



Imperial College
London

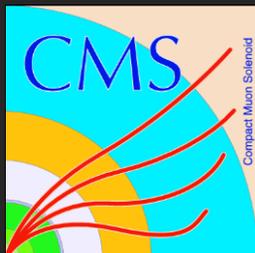
Data analysis and Statistics at the LHC

Dr. Nicholas Wardle



VSOP Quy Nhon, Vietnam

15-26 July 2024

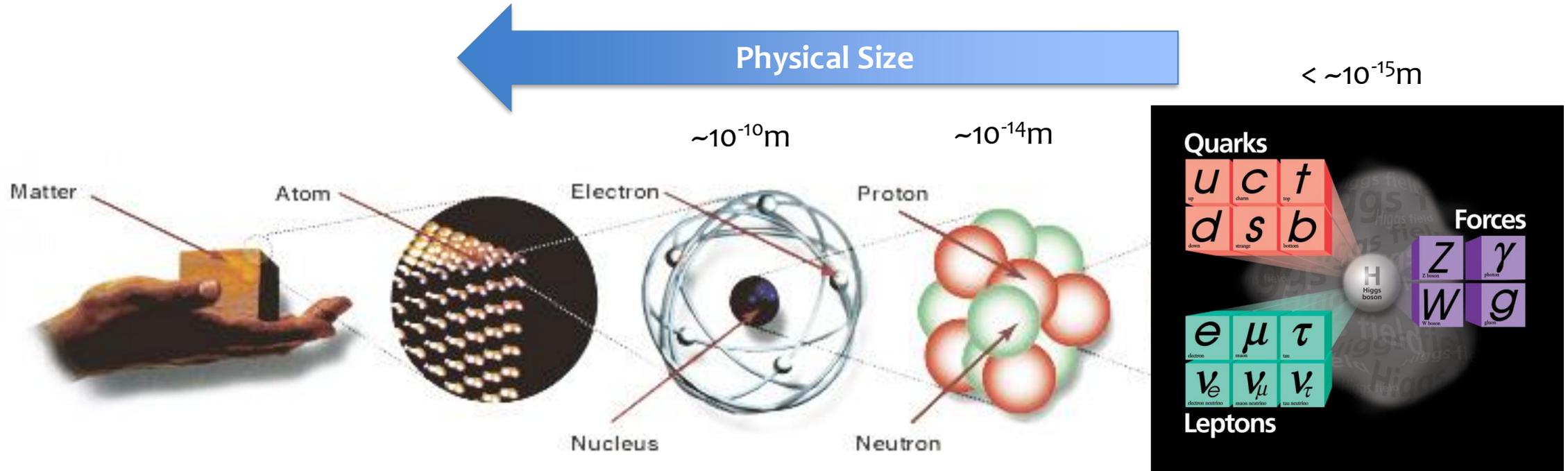


Overview (Exp)

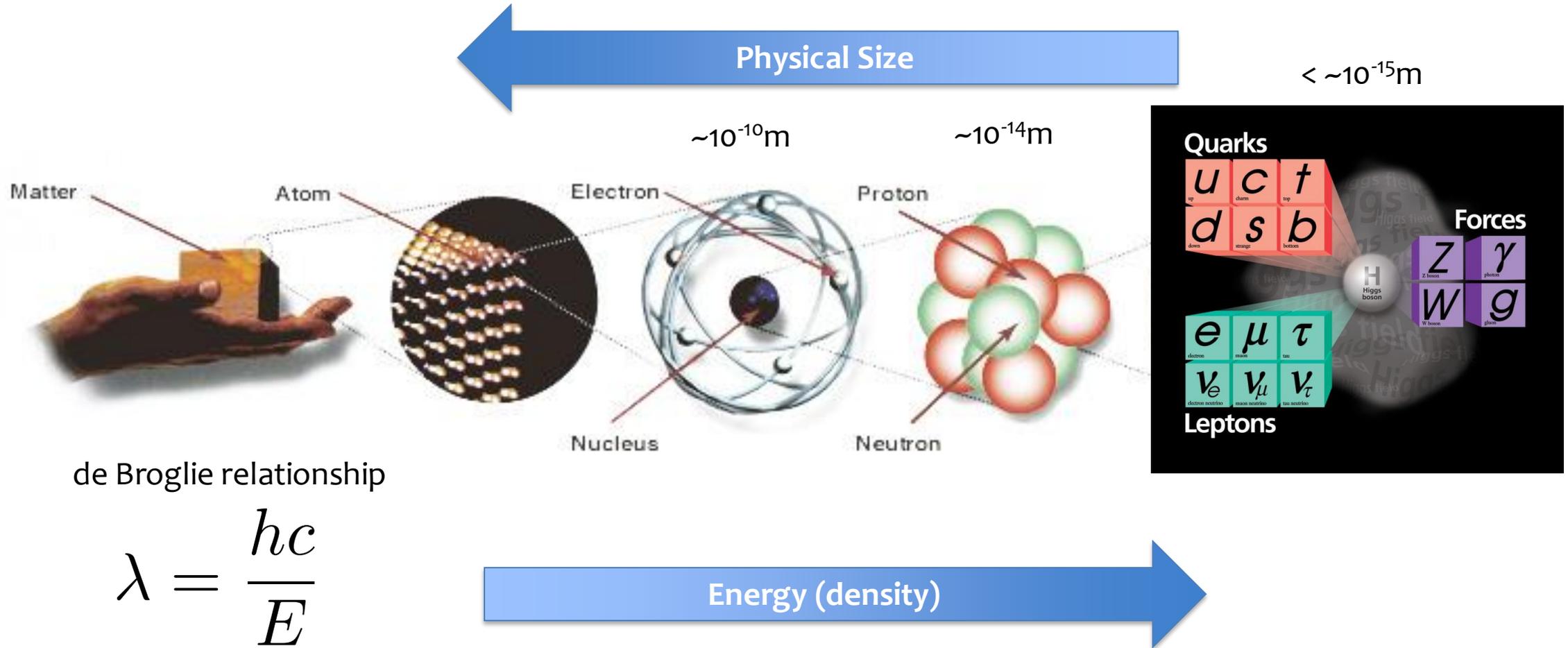
	Morning	Afternoon
Tuesday 16 th	Lecture 1 LHC Data Analysis	Exercise 1
Wednesday 17 th	Lecture 2 LHC Statistics	Exercise 2
Thursday 18 th		Lecture 3 LHC Statistics
Friday 19 th		Exercise 3 (&4)

Slides for today inspired by
Sourabh Dube (VSOP-28 2022)

Our understanding of matter

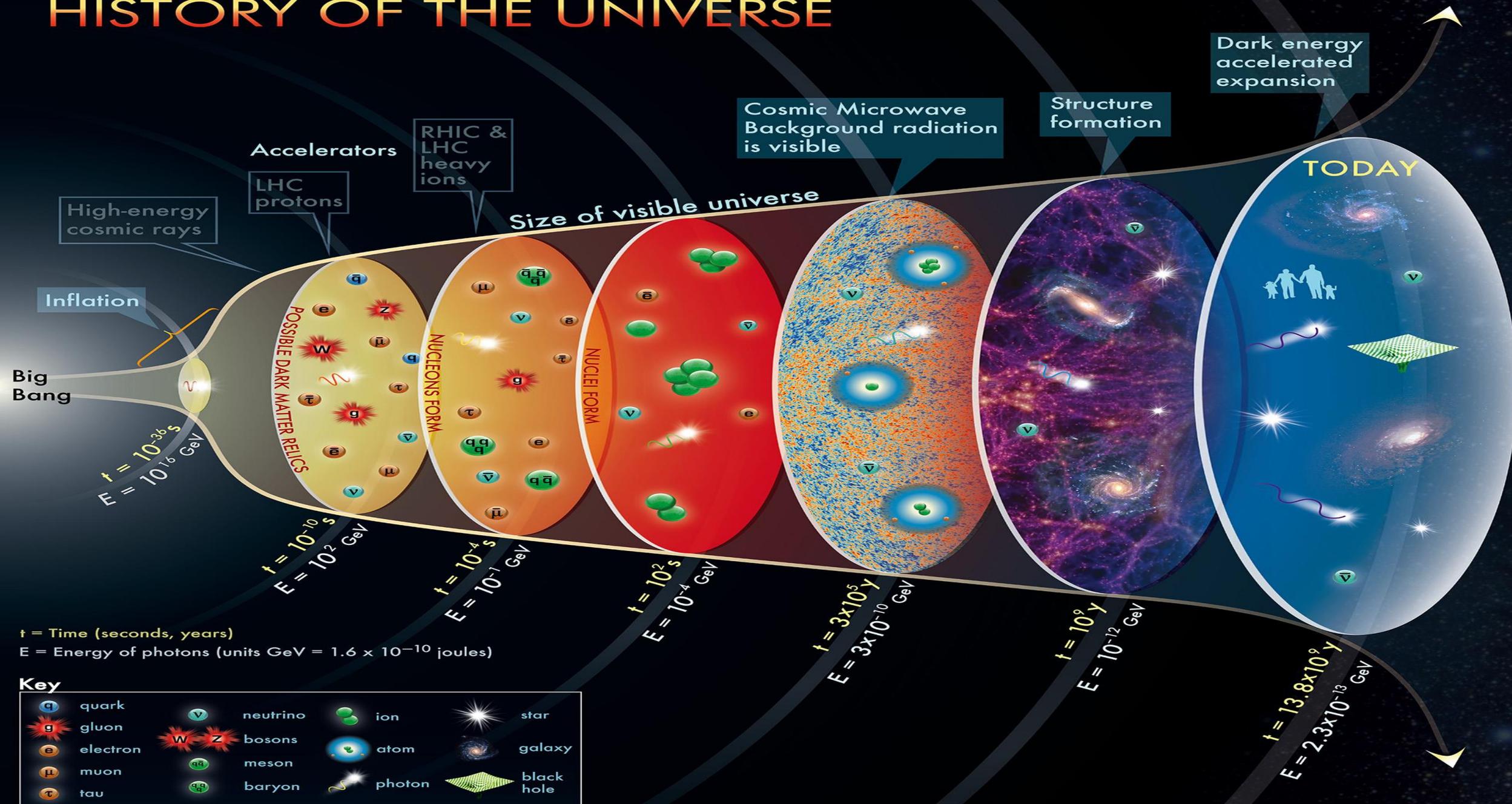


Our understanding of matter



Probing finer structure requires higher energy densities → Particle Collisions probe fine structure of Nature

HISTORY OF THE UNIVERSE



t = Time (seconds, years)
 E = Energy of photons (units GeV = 1.6×10^{-10} joules)

Key

quark	neutrino	ion	star
gluon	bosons	atom	galaxy
electron	meson	black hole	
muon	baryon		
tau	photon		

The concept for the above figure originated in a 1986 paper by Michael Turner.

The LHC



The Large Hadron Collider at CERN is a fundamental physics experiment!

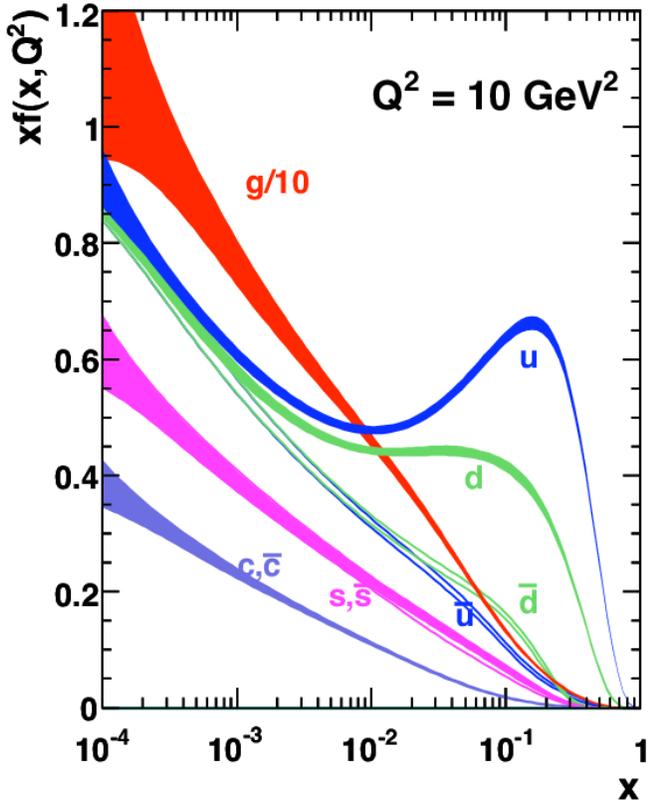
- 27 km in circumference
- 100m underground
- Accelerates protons to 99.9999991% x speed of light
- Proton circles 11,245 times per second!

At center-of-mass energies of **13.6 TeV**, proton collisions probe **physics around the time of the big-bang!**

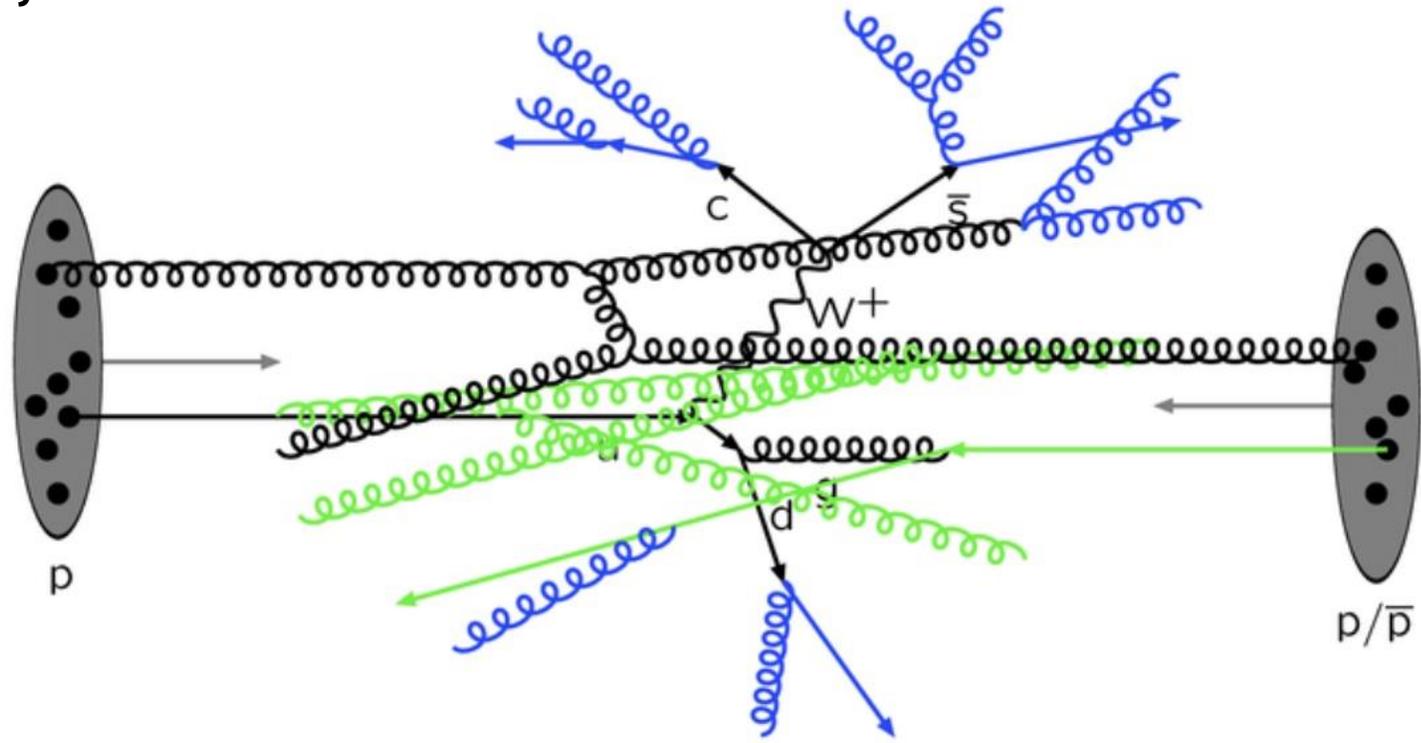
Proton Collisions

Unlike electron-positron colliders, proton collisions are messy but can probe a **huge range of energies simultaneously!**

[F. Betchel](#)

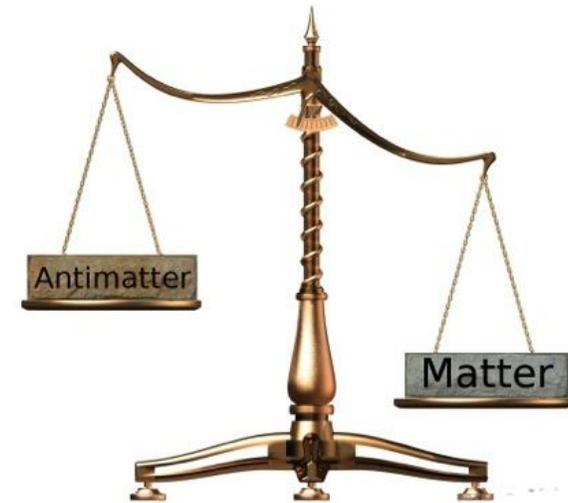
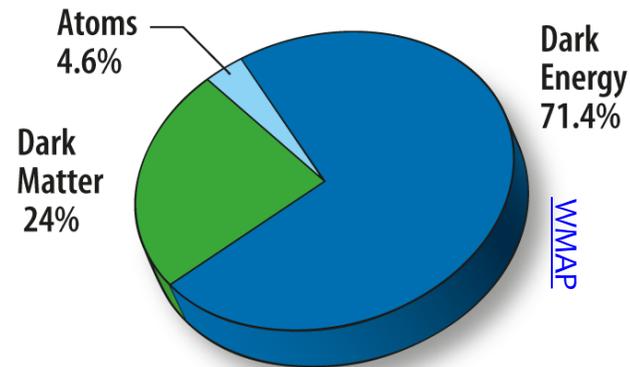
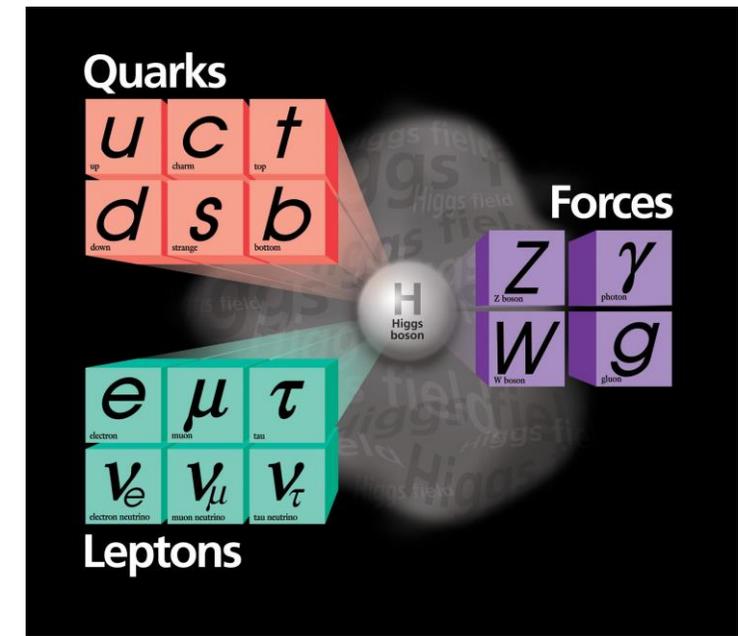
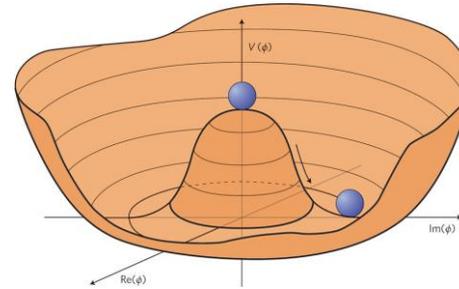


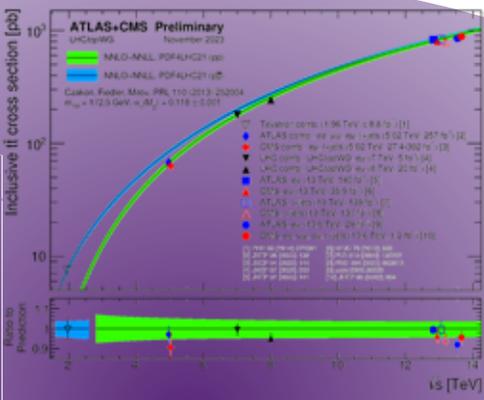
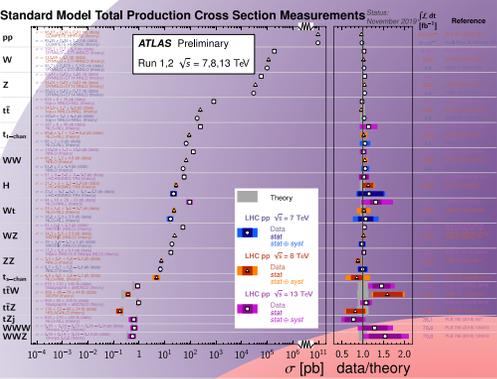
Fraction of proton momentum carried



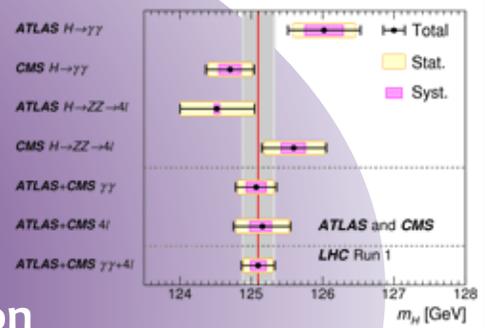
Open questions in Particle Physics

- Is the Higgs sector SM-like ?
- What is Dark Matter (DM)?
- Why is there more matter than anti-matter?
- What is the fundamental nature of neutrinos?
- What is (or is there) a quantum description of gravity?
- ...





