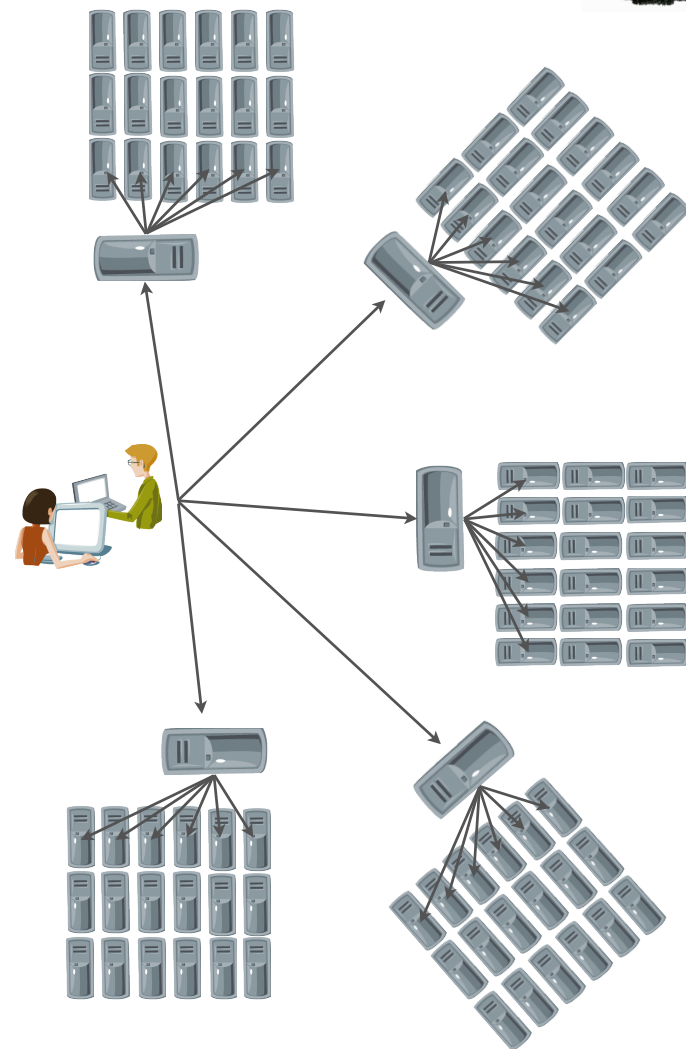
Each a “corner” of a redundant network



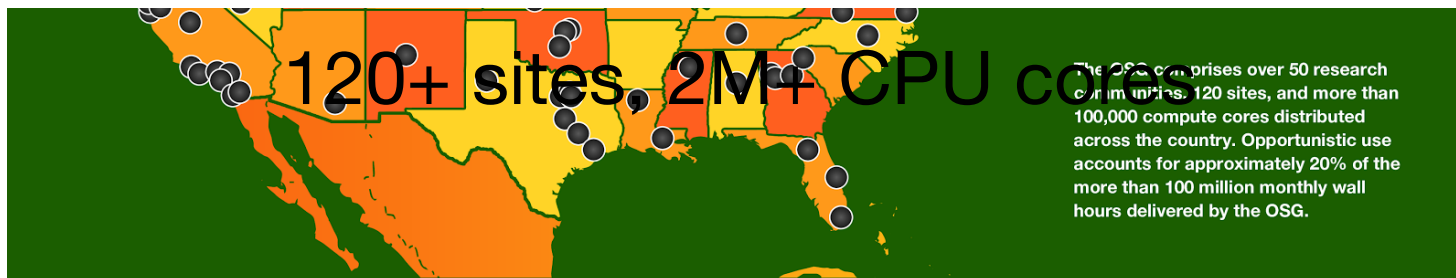
Divide et Impera: the Grid



- 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



Overlapping grids



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 seemingly 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