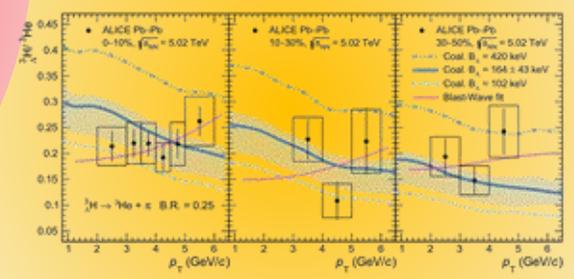
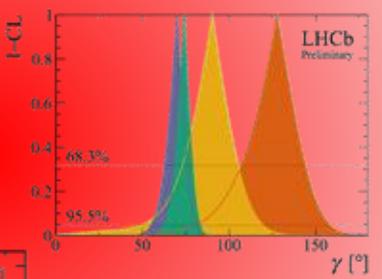
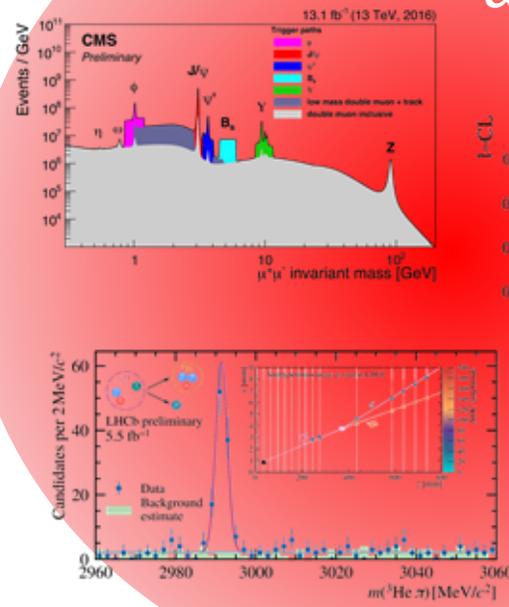
Precision SM/Top/Higgs



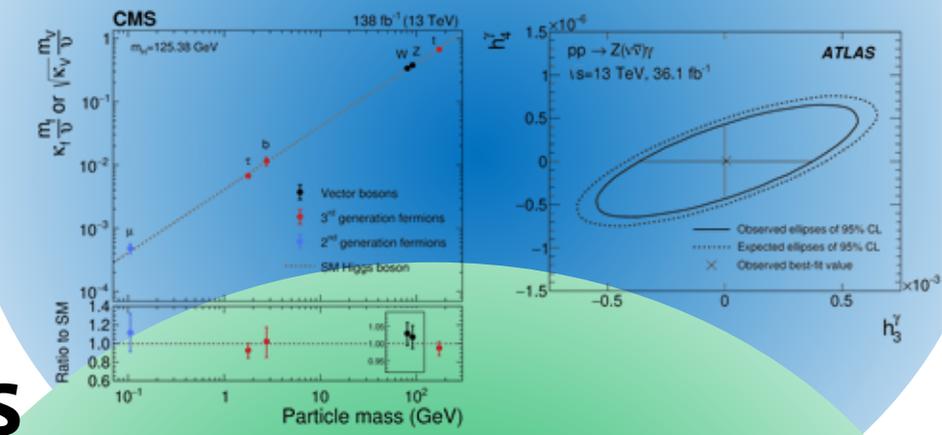
Data analyses at the LHC

Heavy Ions

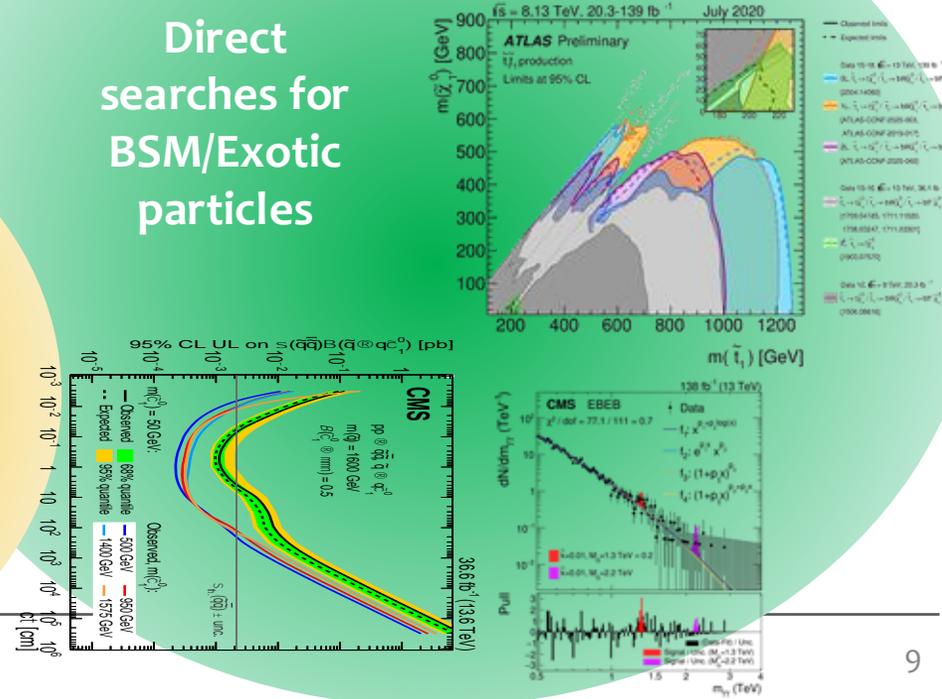
Spectroscopy & Flavour Physics



Indirect searches for BSM



Direct searches for BSM/Exotic particles

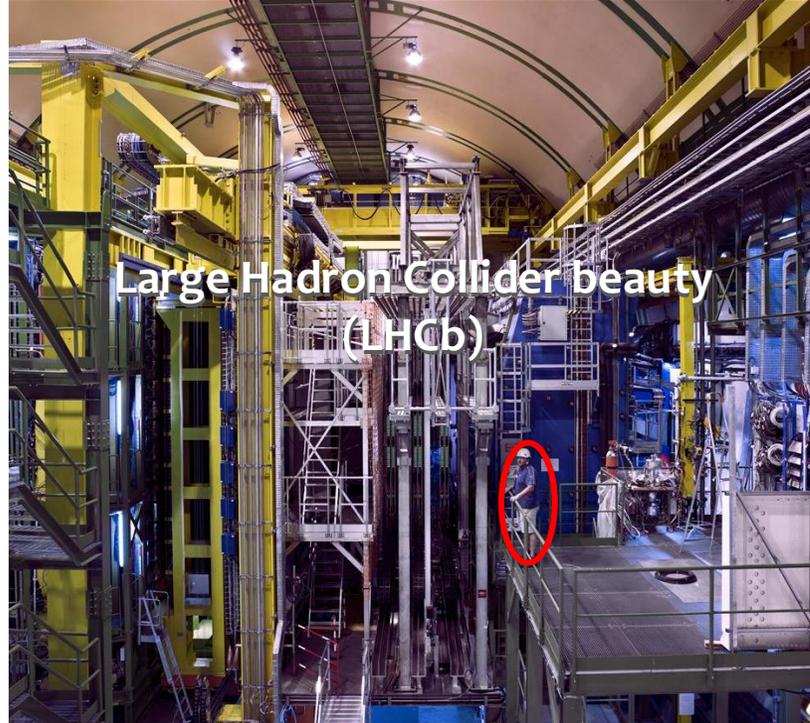


ATLAS and CMS are the two **General Purpose Detectors** at the LHC

LHCb optimized for **flavour** physics and **ALICE** optimized for Heavy Ion collisions

Each is designed to detect the products that are produced in the proton-proton collisions

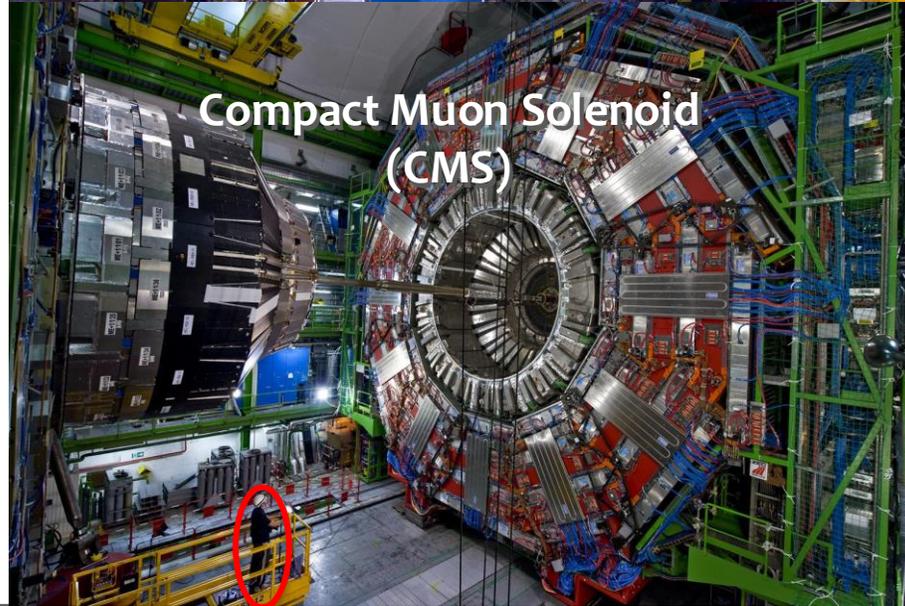
Extremely large-scale machines are required to reconstruct the microscopic events



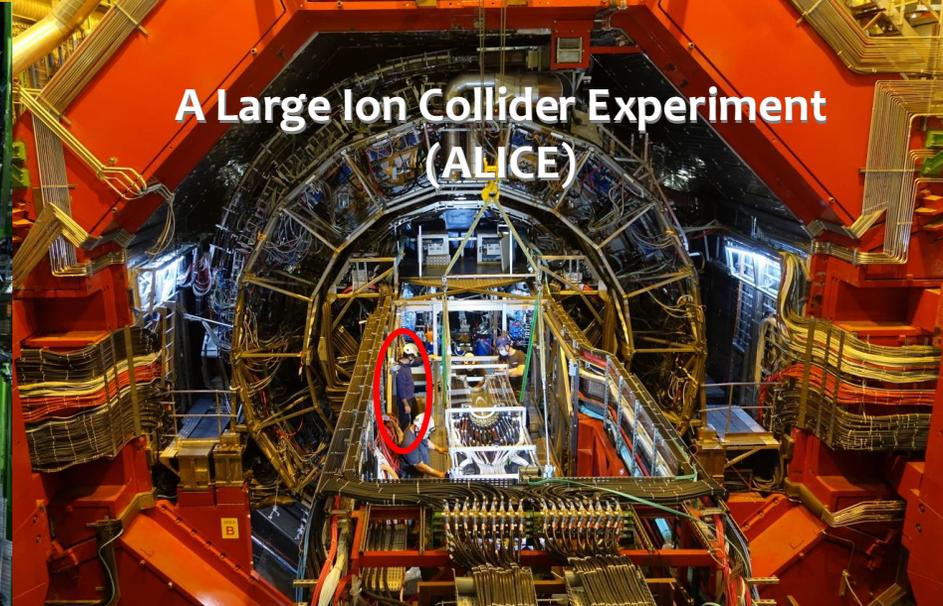
Large Hadron Collider beauty (LHCb)



A Toroidal LHC Apparatus (ATLAS)



Compact Muon Solenoid (CMS)



A Large Ion Collider Experiment (ALICE)

Data



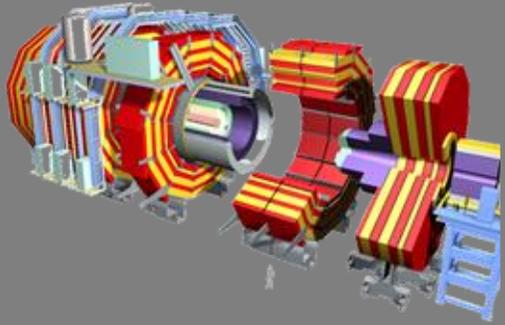
Reconstruction
&
Particle ID

Calibrations

Event Selections
&
Distributions

RESULTS

Data



Reconstruction
&
Particle ID

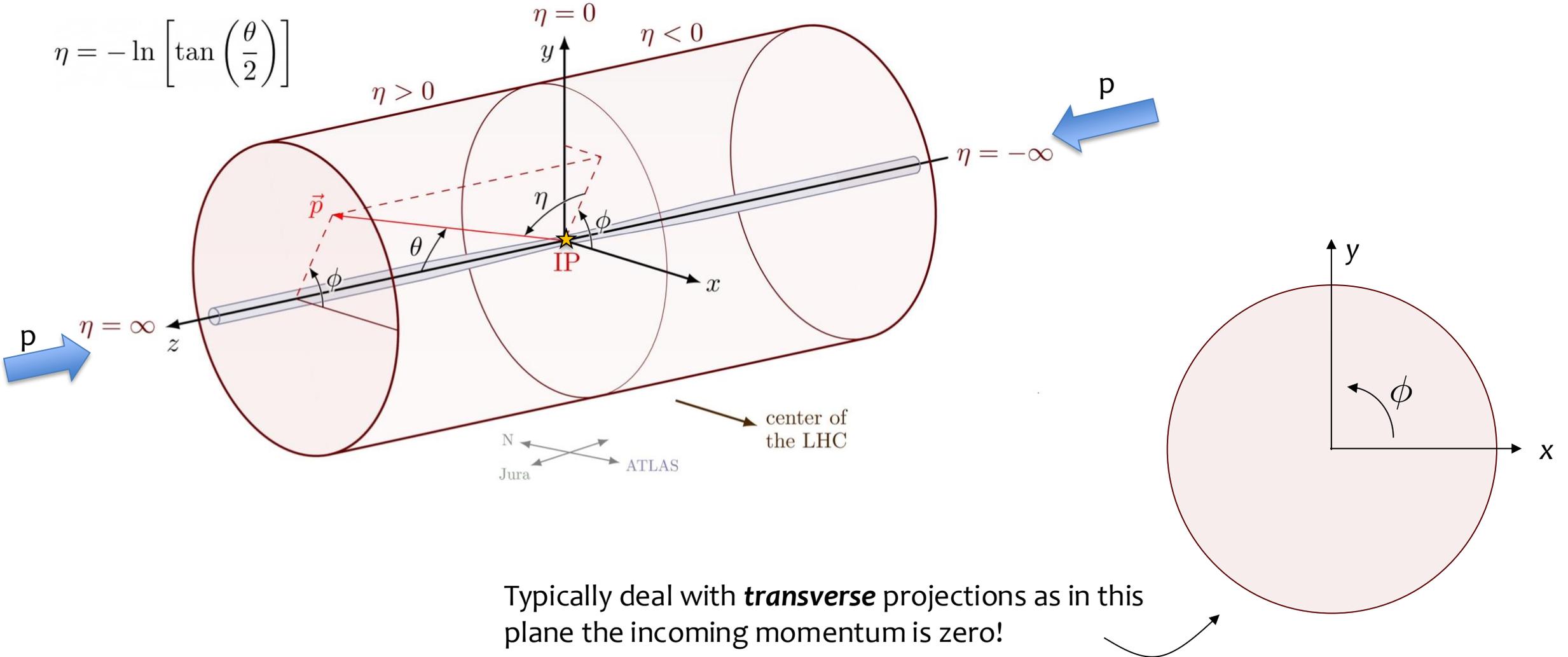
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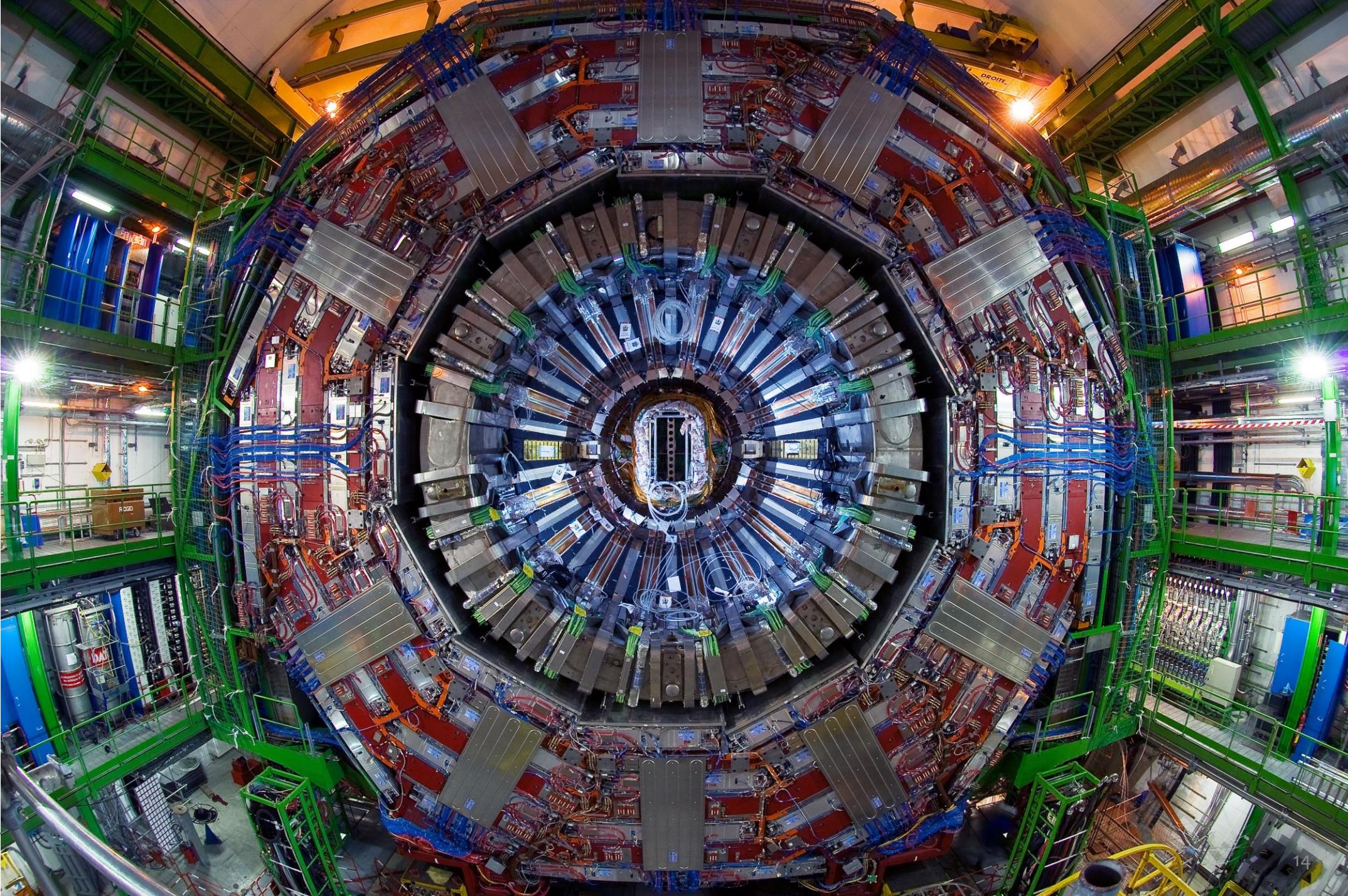
Event Selections
&
Distributions