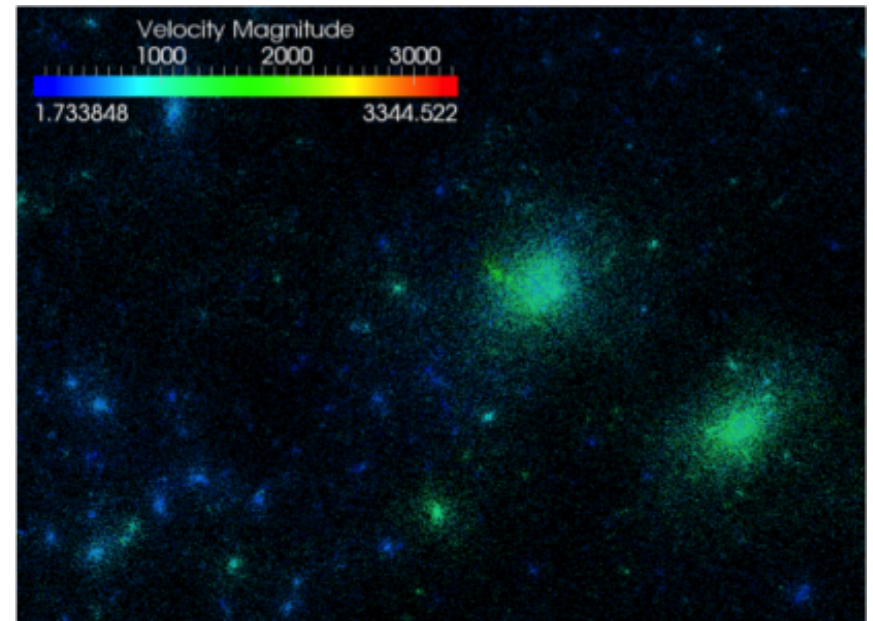


High Performance Computing

- 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

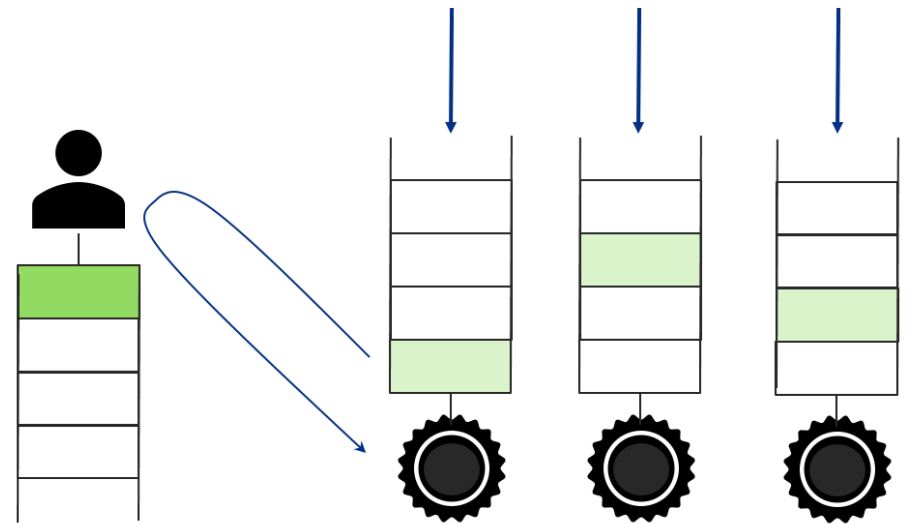


Aurora@Argonne: 20k CPUs 60k GPUs, 10PB memory, 2 DP exaflops

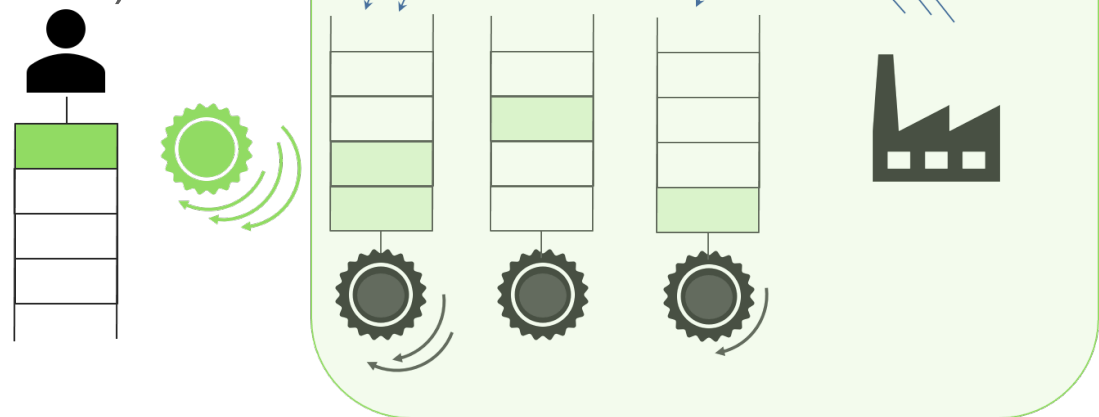


Simplify access: Pilot jobs

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



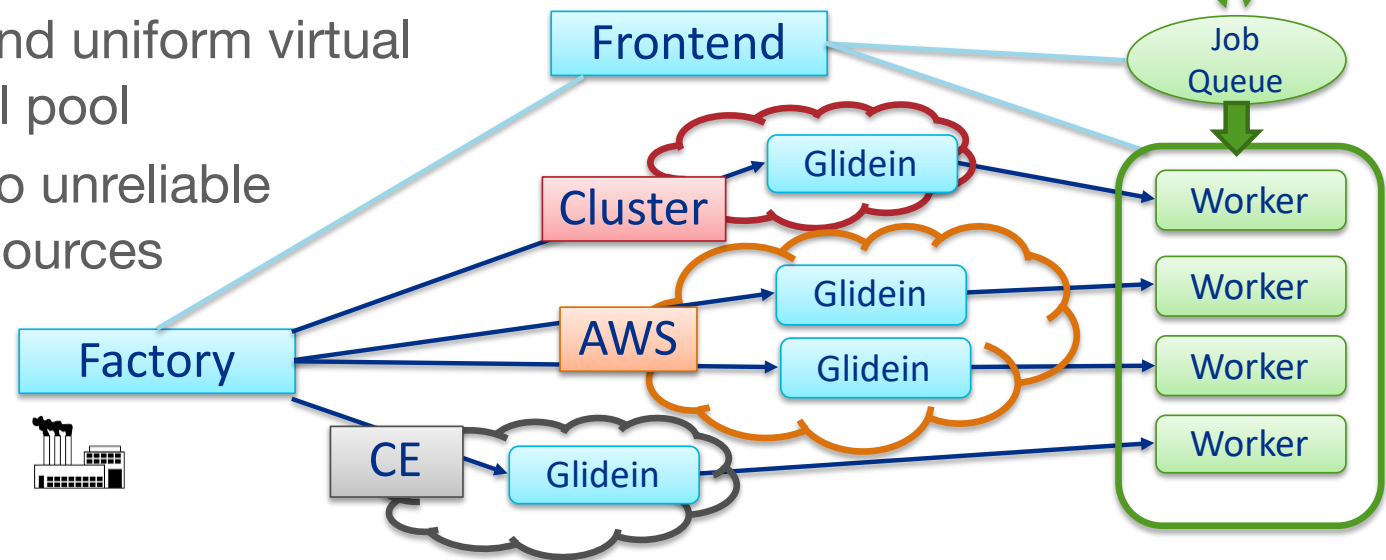
- A Factory will submit pilot jobs for you
 - Use resources as available (pressure)
 - Separation of knowledge