RESULTS

Co-ordinate system

Co-ordinate system chosen around design of detector & collision system



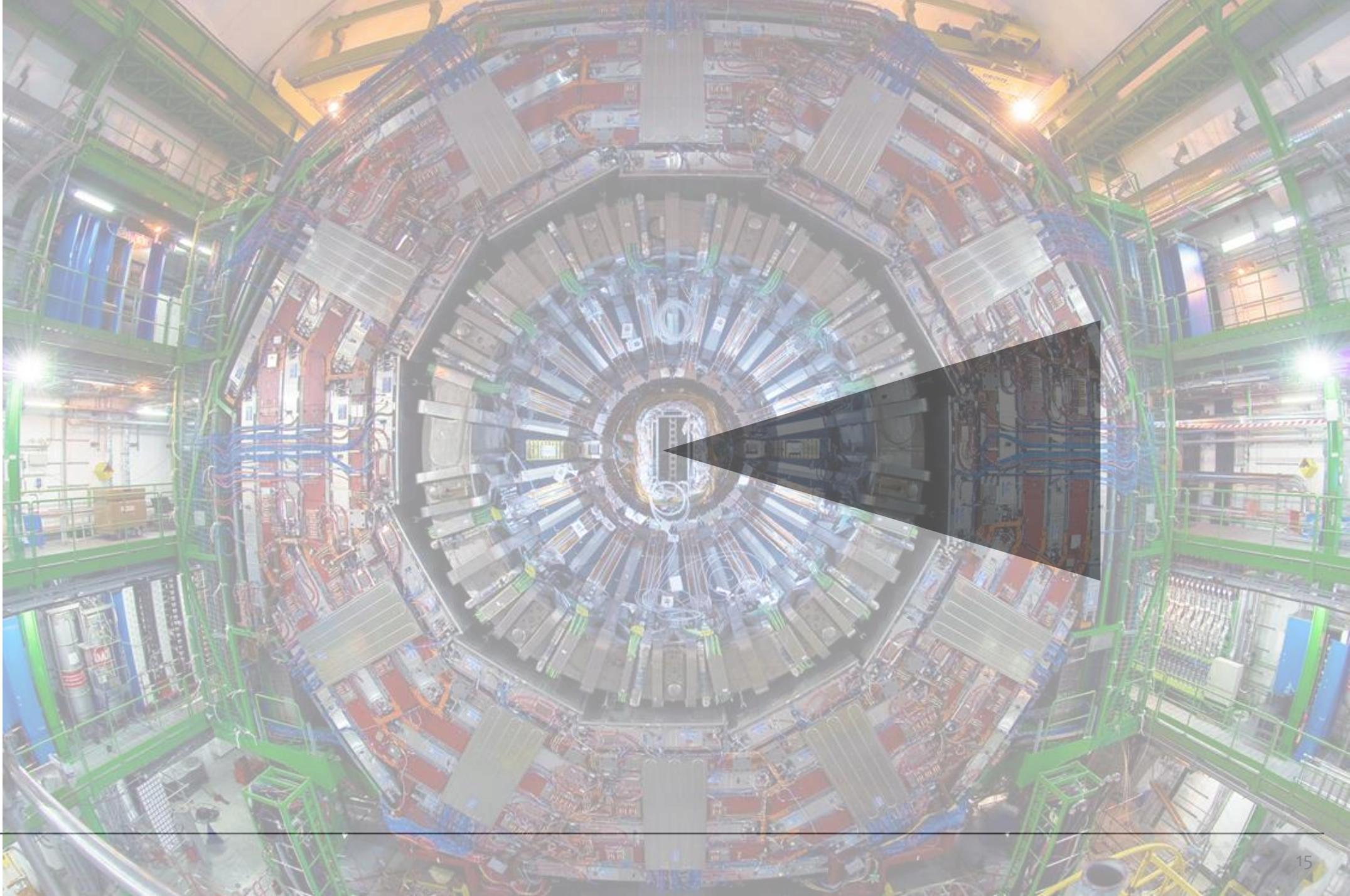


The CMS Detector



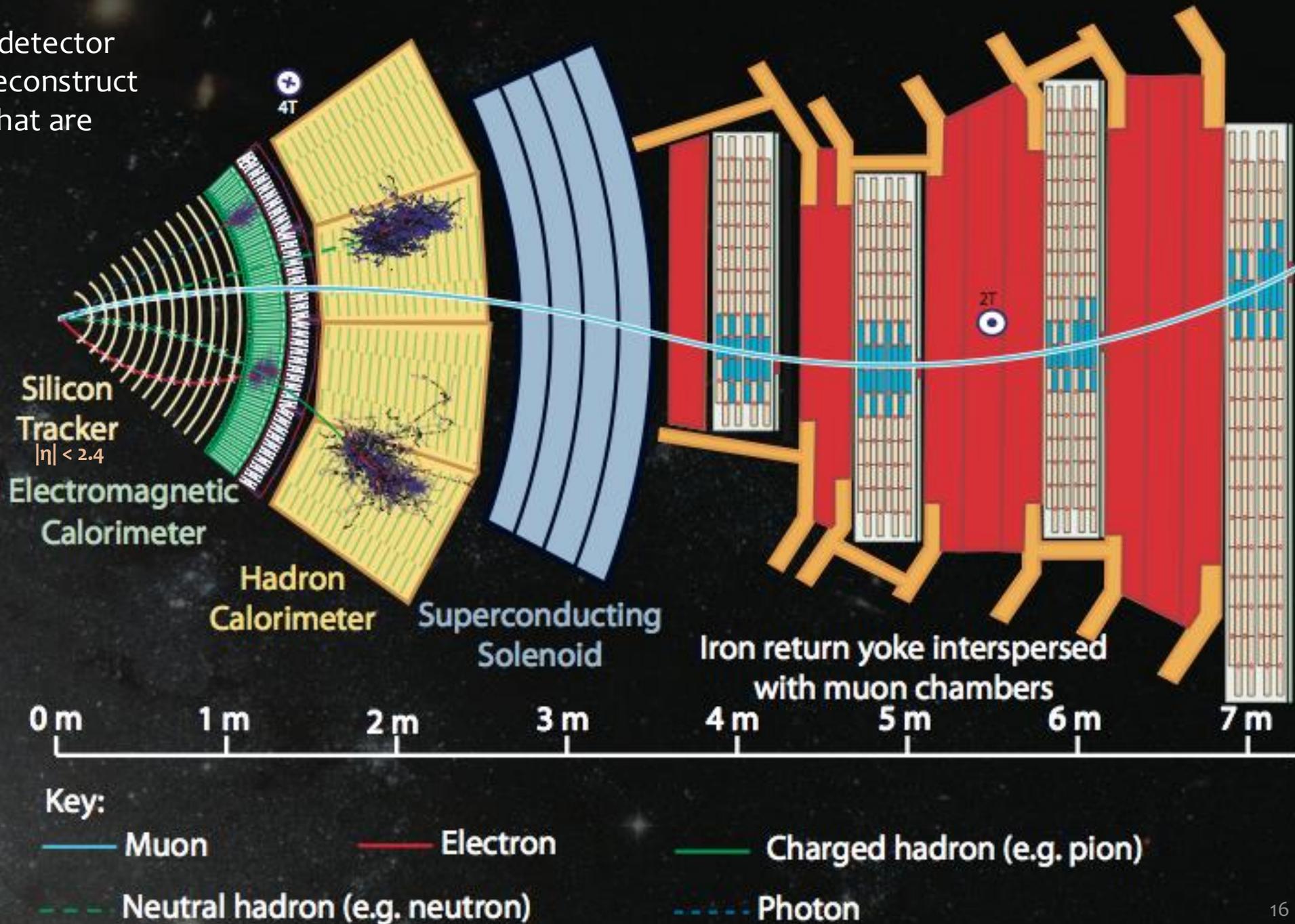


The CMS Detector

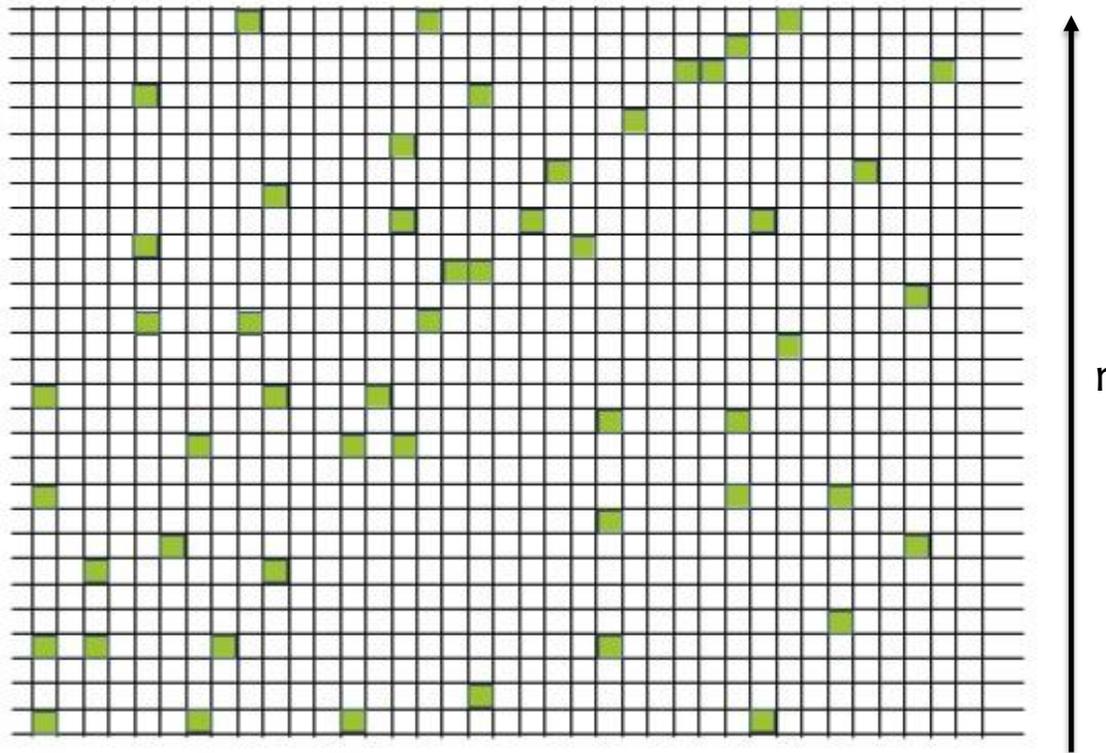
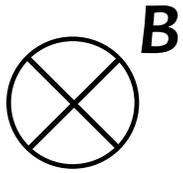


Different elements of the detector designed to identify and reconstruct different stable particles that are produced

The CMS Detector



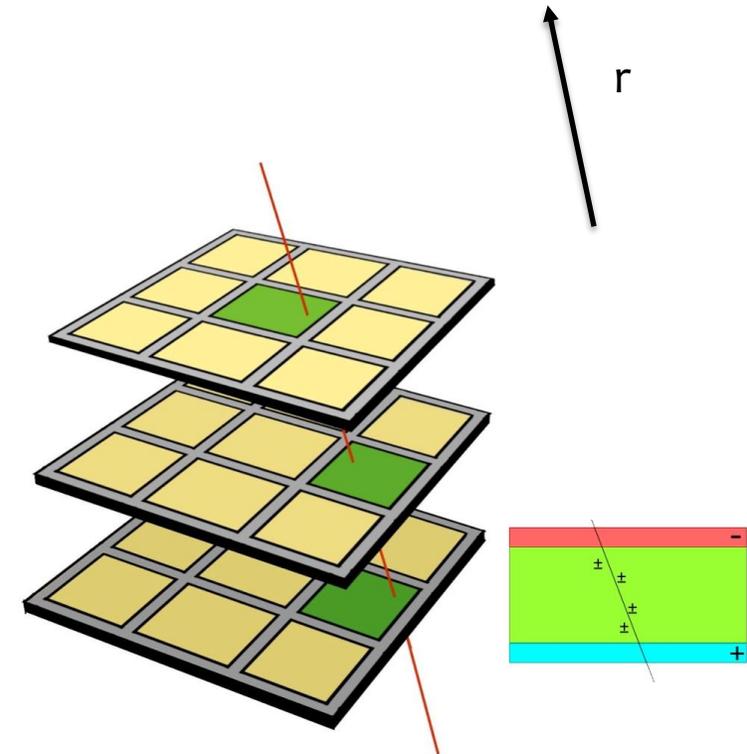
Forming Tracks



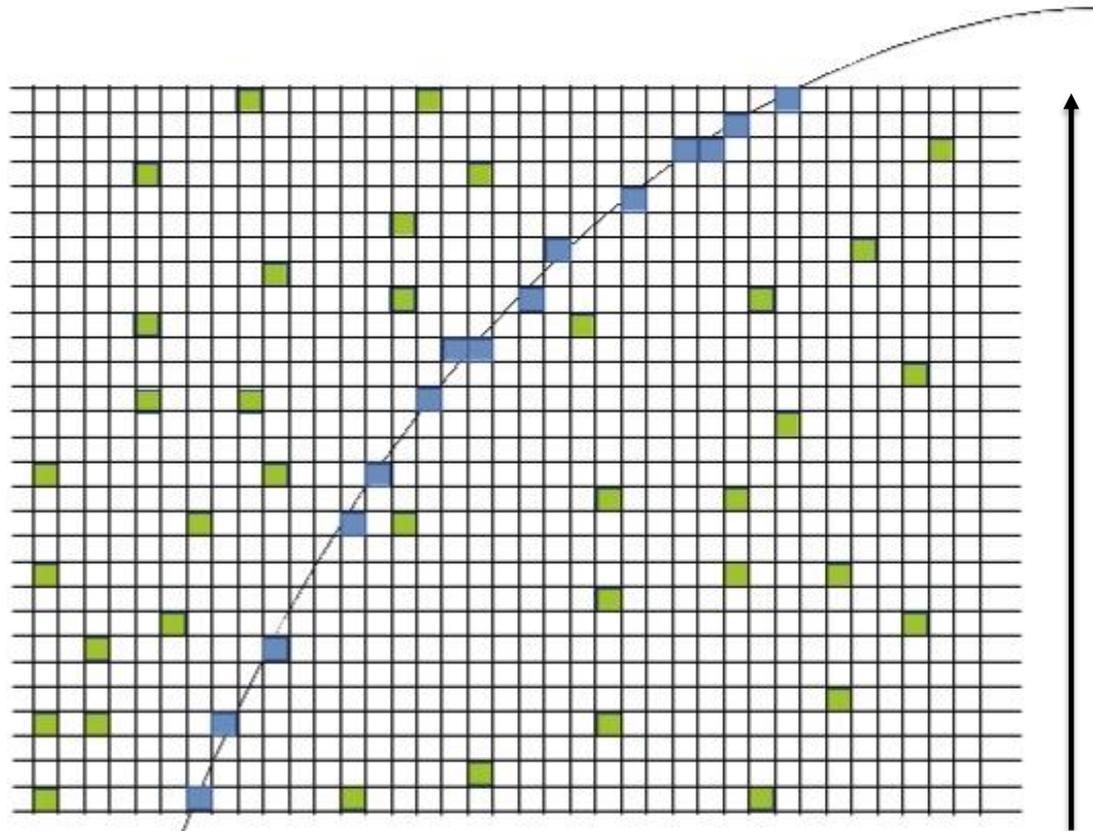
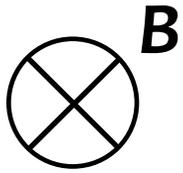
Charged particles travelling through silicon track layers (pixels/strips) will create electrons / hole pairs

→ Electrons drift where charge can be read-out

→ Localized “hits” in the tracker layer



Forming Tracks

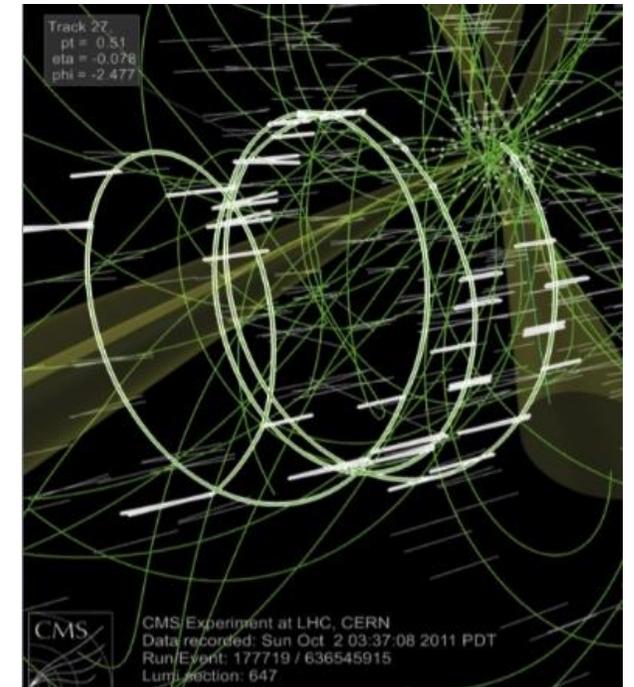


$$\mathbf{F} = q\mathbf{v} \times \mathbf{B}$$

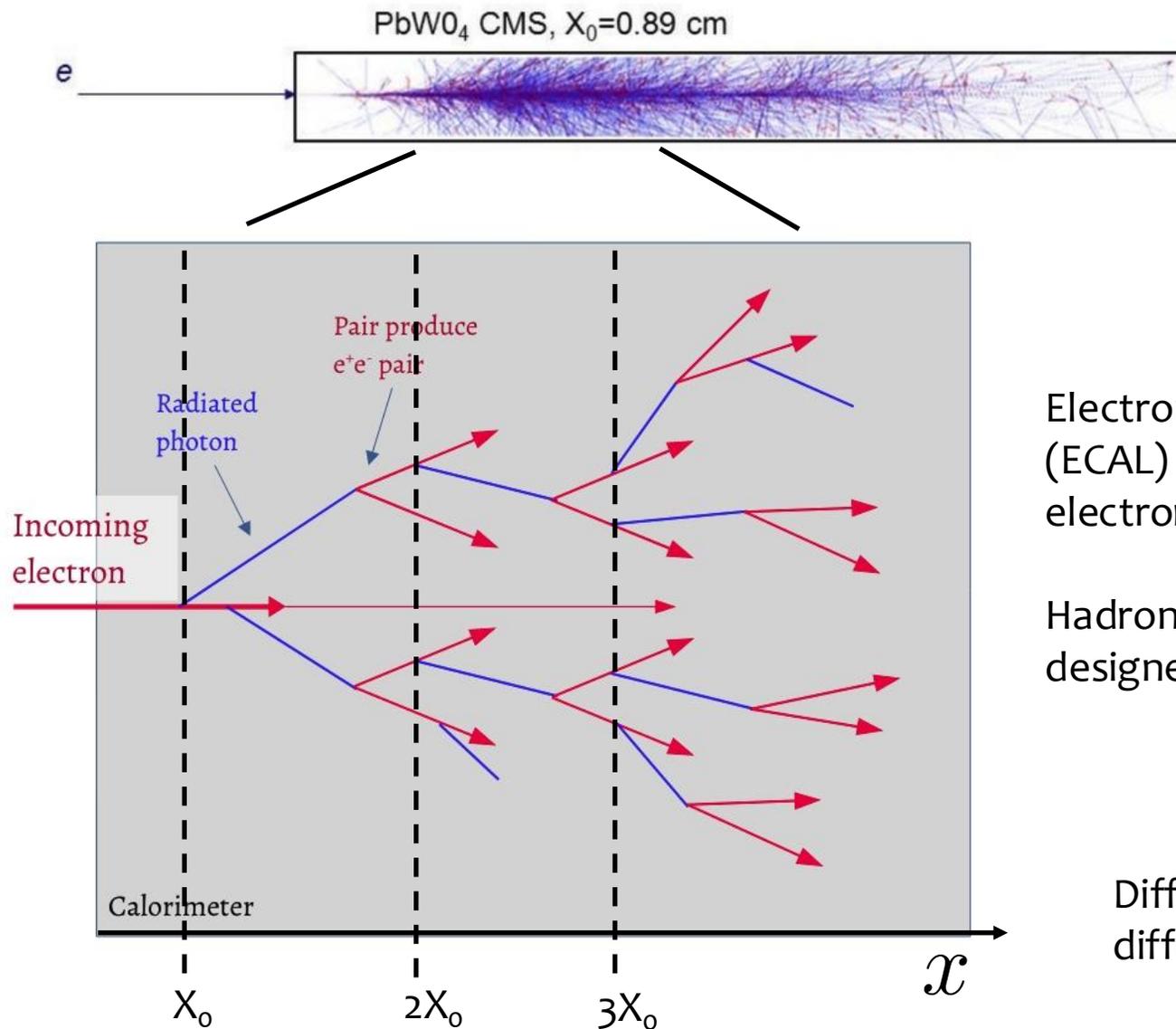


Tracking algorithm combines **hits along path** → track is formed!

- Radius of curvature → momentum
- Charge ID from direction of bending
- Angles of trajectory wrt beamline
- Impact parameters (offset wrt interaction point)



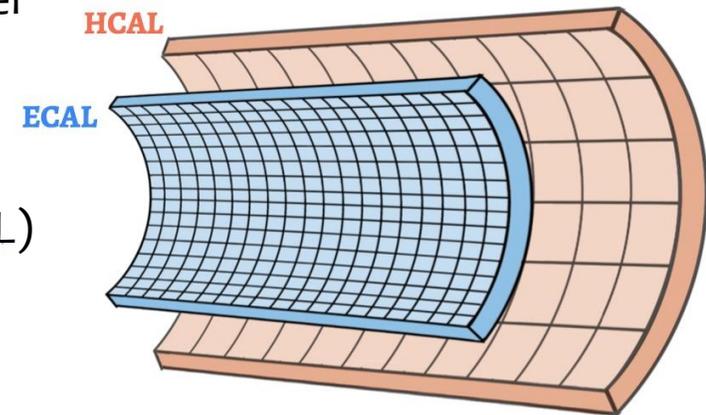
Calorimetry



Calorimeter layers are designed to absorb particle energy: E.g electron bremsstrahlung in ECAL / pair production produces showers which evolve through calorimeter material

Electromagnetic Calorimeter (ECAL) designed to stop electrons/photons

Hadronic Calorimeter (HCAL) designed to stop hadrons

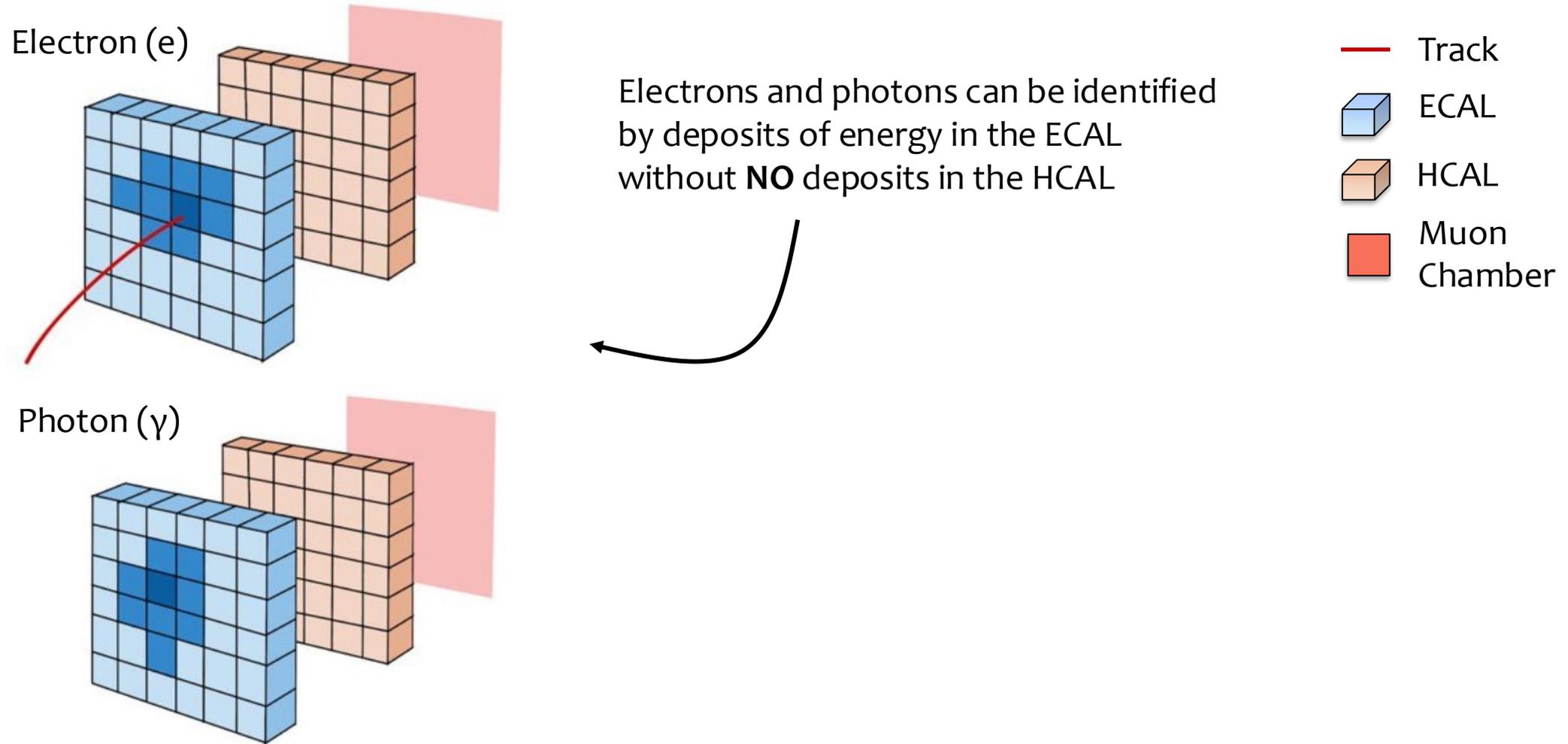


Different materials have different radiation lengths (X₀)

$$-\frac{dE}{dx} = \frac{E}{X_0}$$

Calorimetry

Remember that different components of our detector will respond differently to different particles

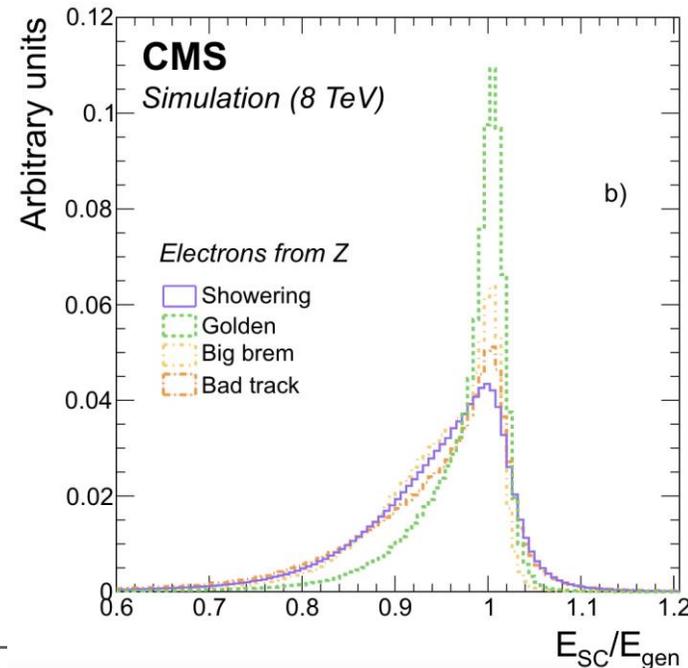
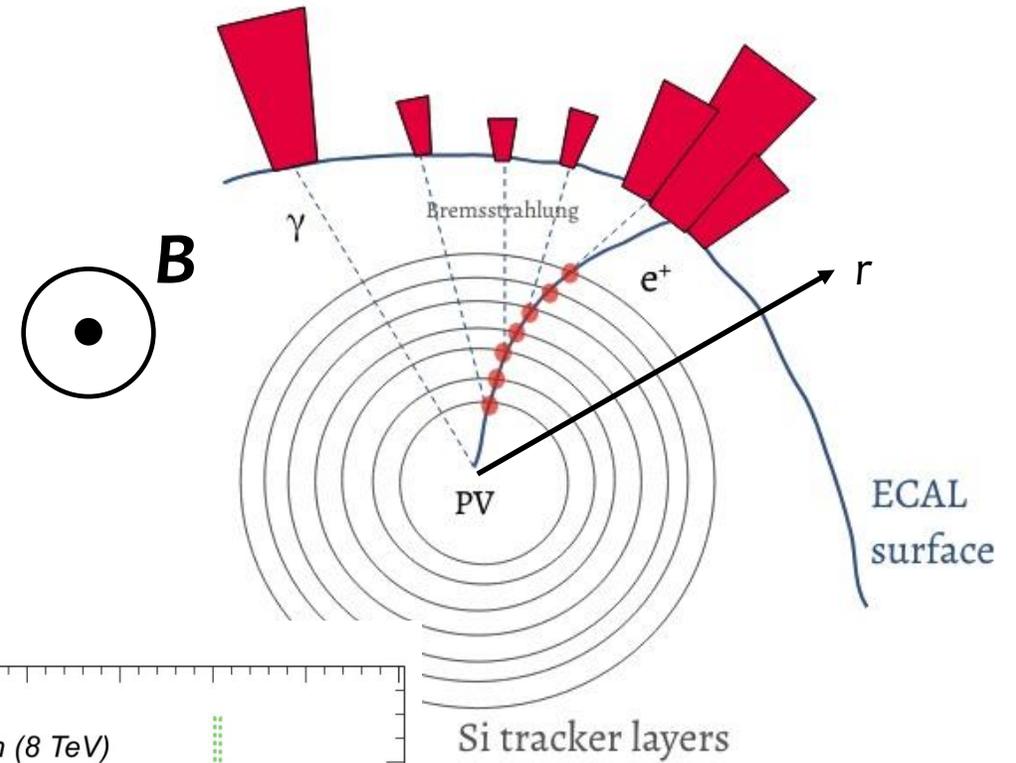
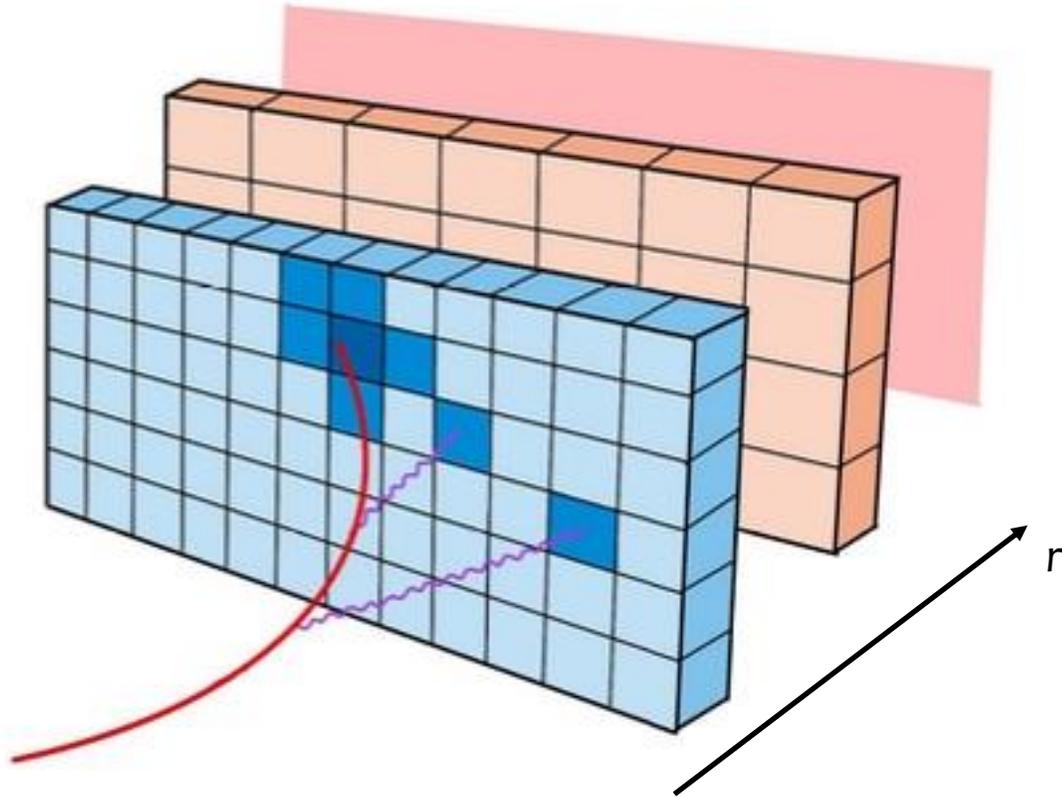


“Super clustering”

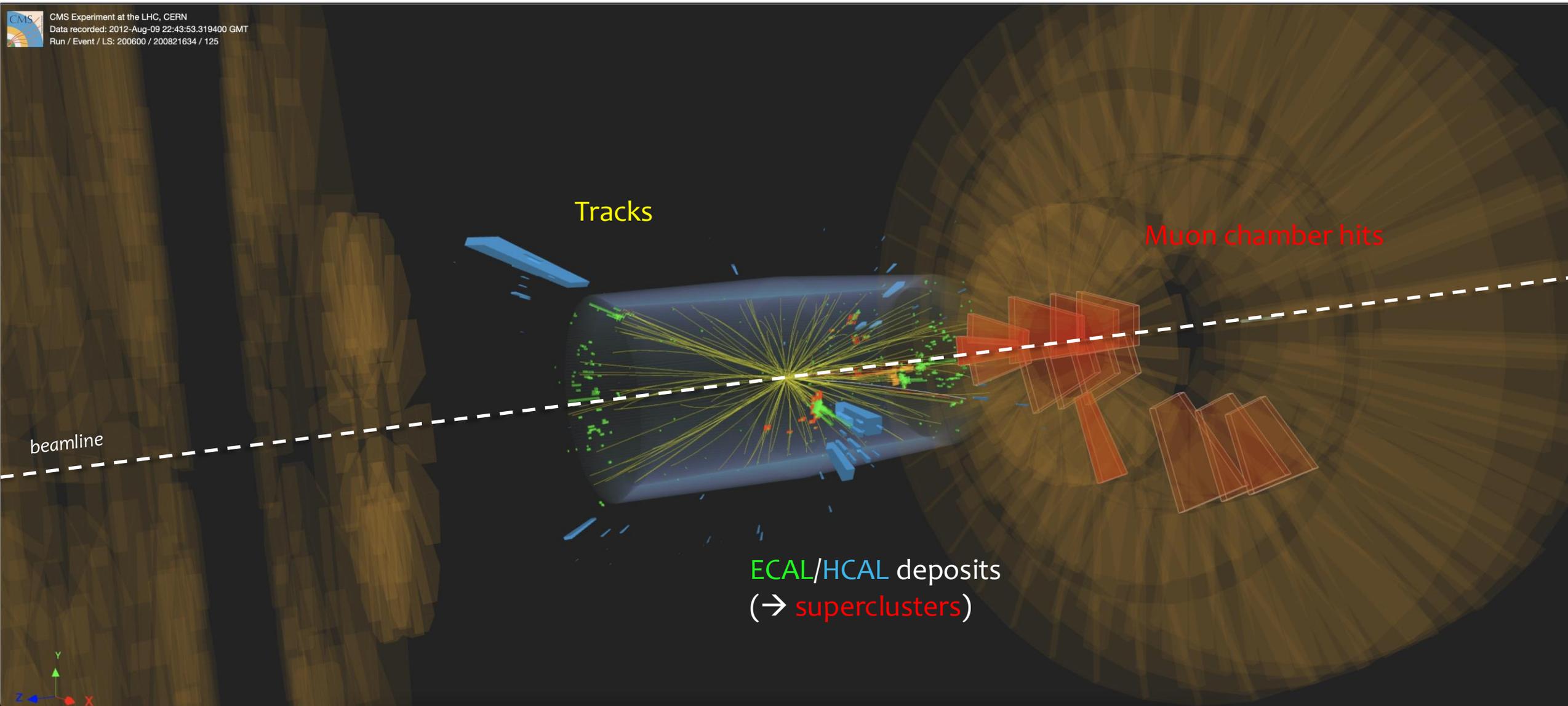
Electrons bend in the presence of a magnetic field B

→ Radiation from acceleration of charged particle

→ Photons must be included in reconstruction of electrons to maintain a good energy measurement

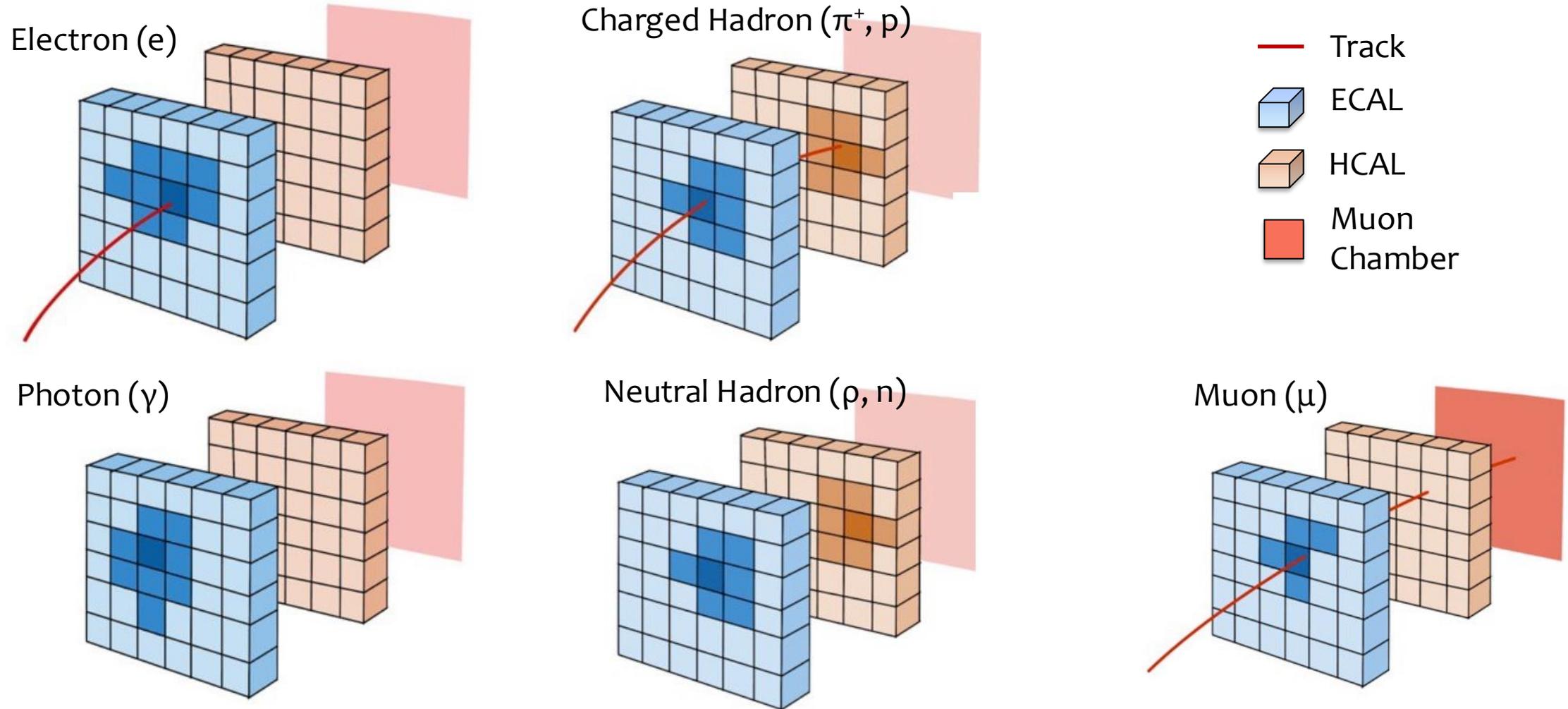


In each collision, the detector components measure energy deposits forming hits / tracks



Calorimeters

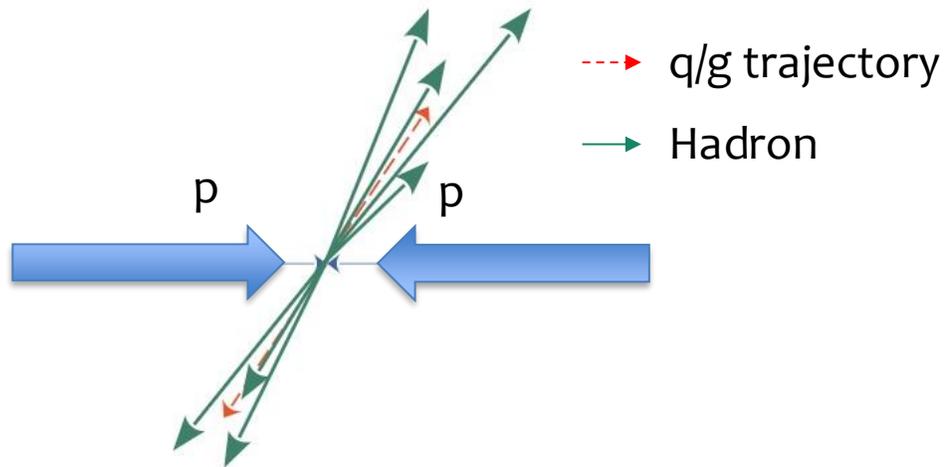
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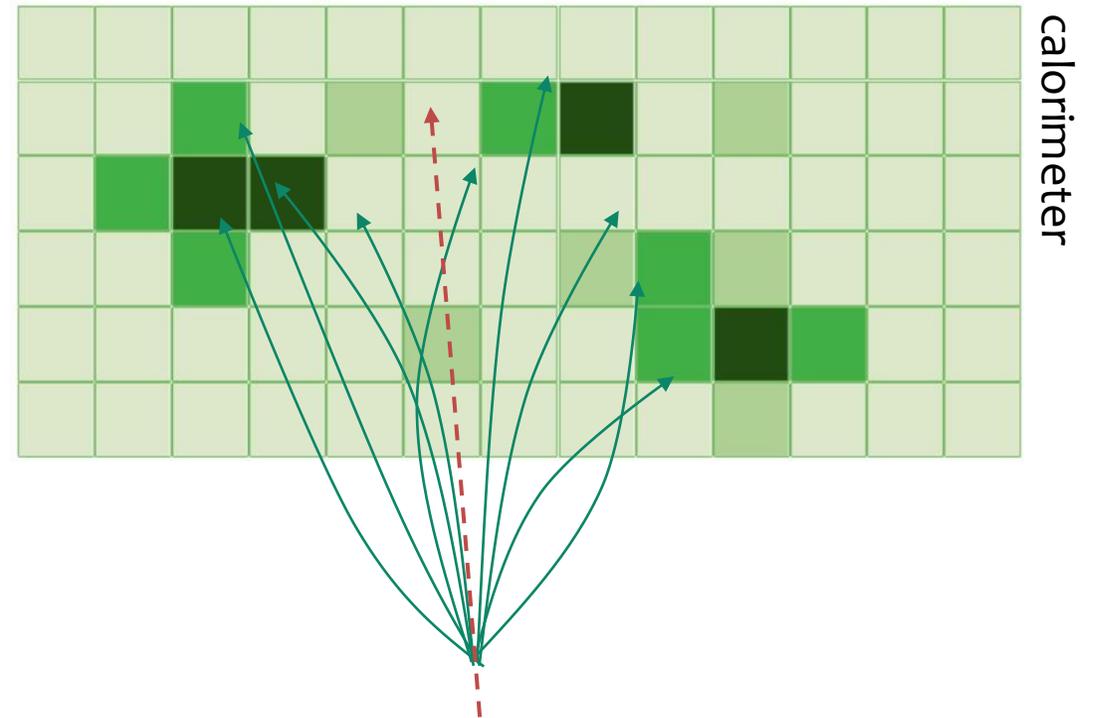
Jet Clustering

Coloured particles (quarks & gluons) produced in proton collisions do not reach the detector components

→ Part of the production energy/momentum is used to produce additional quark/antiquark pairs – which then form hadrons. It is the hadrons that exist/escape from the collision and can be detected



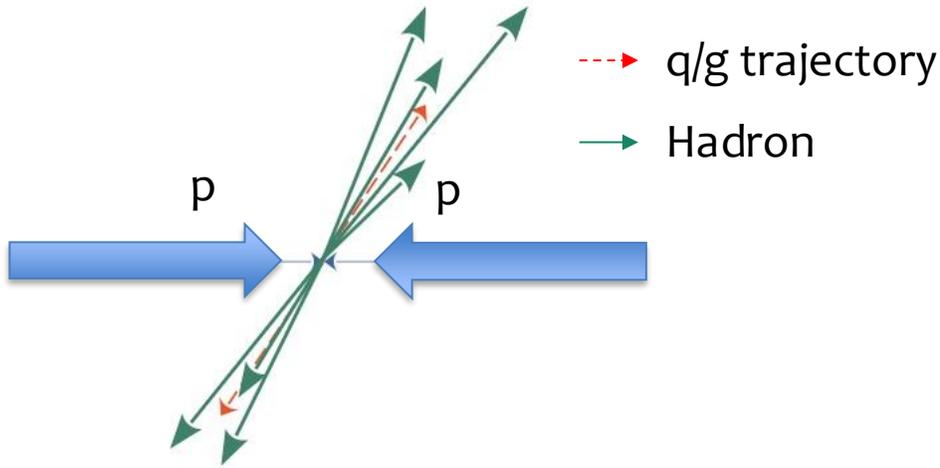
How can we determine energy & momentum of the original coloured particle?