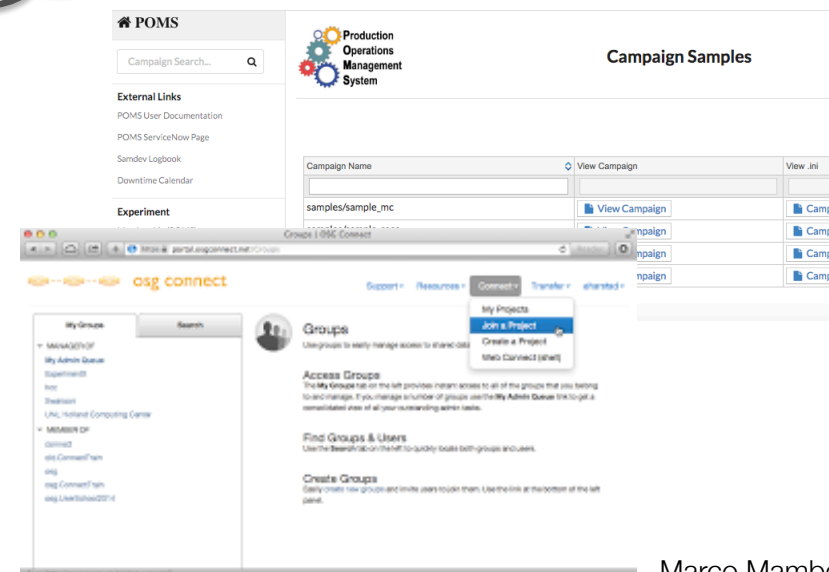
Putting it all together: GlideinWMS

- 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

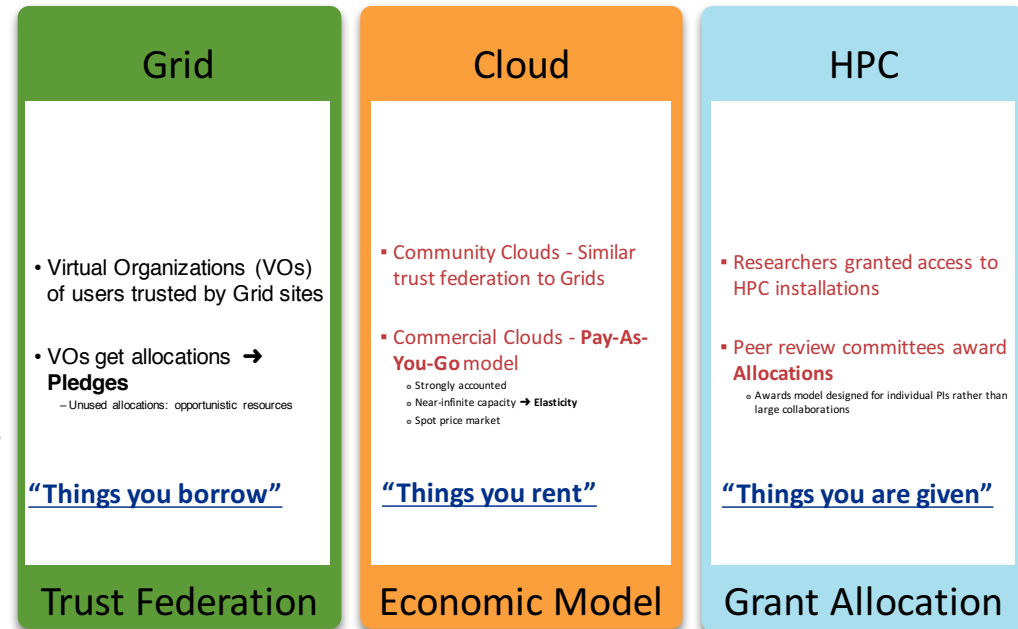


- Knows "how to talk" to all the different systems
- Multiple Frontends and Factories work together to provide High Availability
- Used by: CMS, DUNE, FIFE, OSG, (POMS, Jobsub, OSG-Connect)

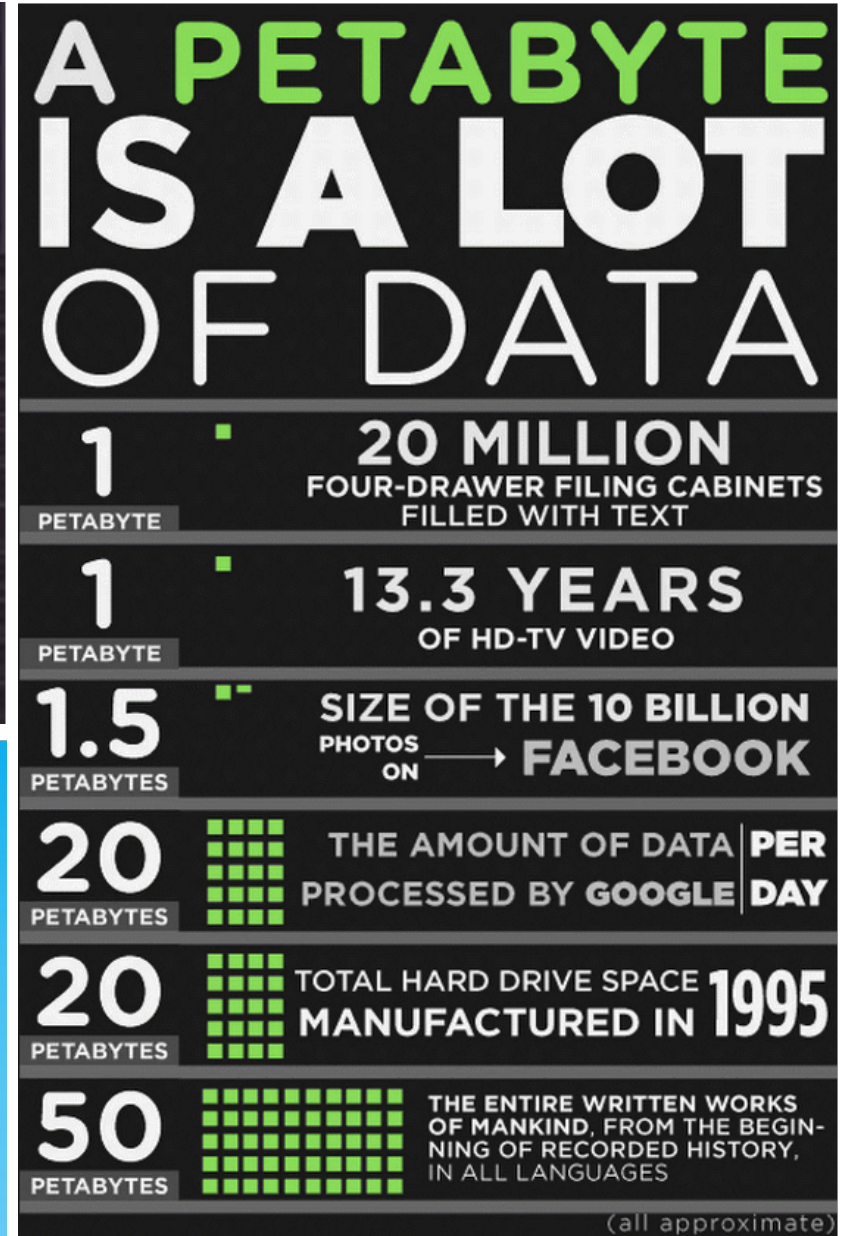
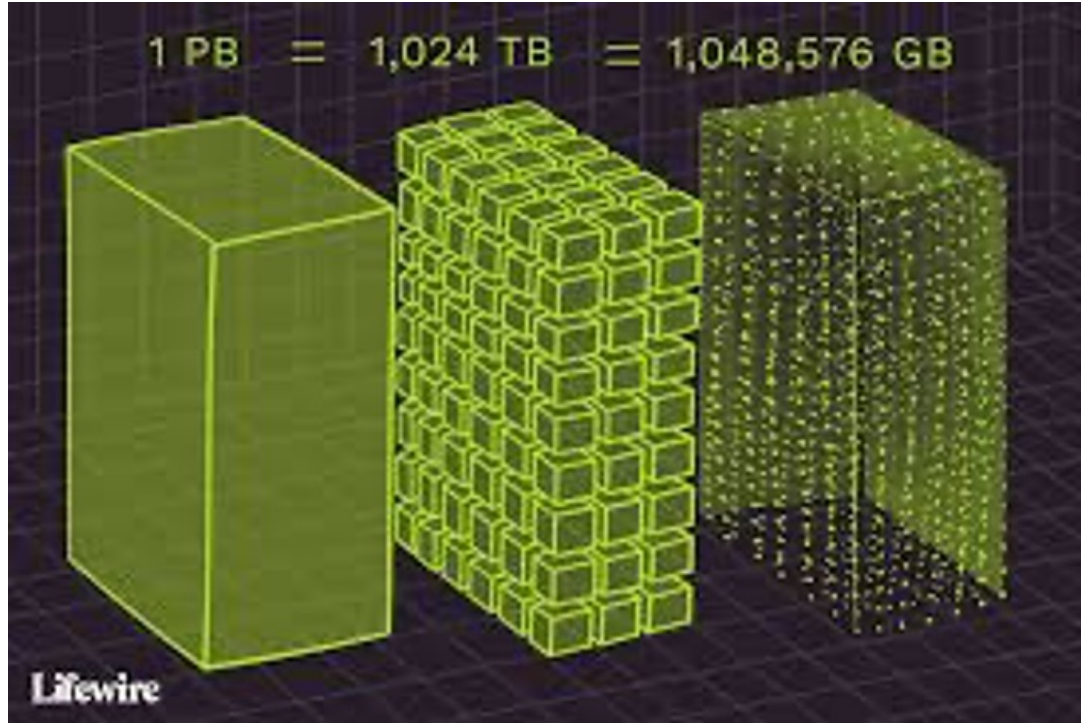


HEPCloud

- 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

1 PB = 1,024 TB = 1,048,576 GB



Scientific data stored at Fermilab = 380 PB!