Jet Clustering

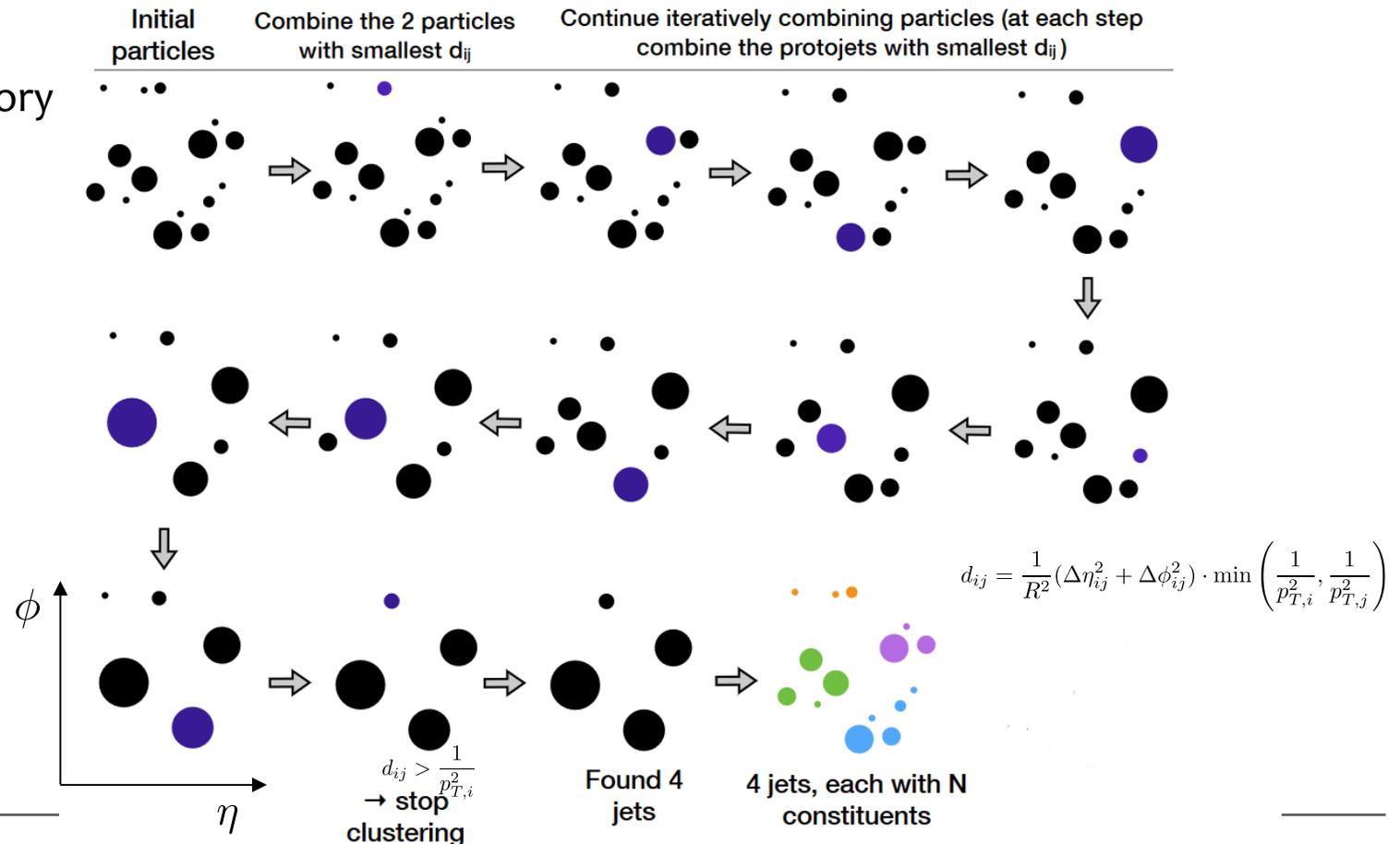
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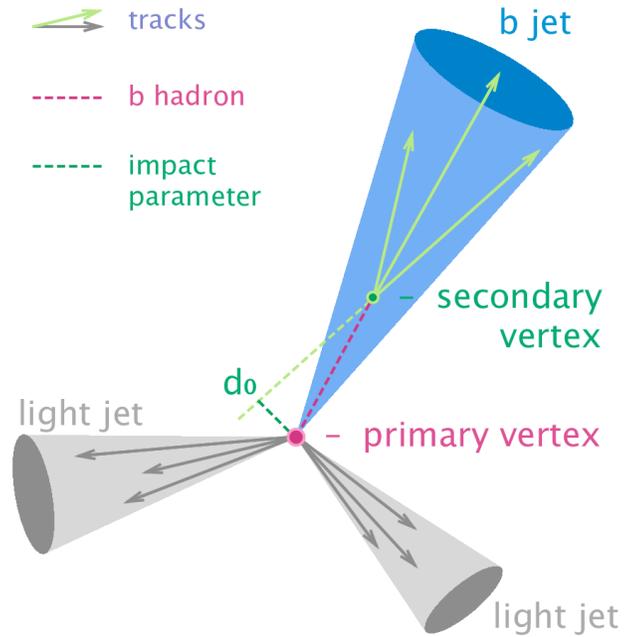
Clustering collects particles* with original quark/gluon into single four-vector using energy-momentum conservation!

*or tracks/energy deposits ...

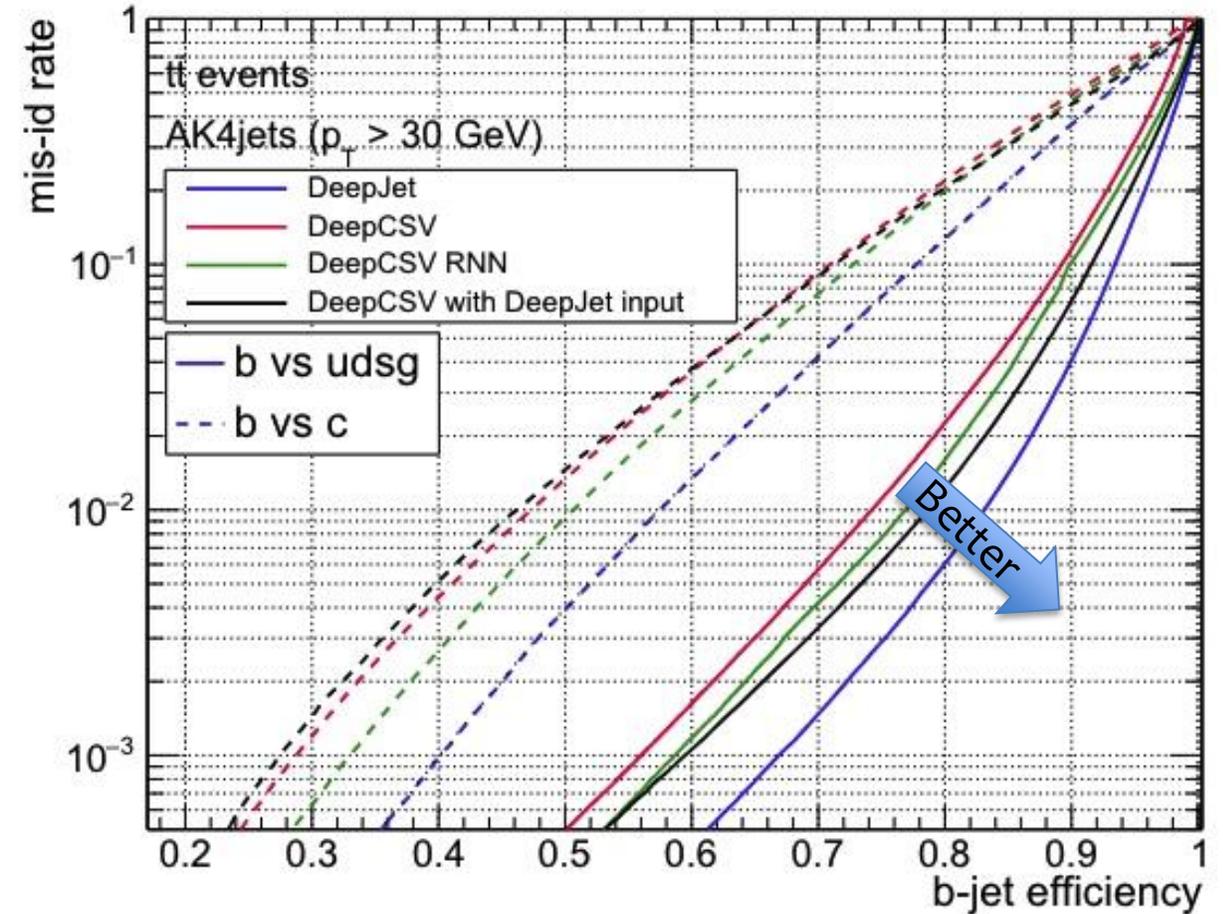
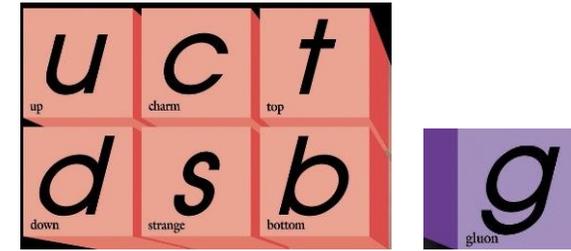


b/c-jet

Identifying which particle initiated each jet requires lots of combined information about the constituents of the jet and the vertices it contains

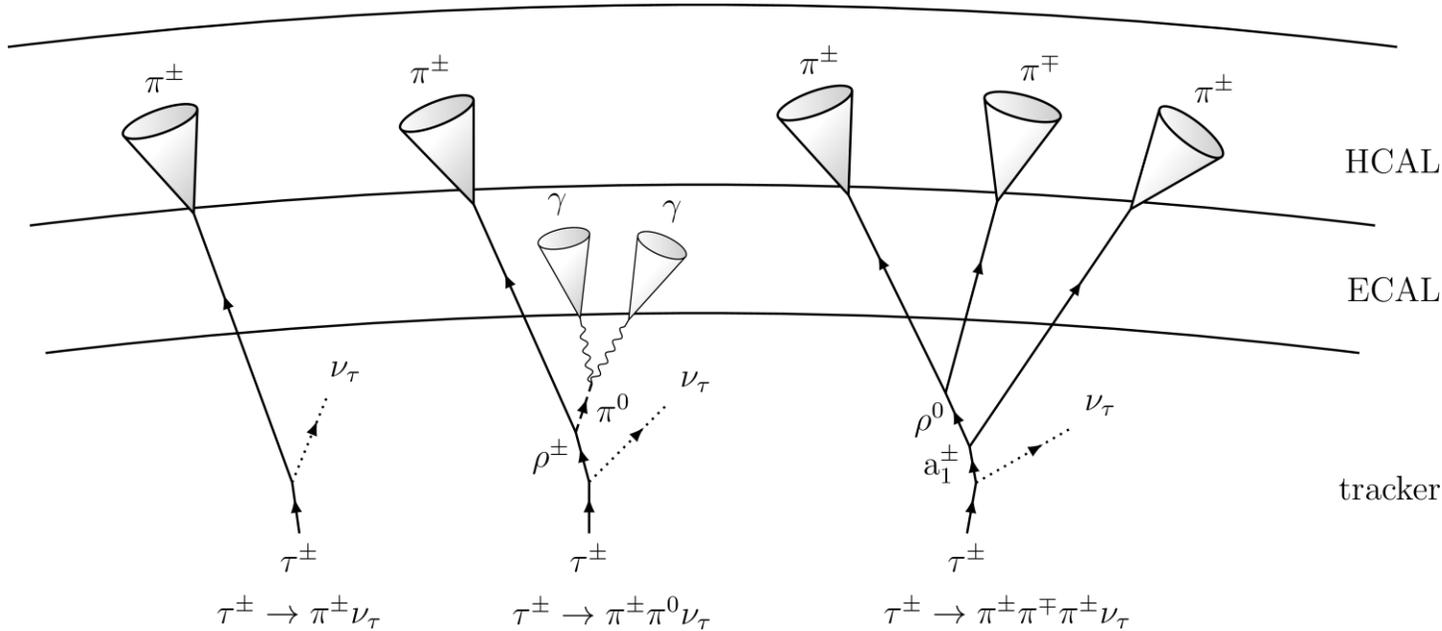


We use sophisticated **machine learning methods** to perform this task

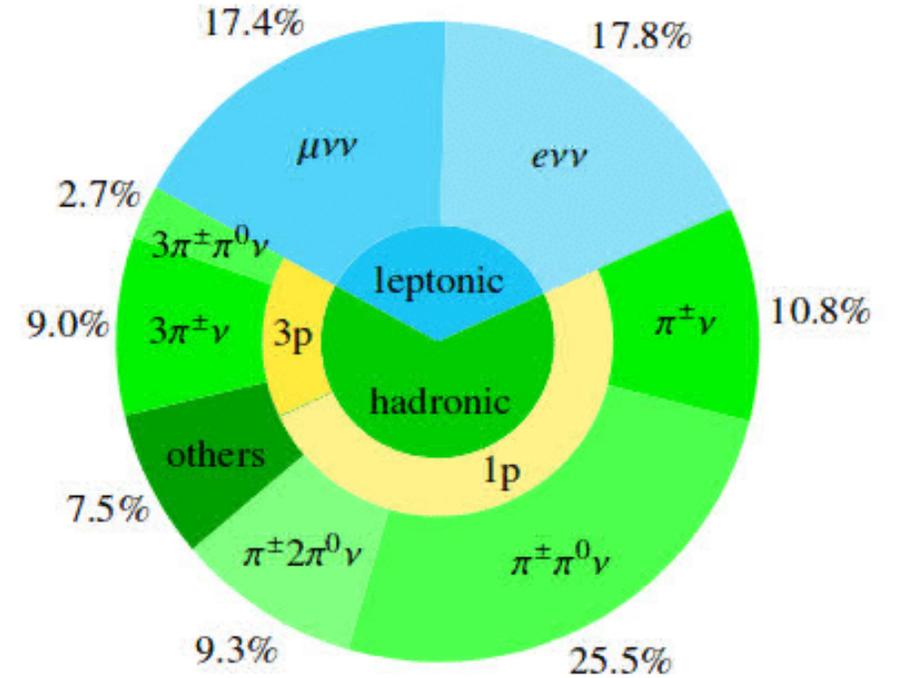


Tau-leptons

τ leptons have very short lifetime \rightarrow they decay into **leptons** or **hadronically**



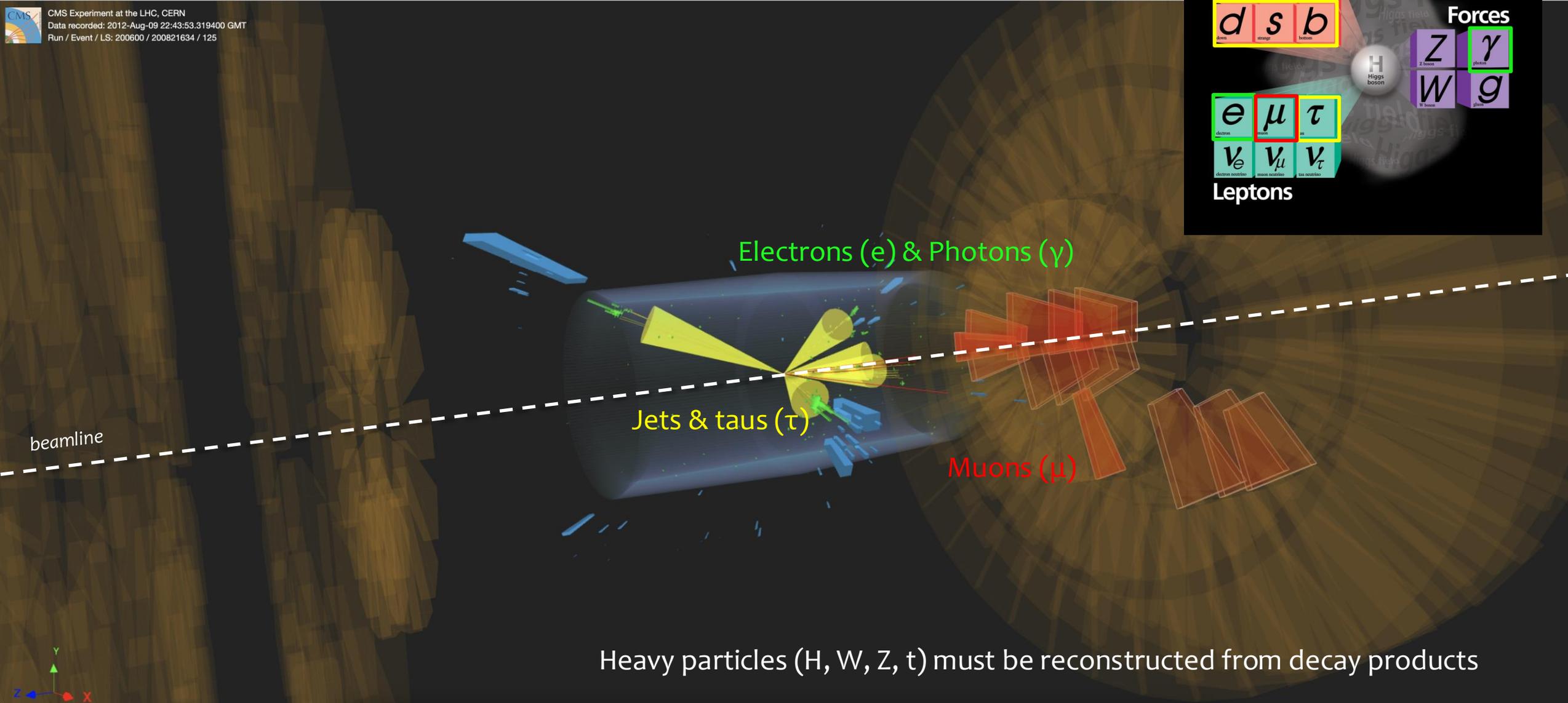
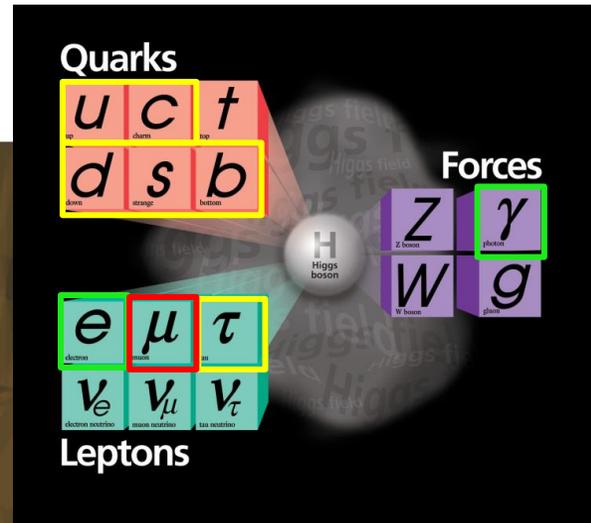
τ decay mode probabilities



Most modern τ -ID strategies use **machine learning** to identify the decay mode and reconstruct the τ four-momentum

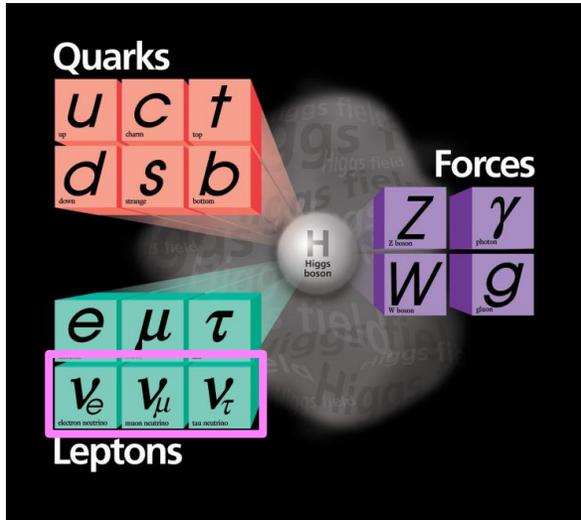
Physics objects are formed by clustering certain tracks & energy deposits

CMS Experiment at the LHC, CERN
Data recorded: 2012-Aug-09 22:43:53.319400 GMT
Run / Event / LS: 200600 / 200821634 / 125