It would take over



20 YEARS TO
TWEET 150 PB

if everyone in the world tweeted
once per day.

A **PETABYTE**
IS A LOT
OF DATA

1 MILLION
FOUR-DRAWER FILING CABINETS
FILLED WITH TEXT

1 PETABYTE
13.3 YEARS
OF HD-TV VIDEO

1.5 PETABYTES
SIZE OF THE 10 BILLION
PHOTOS ON **FACEBOOK**

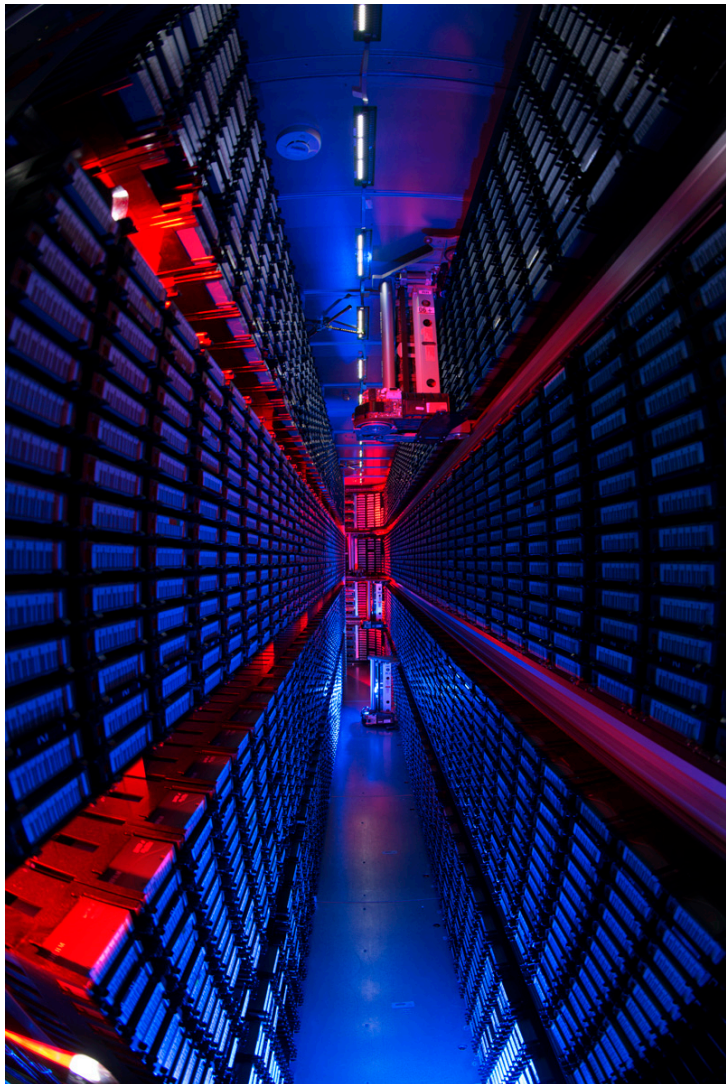
20 PETABYTES
THE AMOUNT OF DATA
PROCESSED BY **GOOGLE** **PER DAY**

20 PETABYTES
TOTAL HARD DRIVE SPACE
MANUFACTURED IN **1995**

50 PETABYTES
THE ENTIRE WRITTEN WORKS
OF MANKIND, FROM THE BEGIN-
NING OF RECORDED HISTORY,
IN ALL LANGUAGES

(all approximate)

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=1caM9JoUONtflgqxcJBbN5OImGrpnYy5>

Storage: disk

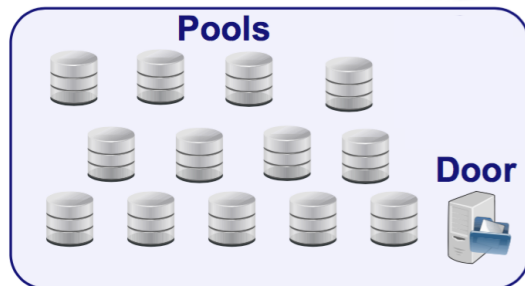
Harddrive



Diskpool



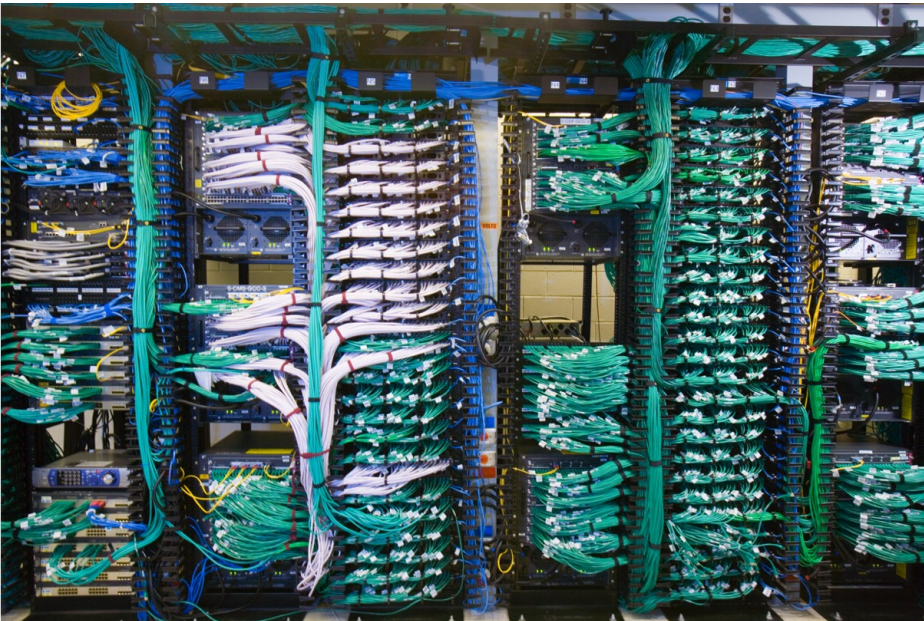
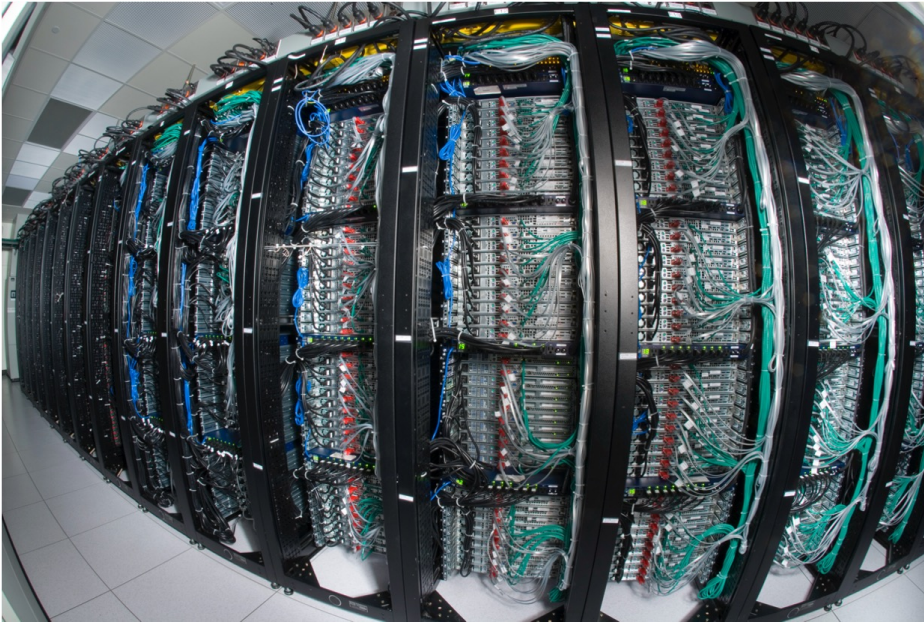
Storage System



- ~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**

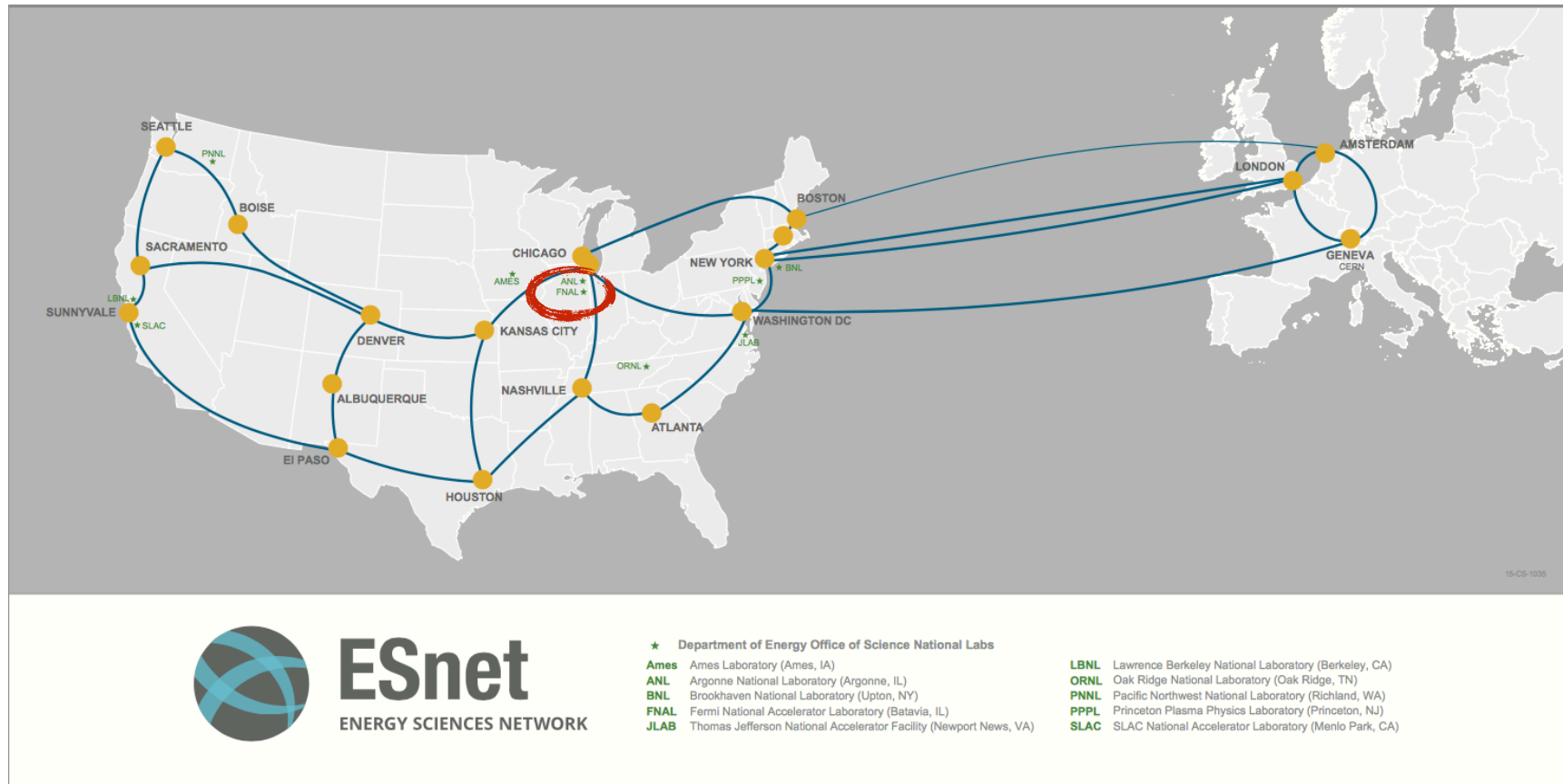


Networking: Local Networks

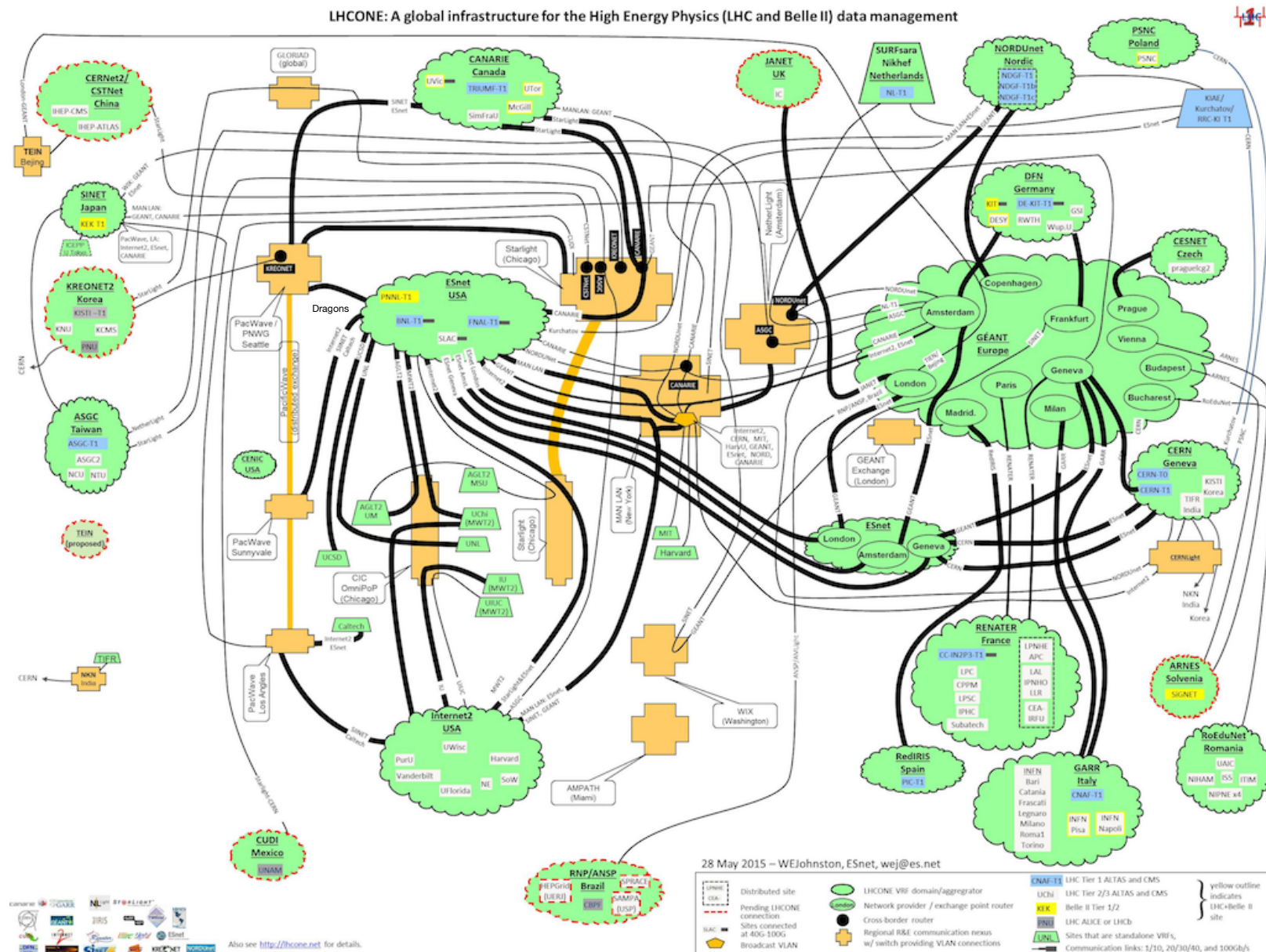


Networking: Wide Area Networks

- Distributed computing requires **fast** and **reliable networks**