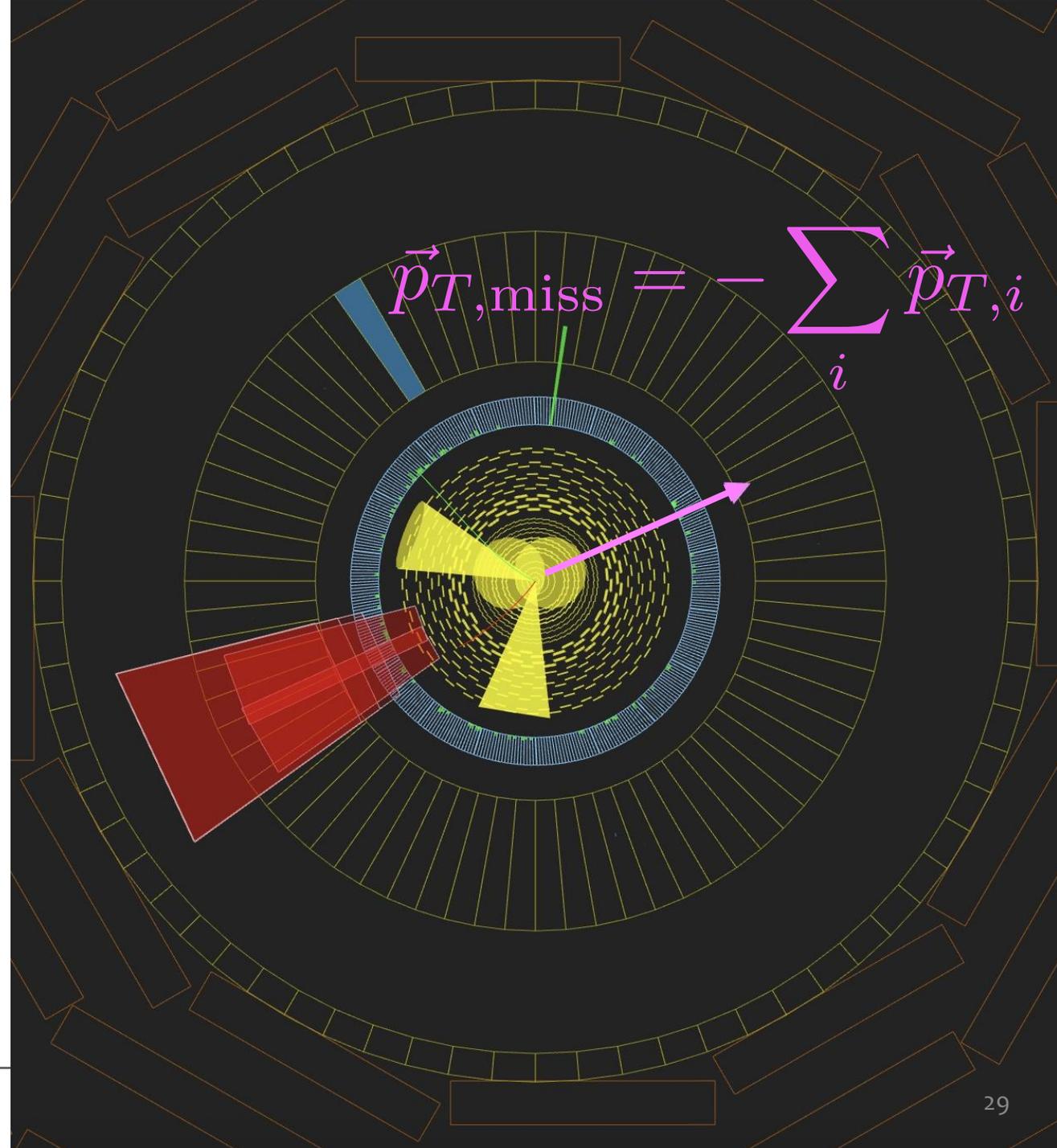
Heavy particles (H, W, Z, t) must be reconstructed from decay products

Missing momentum

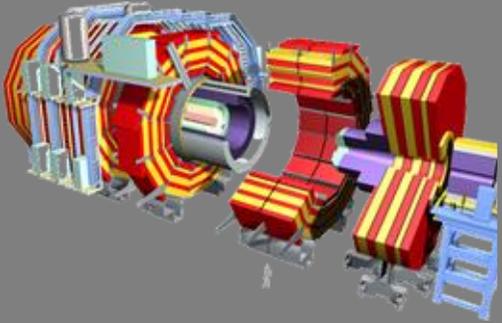


Neutrinos do not interact with any component of the detector

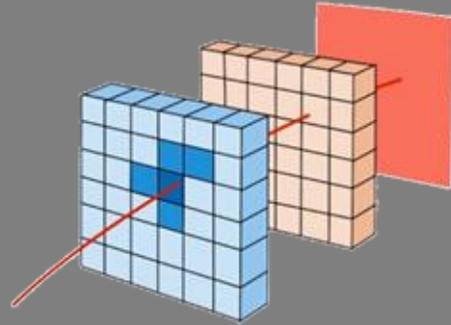
We infer the presence of neutrinos through an imbalance of momentum in the transverse plane
→ missing transverse momentum



Data



Reconstruction
&
Particle ID



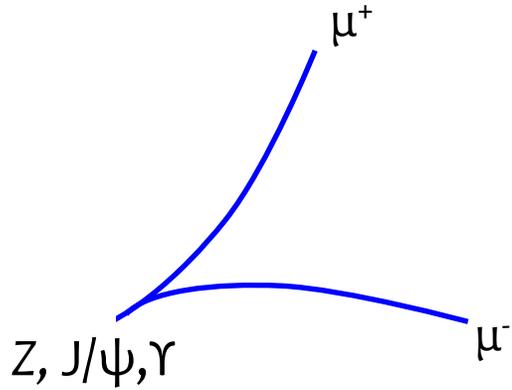
Calibrations

Event Selections
&
Distributions

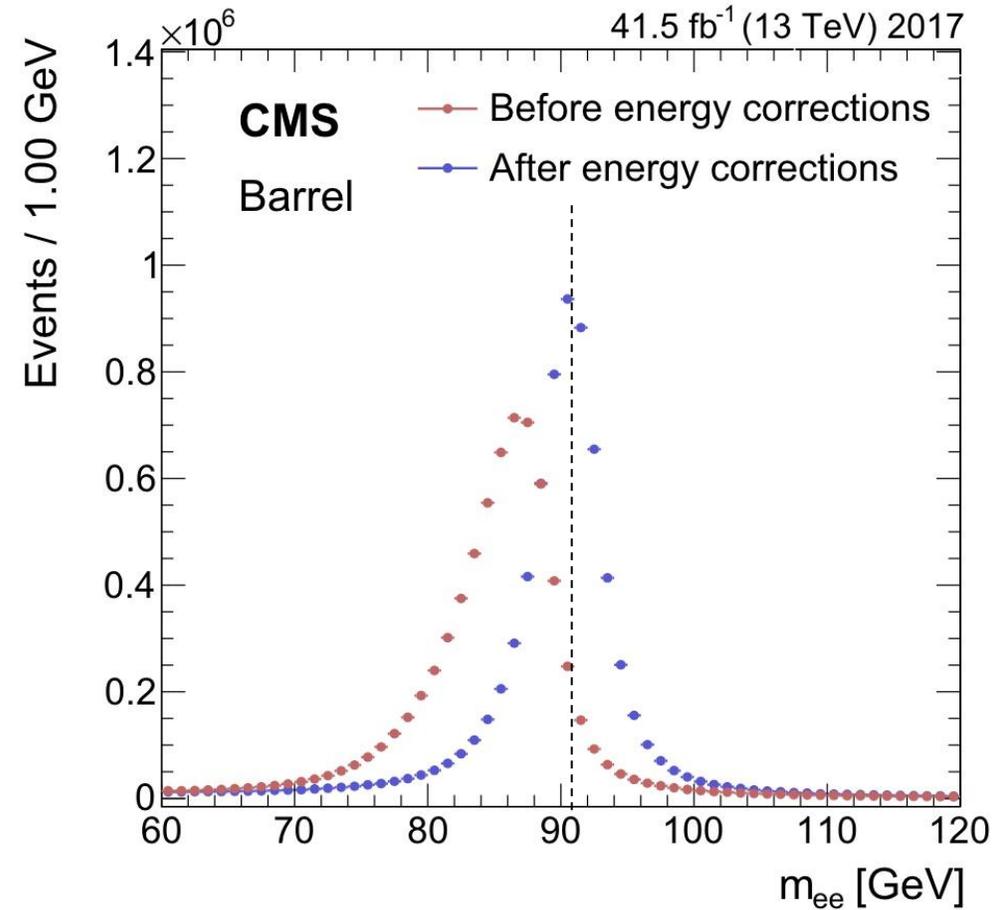
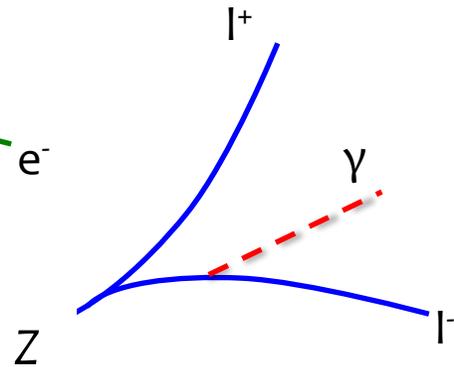
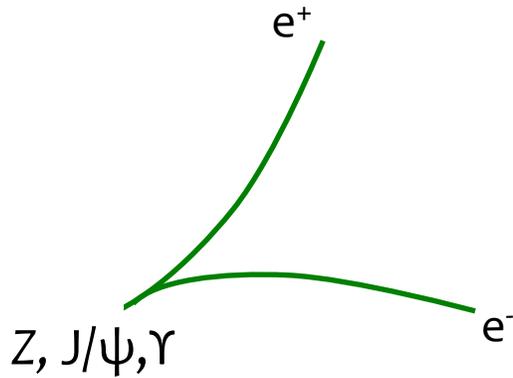
RESULTS

Standard Candles

Z, J/ψ and γ decays in data provide standard candles to calibrate energy/momentum measurements

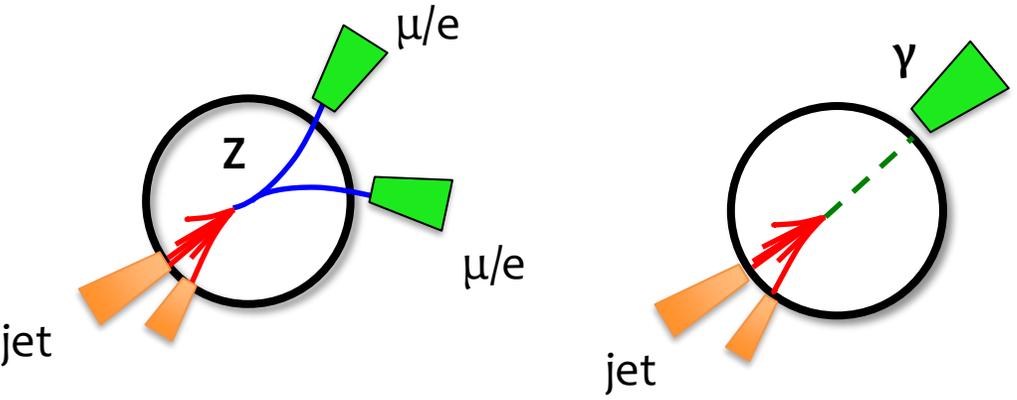


Large source of clean events with a well described mass peak

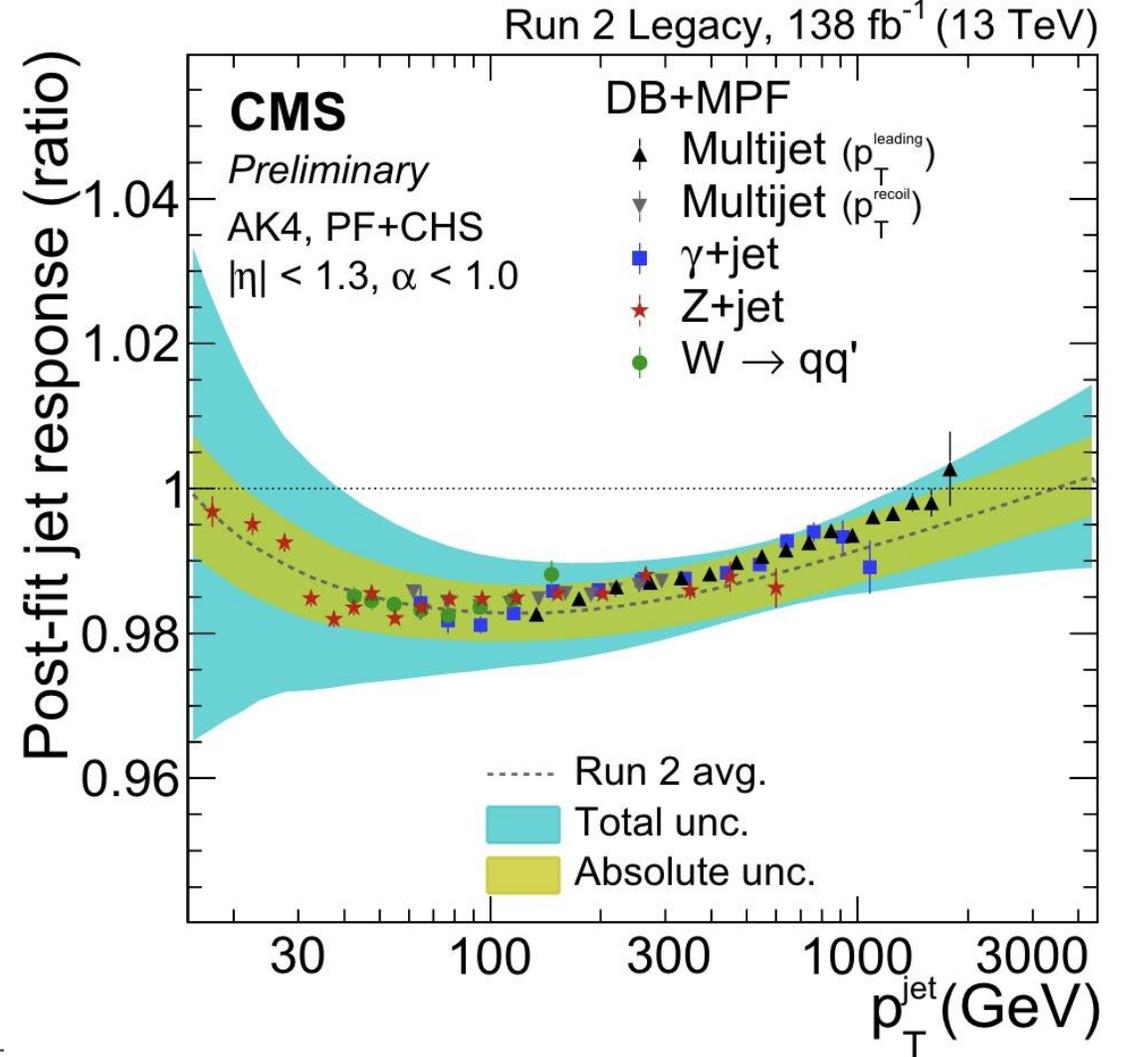


Relative corrections

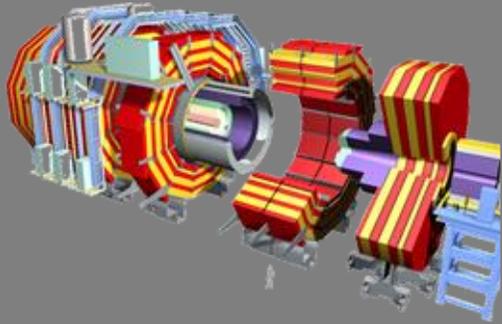
More complicated objects (eg jets) require several stages of correction → Use previously calibrated objects to calibrate jet momentum!



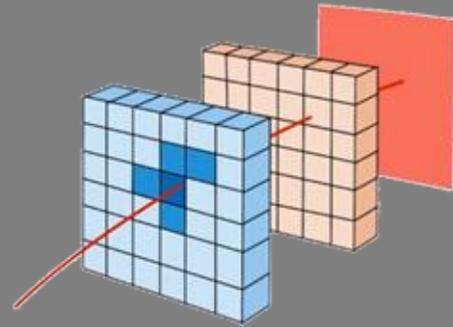
$$R_{jet, p_T} = \frac{p_{T, jet}}{p_{T, Z/\gamma}}$$



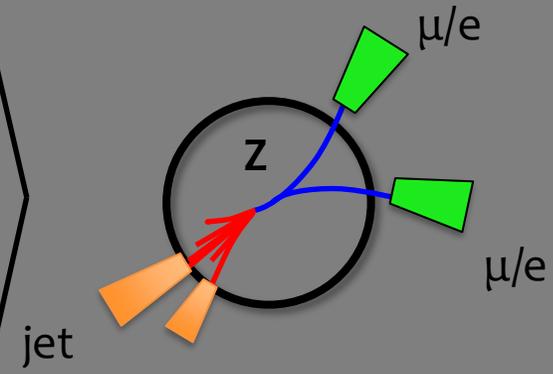
Data



Reconstruction
&
Particle ID

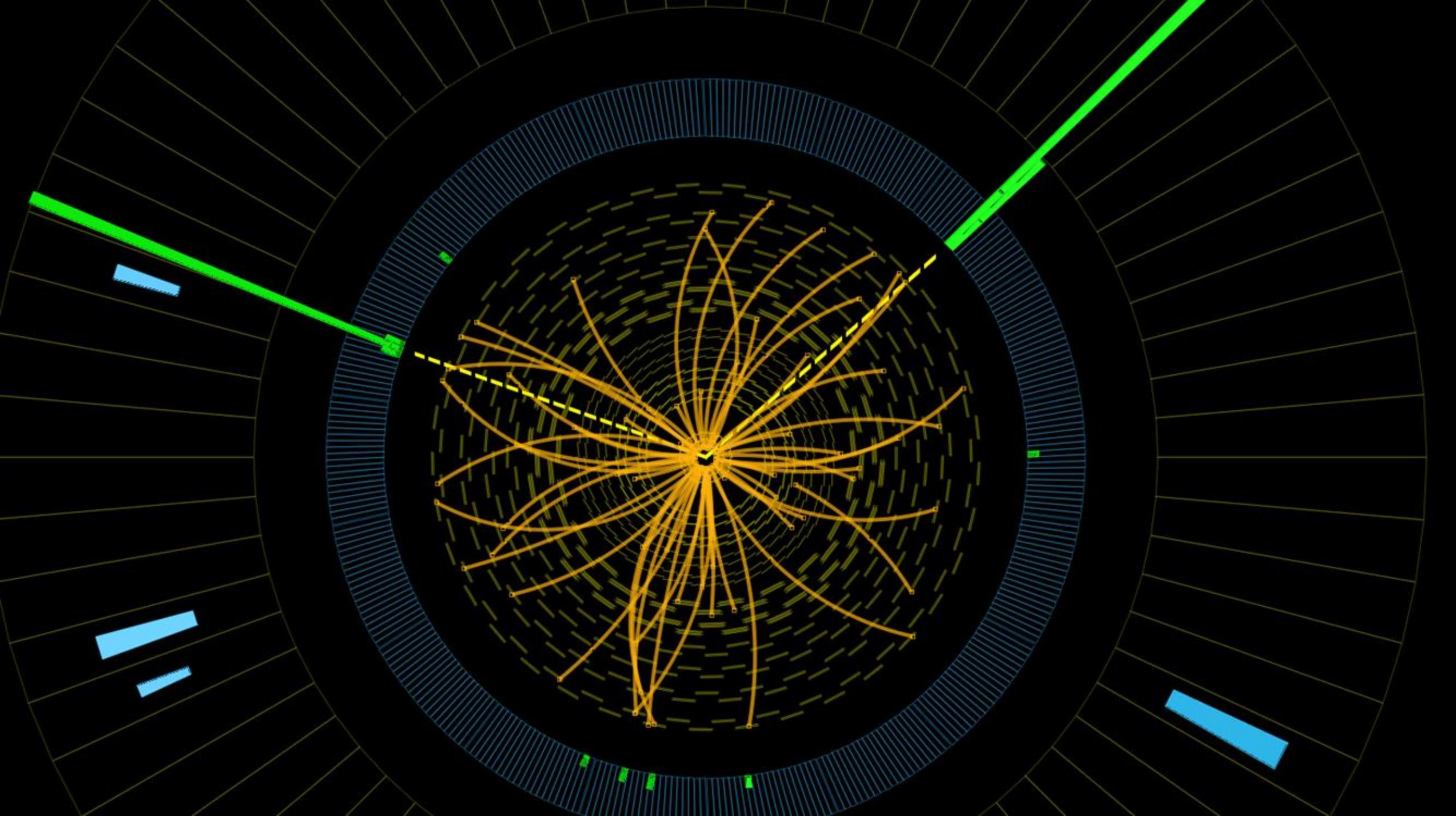


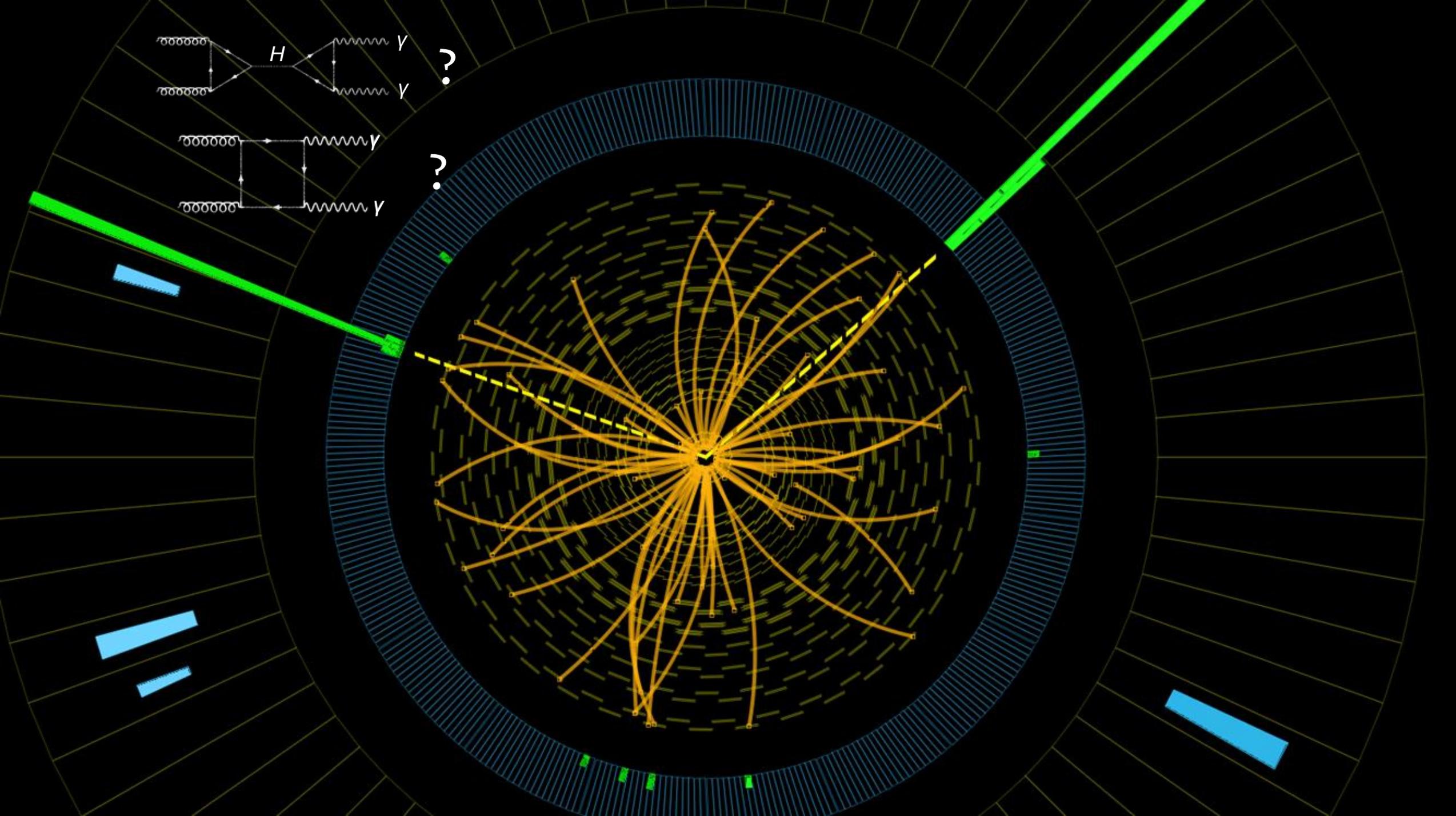
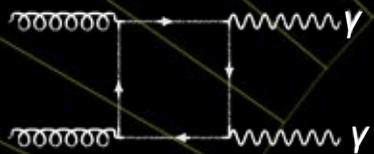
Calibrations

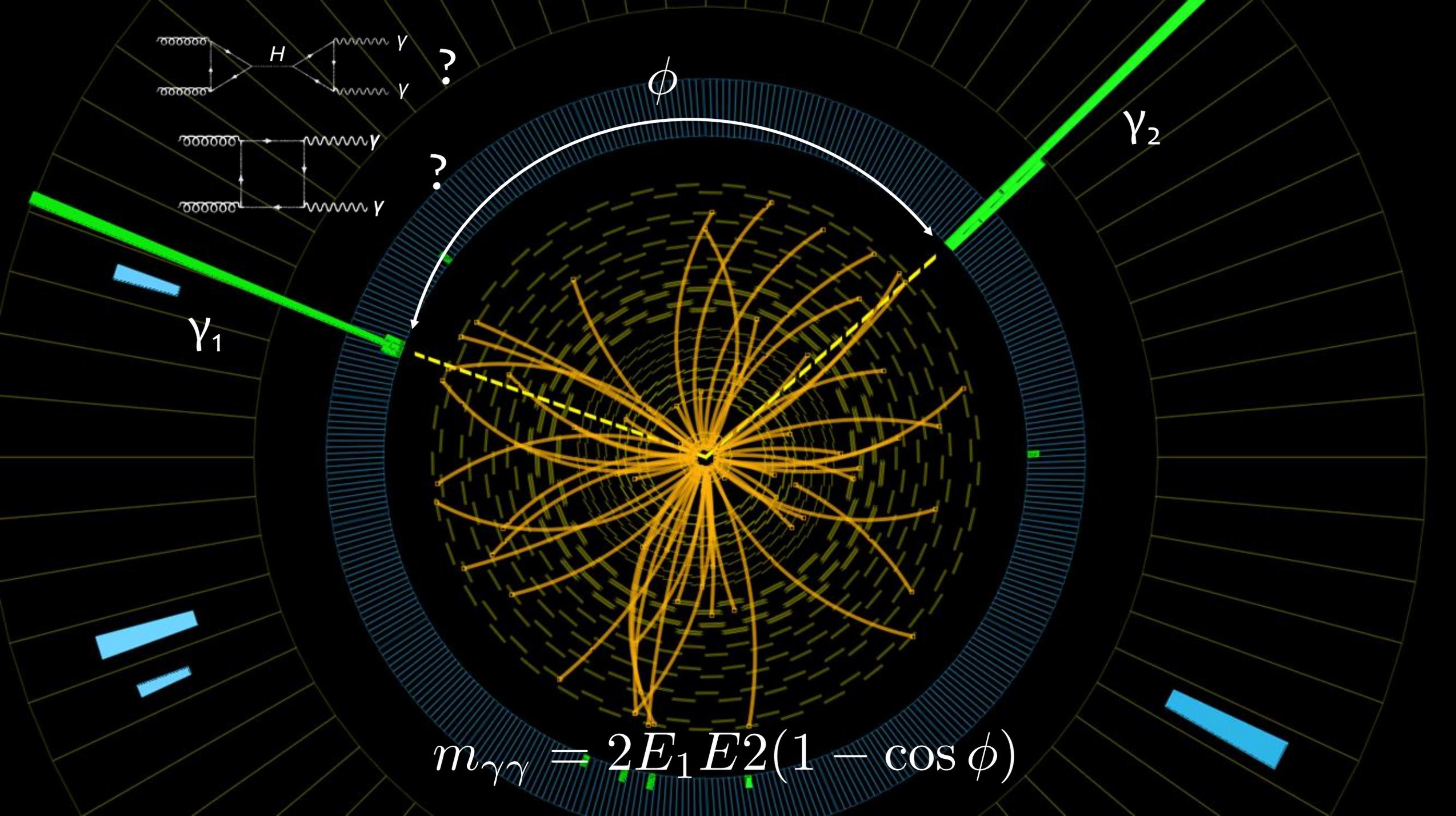


Event Selections
&
Distributions

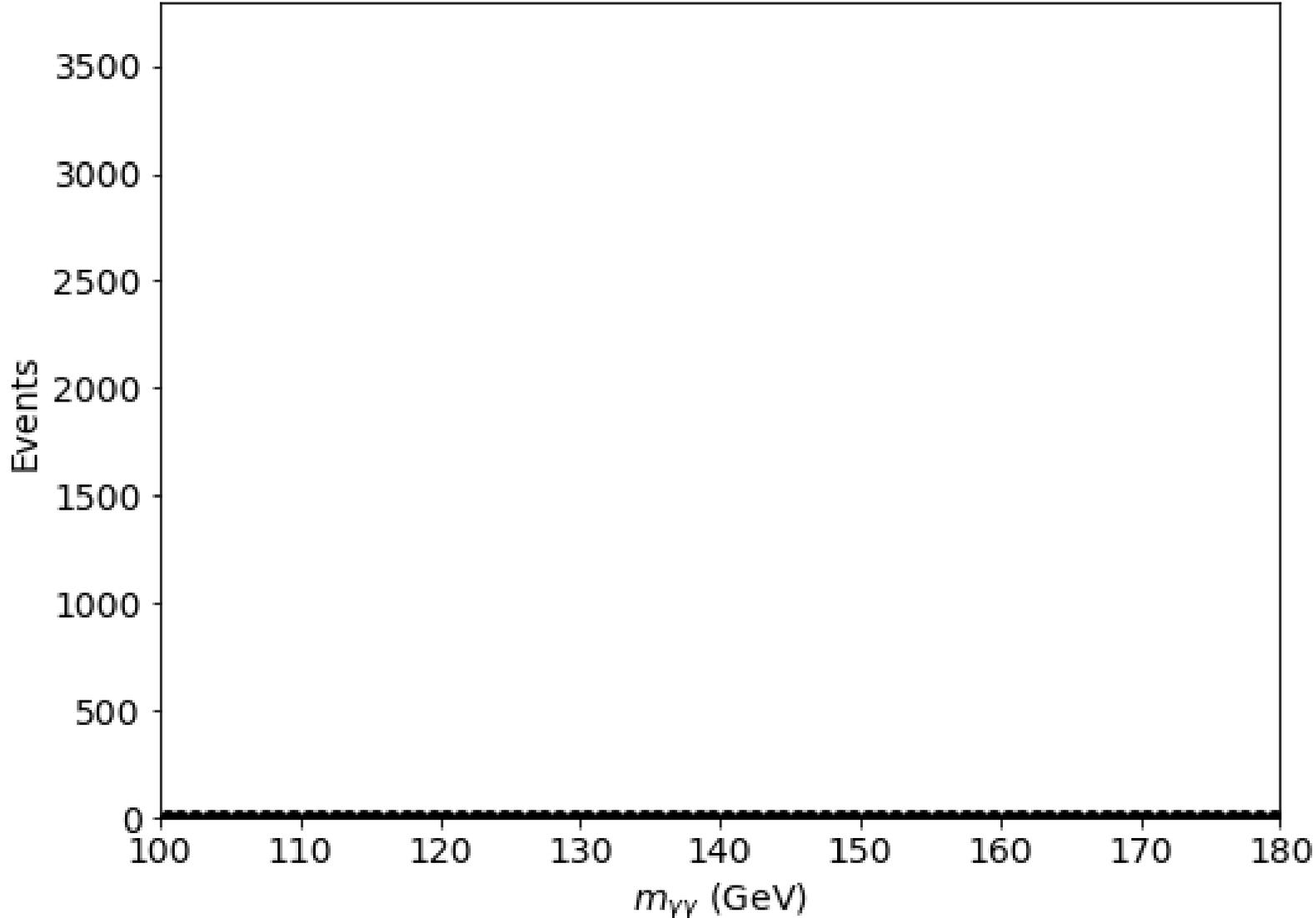
RESULTS







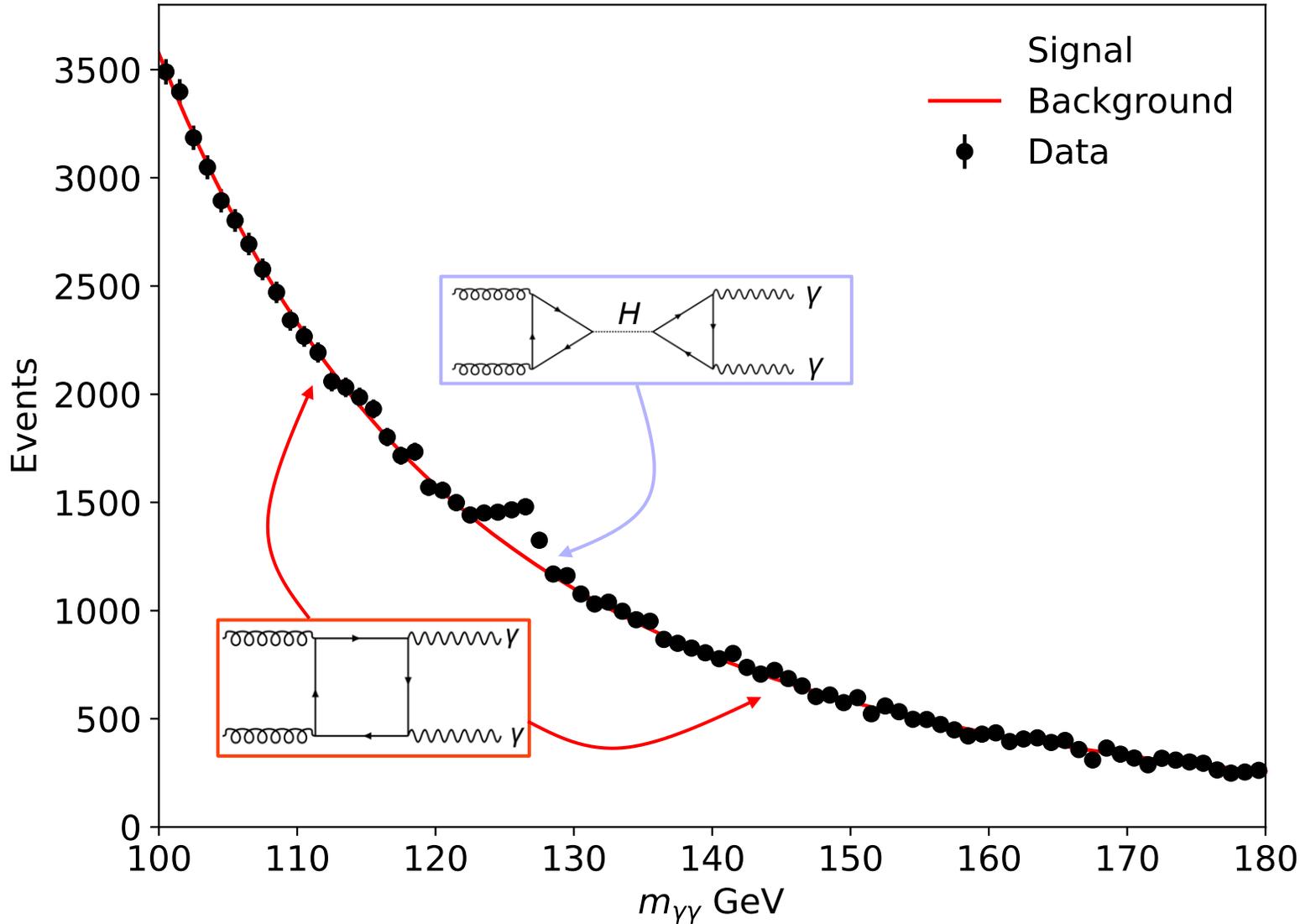
Collecting events



Each event that we select this way builds a picture of the underlying physics → if we're lucky, we might find something new

To extract the Physics, we use **distributions of observables** across many events

Collecting events



Each event that we select this way builds a picture of the underlying physics \rightarrow if we're lucky, we might find something new

To extract the Physics, we use **distributions of observables** across many events

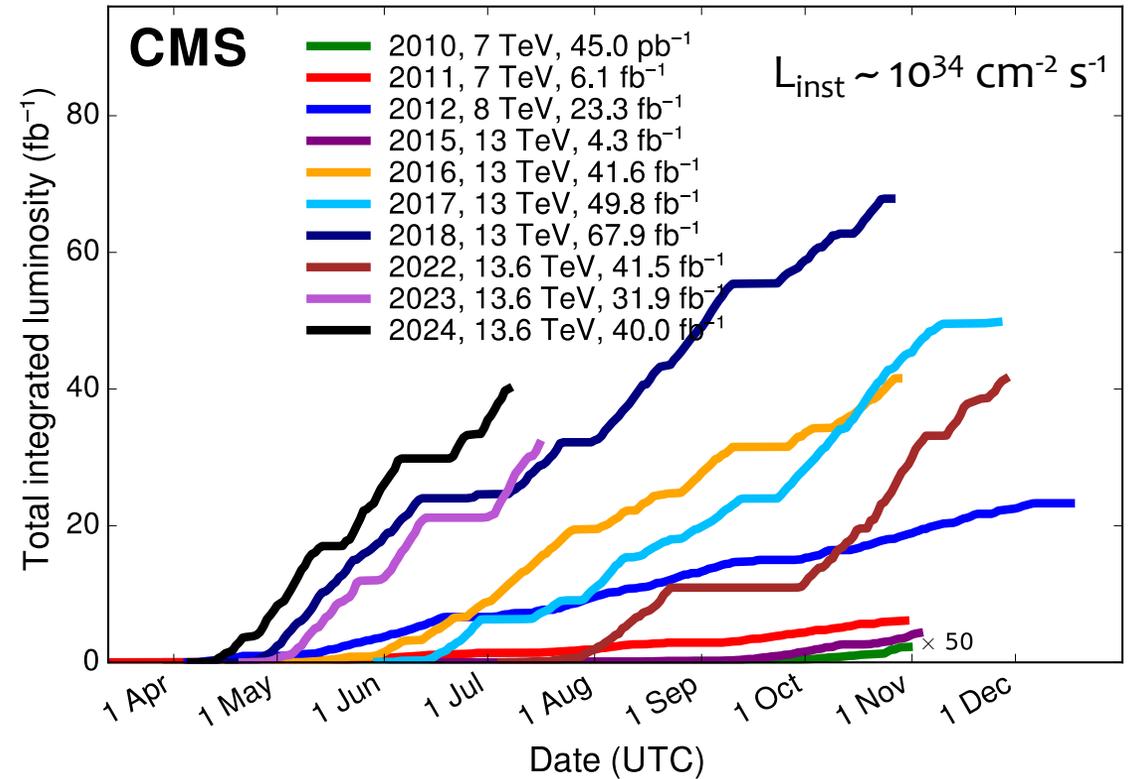
Collecting events

Collisions (i.e. bunch crossings) happen at **40 MHz**

$$\text{Integrated luminosity (units of 1/area)} = \int c \cdot \frac{N_1 N_2}{4\pi\delta_{x,y}} \cdot f n_b dt$$

$$N = L\sigma$$

Number of events



Collecting events

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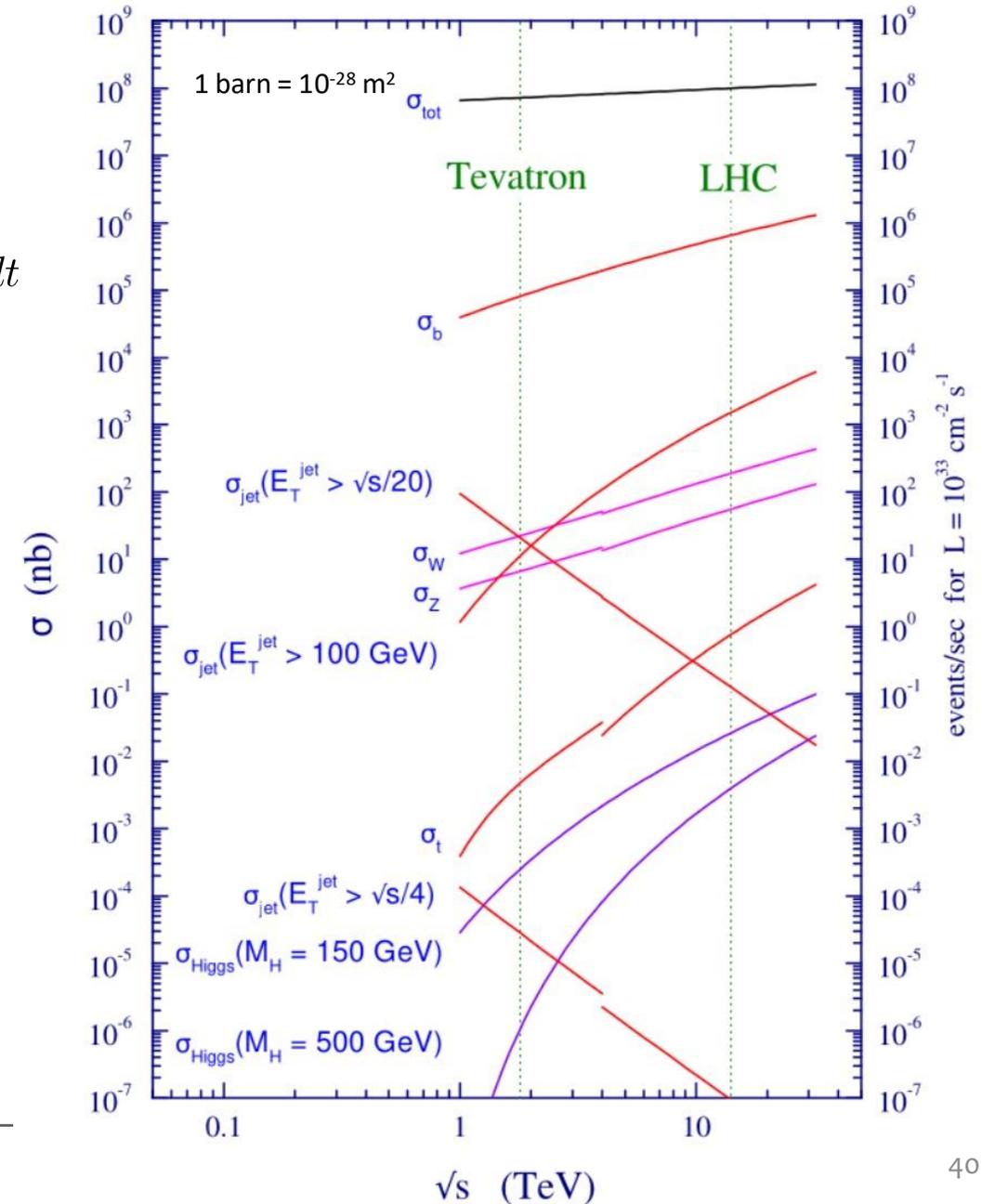
Number of events