- **Dedicated fiber optic links** (typically 100 Gbps) connect **experiments** and major labs
- e.g. ESNet at national labs, GÉANT in Europe
- Dedicated LHC links connect these together (LHCONE, LHCOPN)



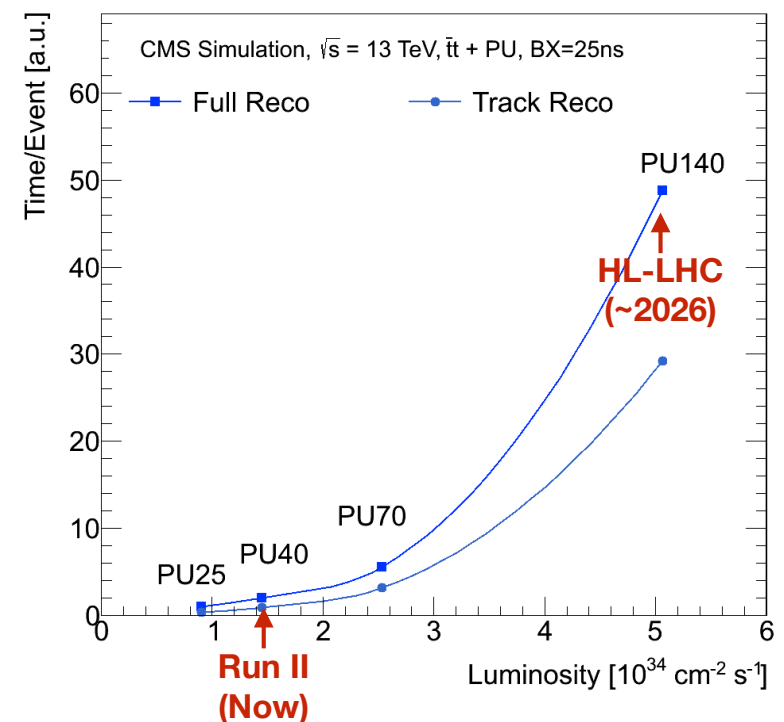
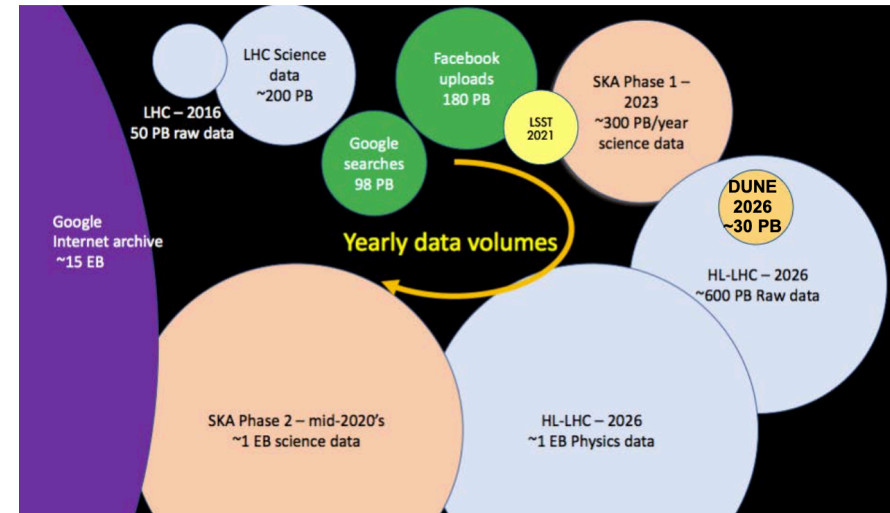
Networking



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

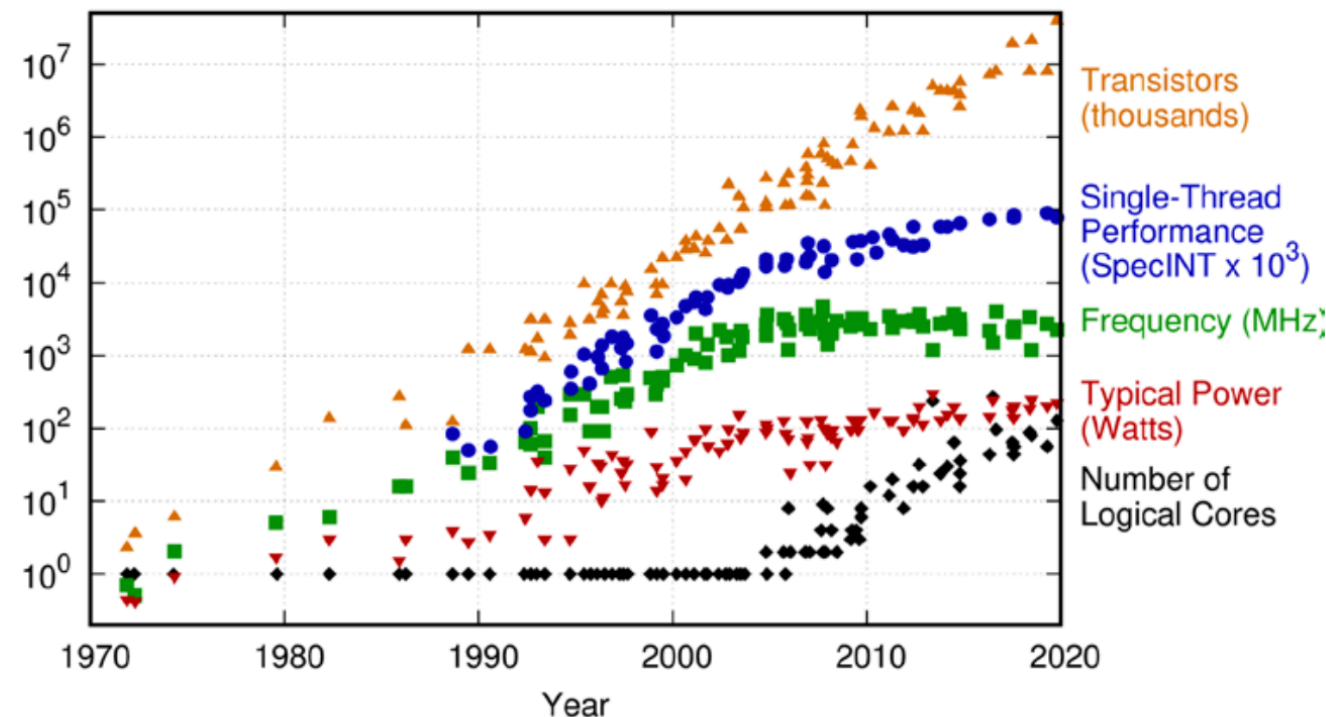


Moore's Law probably won't help

- 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

To take advantage of these:

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



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.

PF = petaflops, floating point operations per second

1,000 PF = 1 exaflops

“Excascale”

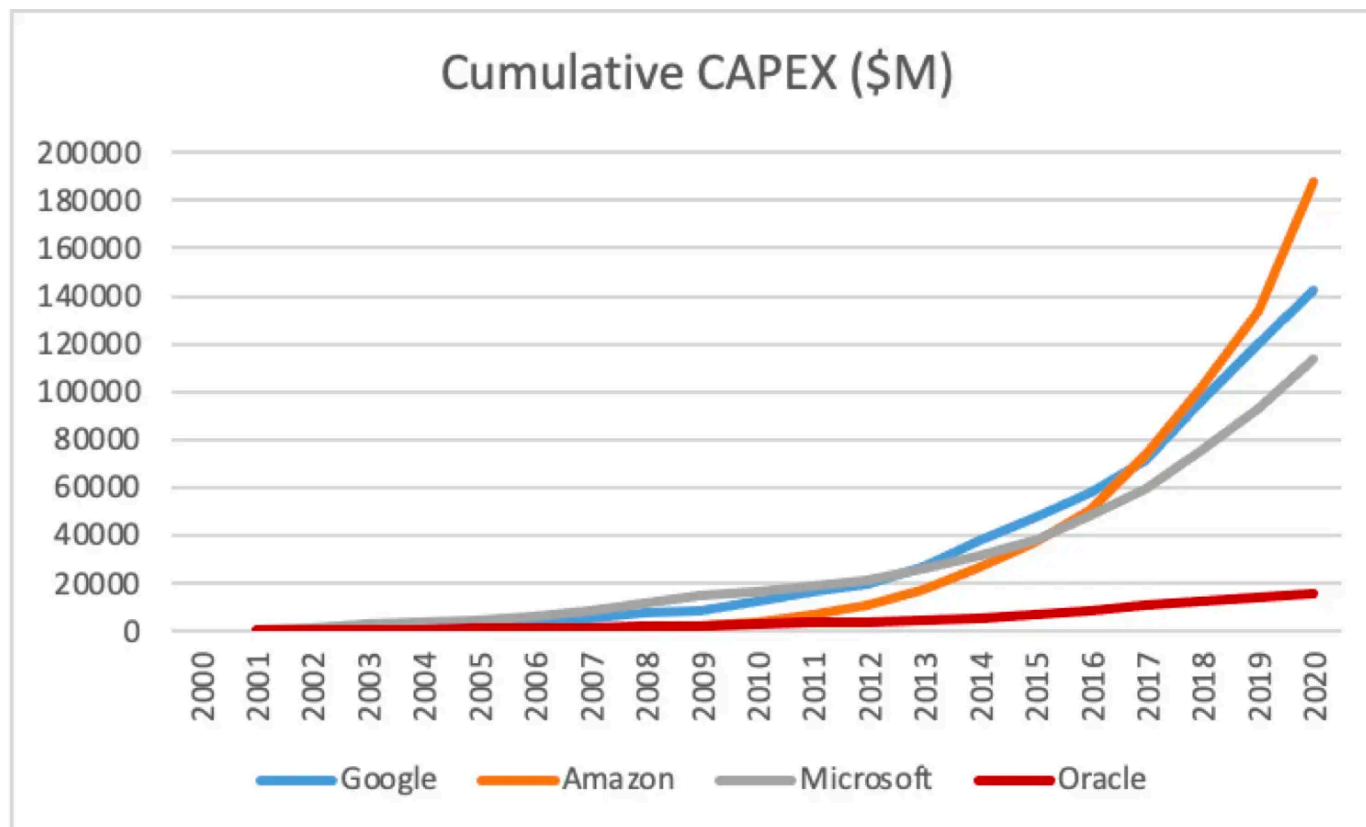


*Advanced Scientific Computing Research
(Dept. of Energy)

<http://exascaleproject.org/>

Commercial cloud computing (>> HEP computing)

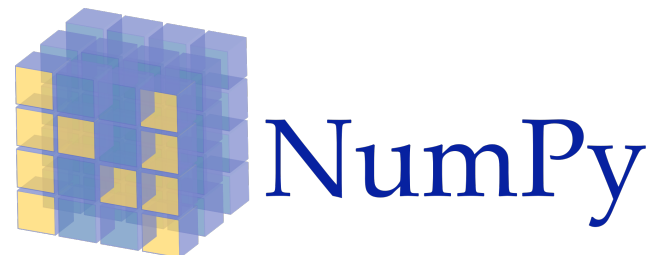
- Total spending on cloud computing is now > \$200 billion per year
- Many huge companies (Netflix, for example) don't buy their own clusters but rely entirely on cloud computing
- HEP experiments are using these resources as well



<https://www.infoworld.com/article/3639017/cloud-costs-a-lot-of-money.html>

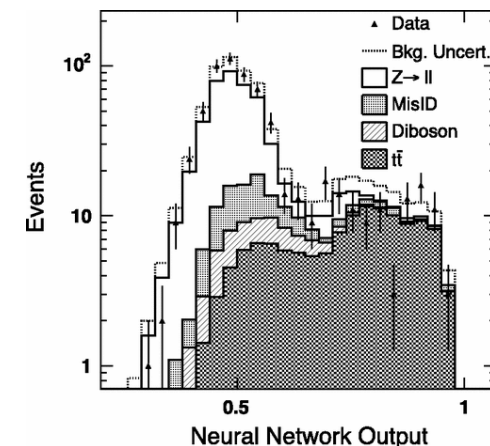
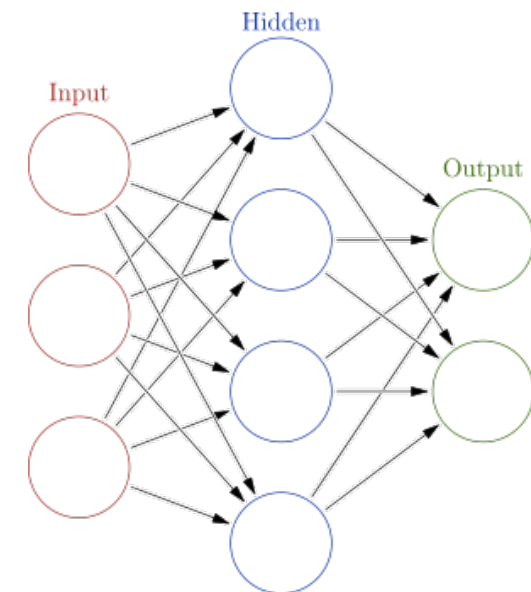
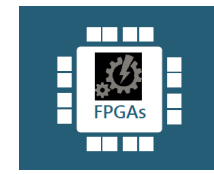
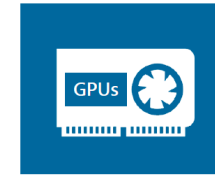
Analysis techniques: tools from industry

- 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



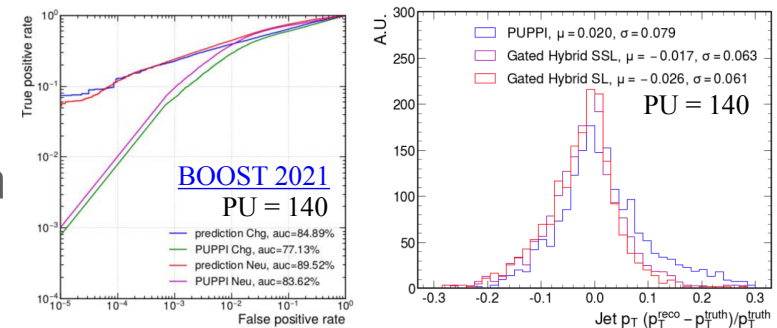
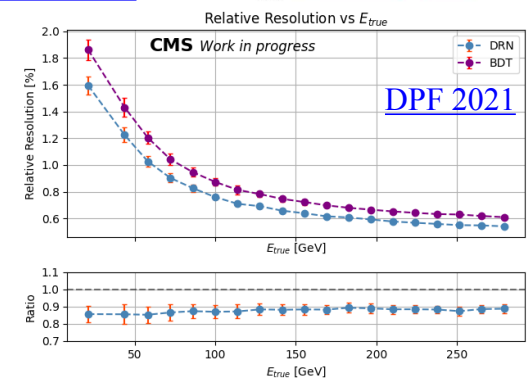
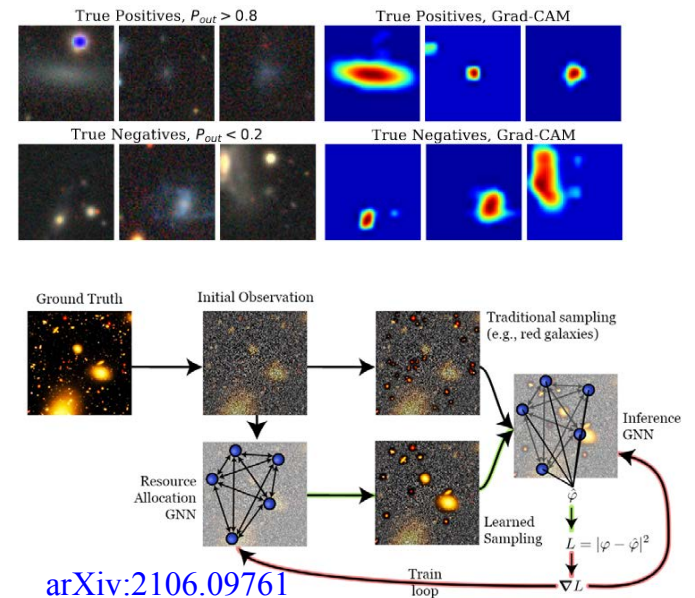
Analysis techniques: Machine Learning

- Machine Learning is a very good fit for heterogeneous computing
- ML is function approximation $\vec{x} \mapsto \vec{y}$
 - Maps inputs to outputs by optimizing weights $\vec{y} \approx F'(\vec{x}; \vec{w})$
- Deep Learning uses Neural Networks with many hidden layers to derive features from inputs (the Neural Networks Zoo)
- More neurons --> more multiplications, weights
 - Training: optimizing weights to improve function approximation
 - Inference: applying optimized function to new data to make predictions
- Used in HEP and Astrophysics since the turn of the century
- Fermilab Artificial Intelligence Project



AI at Fermilab

- DeepShadows: ([arXiv:2011.12437](https://arxiv.org/abs/2011.12437))
 - Convolutional NN to distinguish Low Surface Brightness Galaxies from artifacts in DES data
 - 92% accuracy, vs. ~80% accuracy for simpler ML methods
- Graph NN for unsupervised optimization of telescope time: pick best galaxies to observe
 - Outperforms conventional strategies
- Dynamic Reduction Network ([arXiv:2003.08013](https://arxiv.org/abs/2003.08013))
 - Learn best graph of inputs & use it for regression
 - Improve electron resolution by 10% (vs. state of the art)
 - Work in progress: apply to missing energy
- Semi-supervised Graph NN to reject pileup: trained on charged particles → can use data!
 - Significantly improves on classical algorithm
- See K.Pedro, AI at Fermilab, for more



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>