Cross-section (units of area)

Example: For every **1,000,000,000** inelastic proton-proton collisions, **only expect one of them to produce a Higgs boson!**

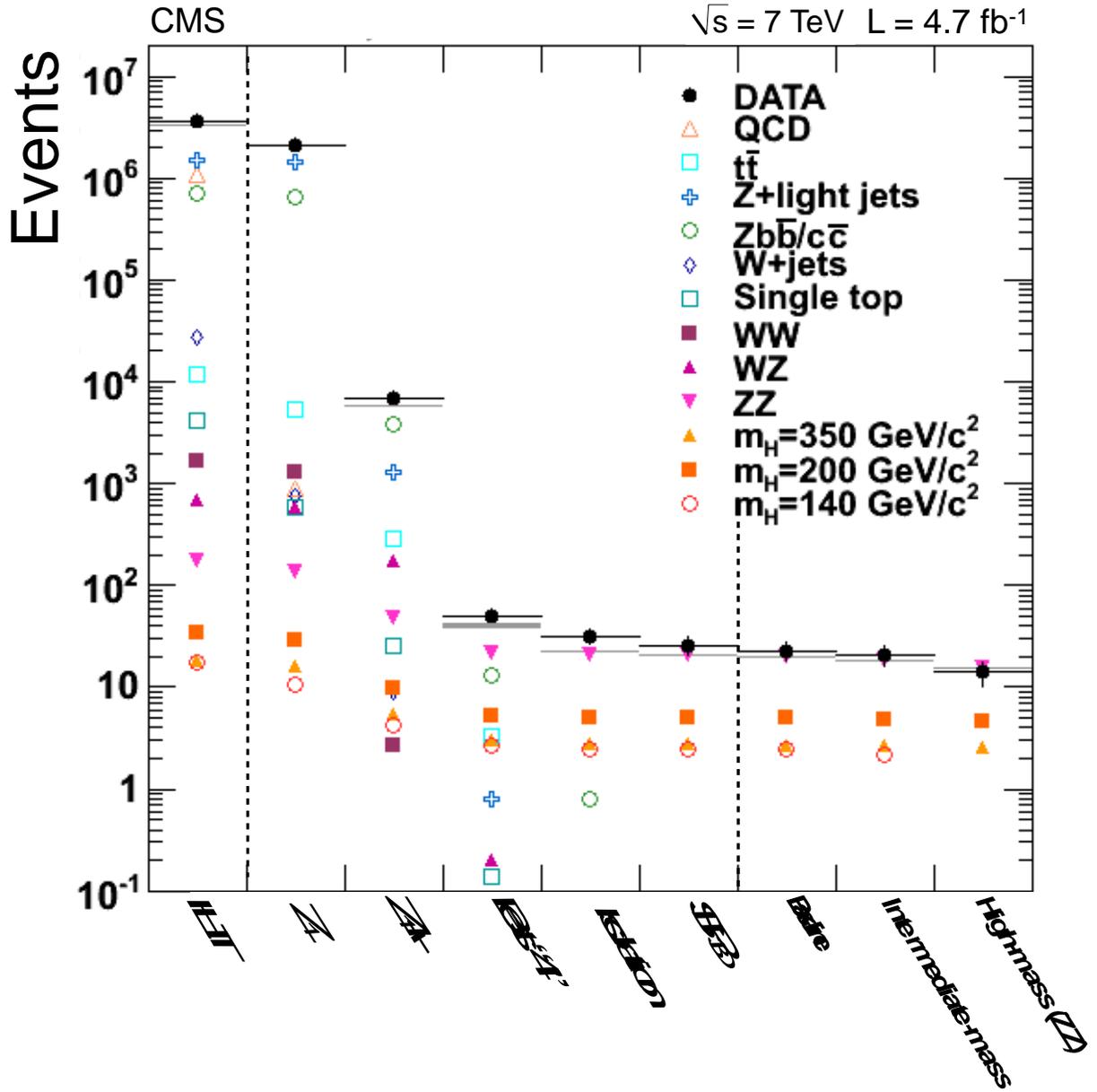
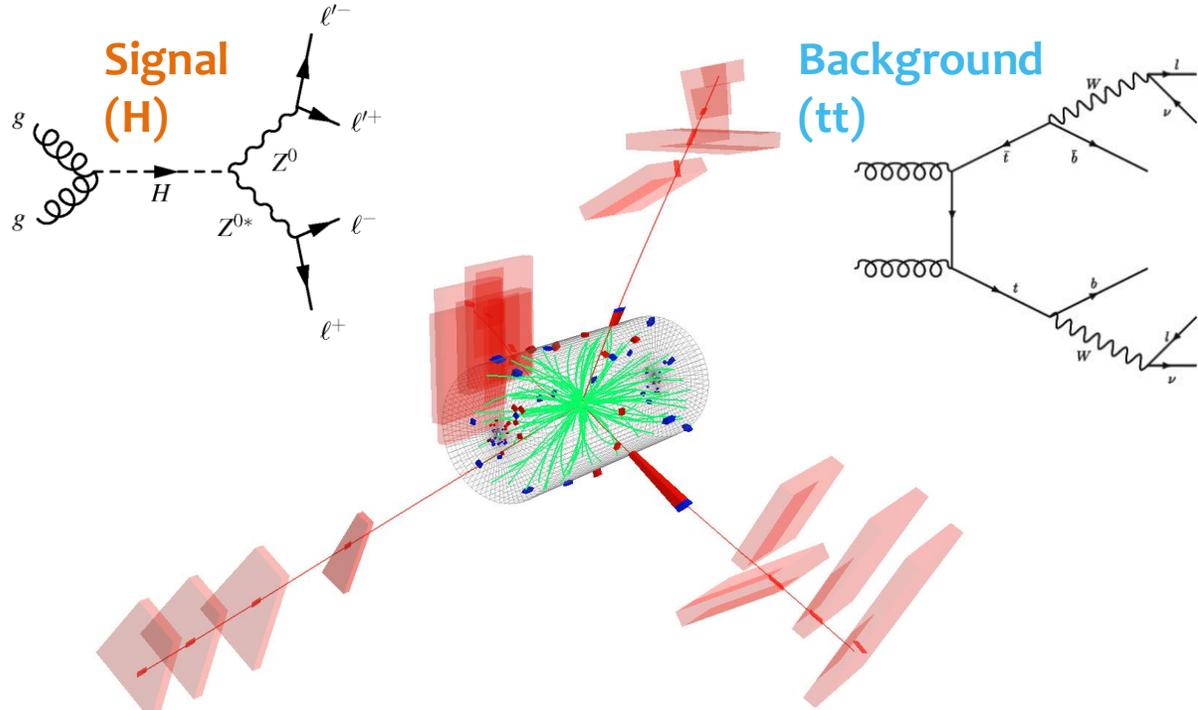
We typically have to select events based on these observables to dig out the signal from the background (noise!)

proton - (anti)proton cross sections



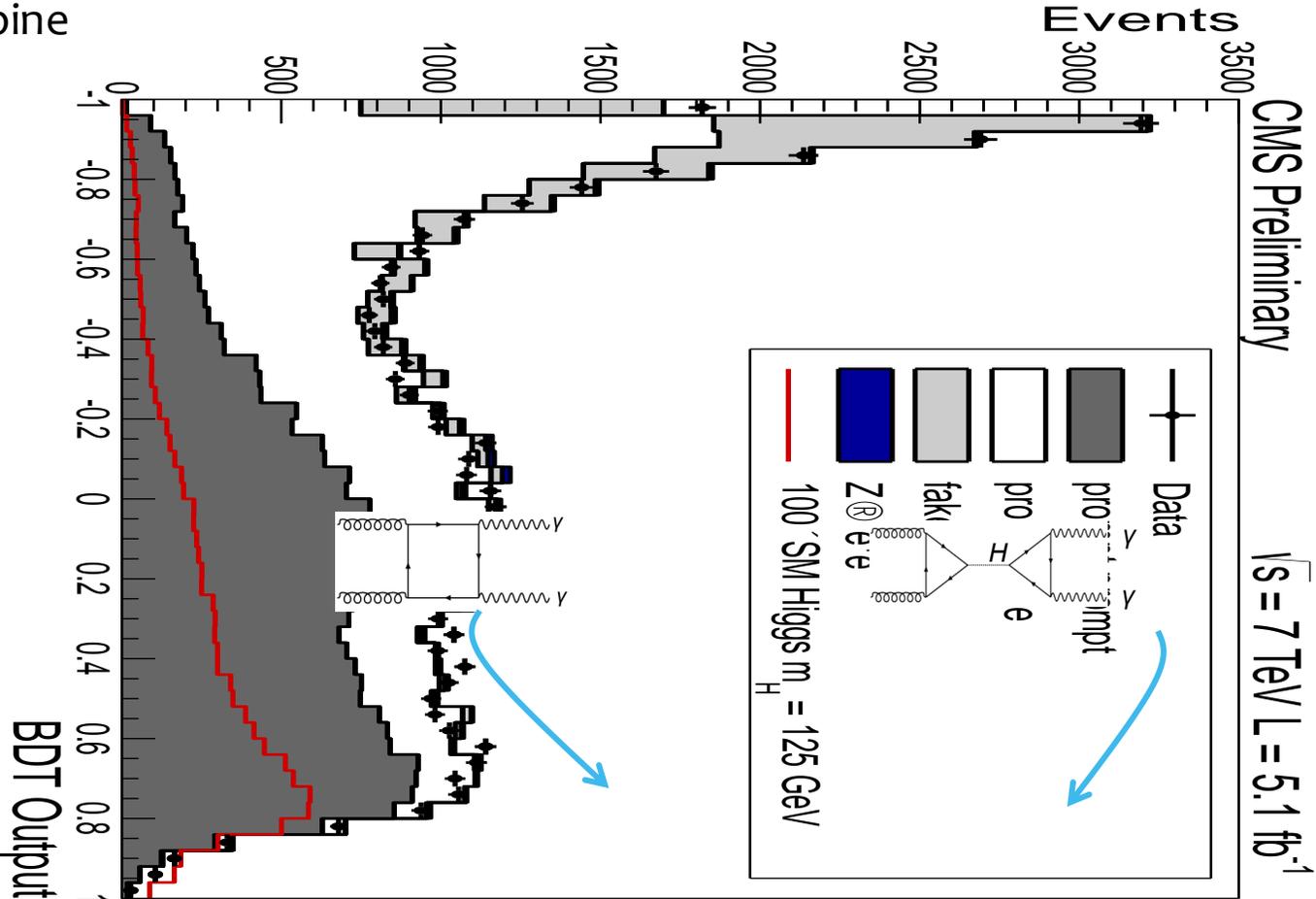
Event Selection

By knowing ahead of time the kind of events we are interested in, we impose selections on the events to reduce the background as much as possible while maintaining the signal

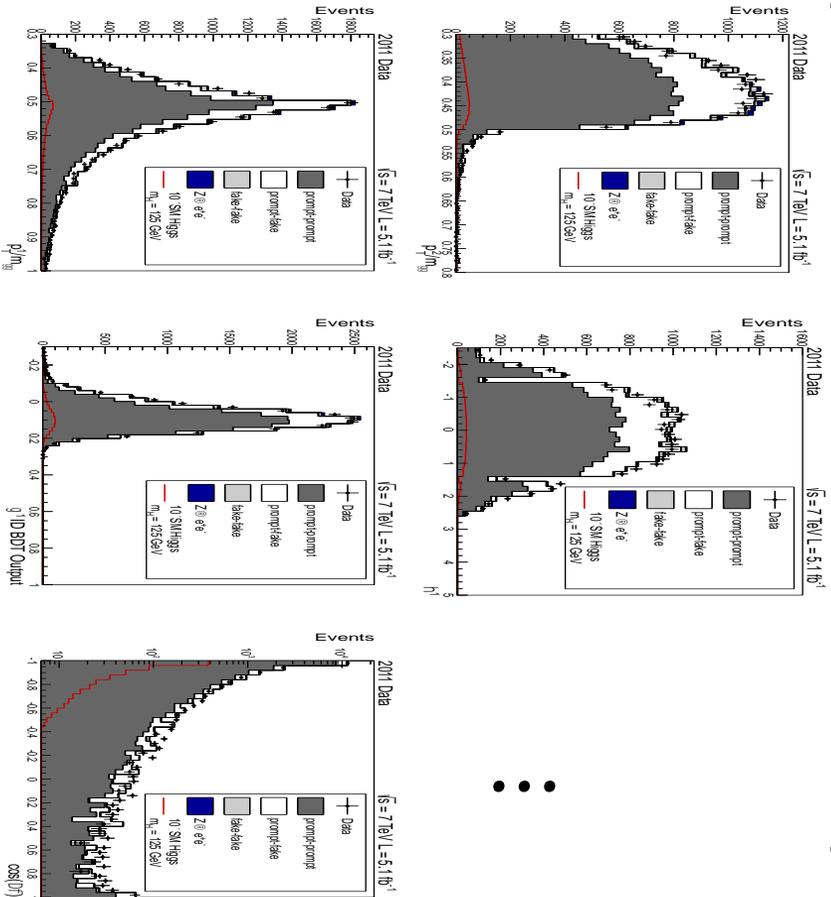


Event Selection

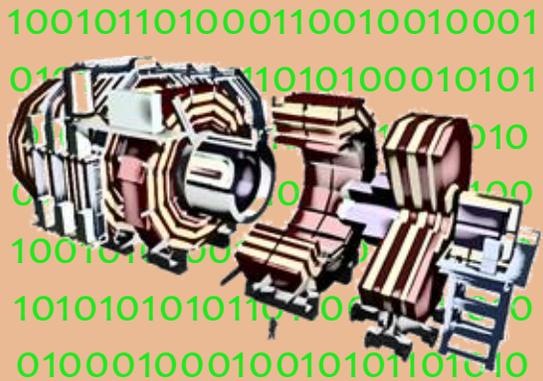
We often use **Machine Learning techniques** to combine as much information as possible for this task



How can we choose these selections/train our Machine Learning models?

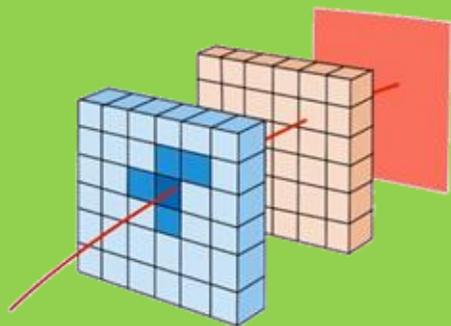


Data

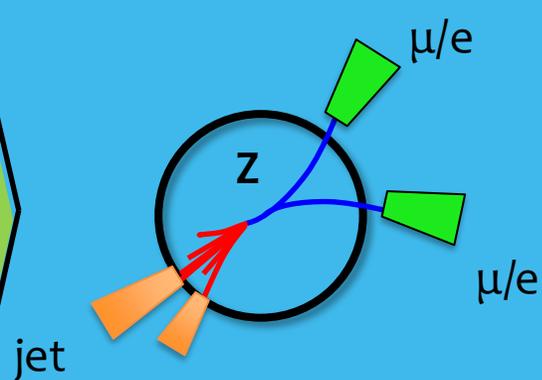


Simulation

Reconstruction & Particle ID



Calibrations



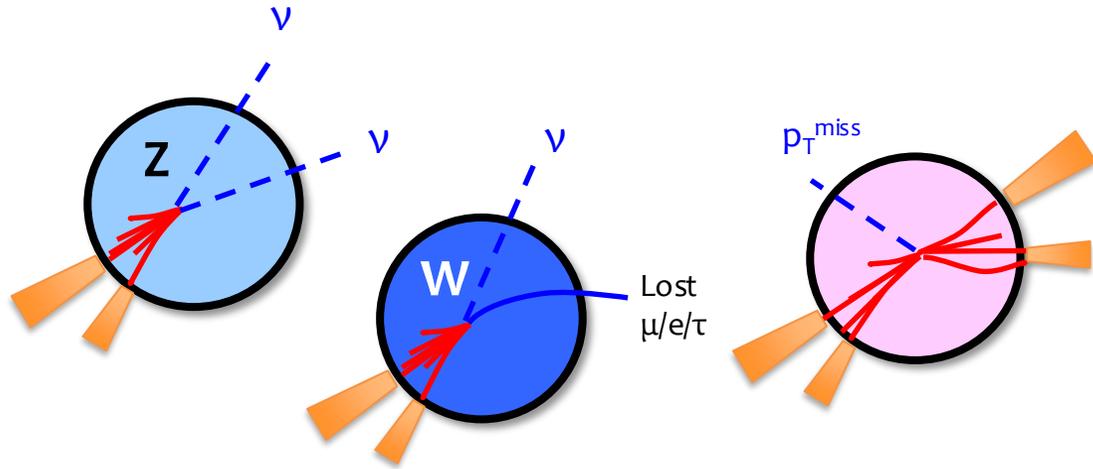
Event Selections & Distributions



RESULTS

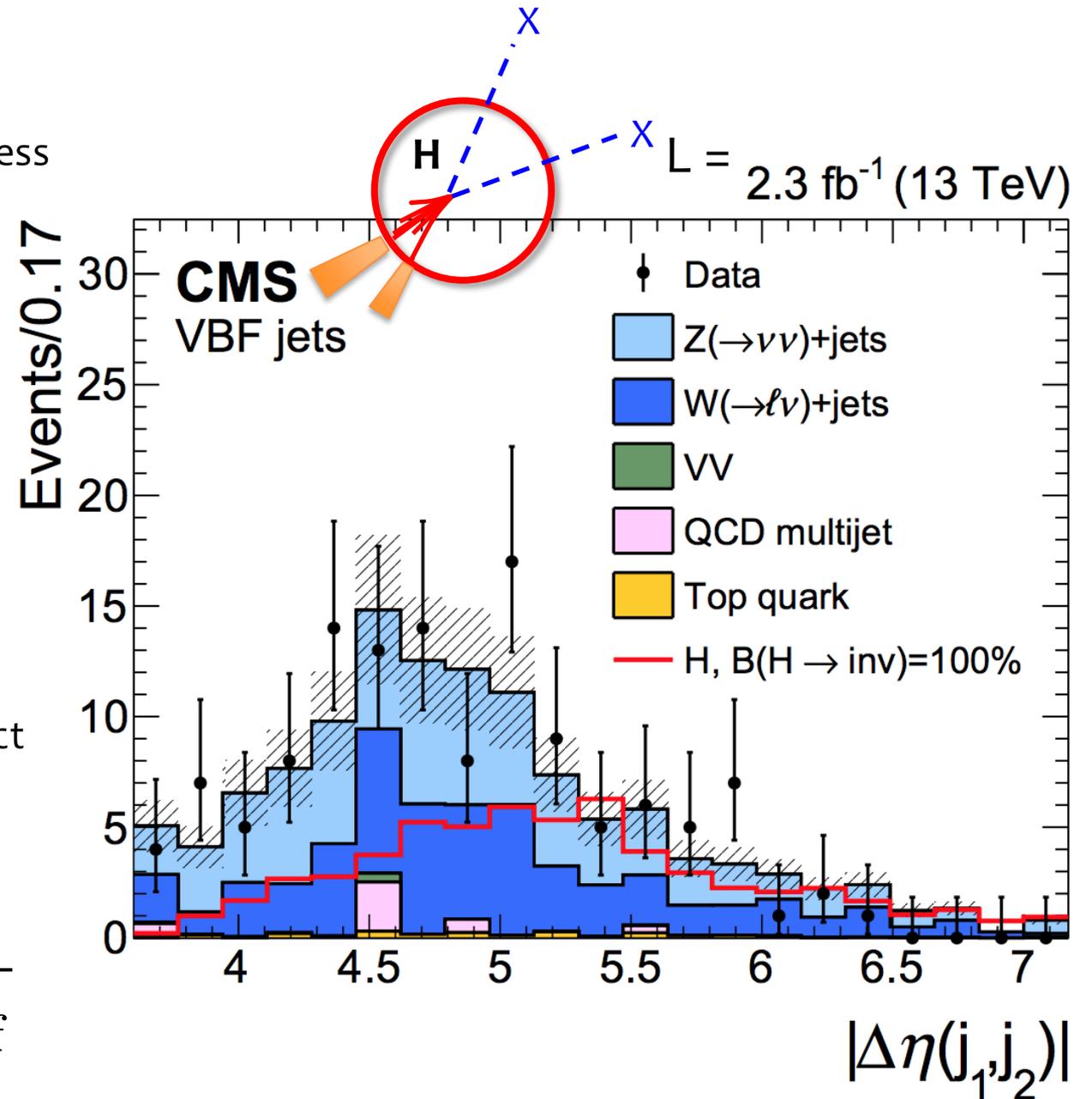
Simulation

Generate large number of simulated events for each process contributing to our analysis (**signals** and backgrounds)

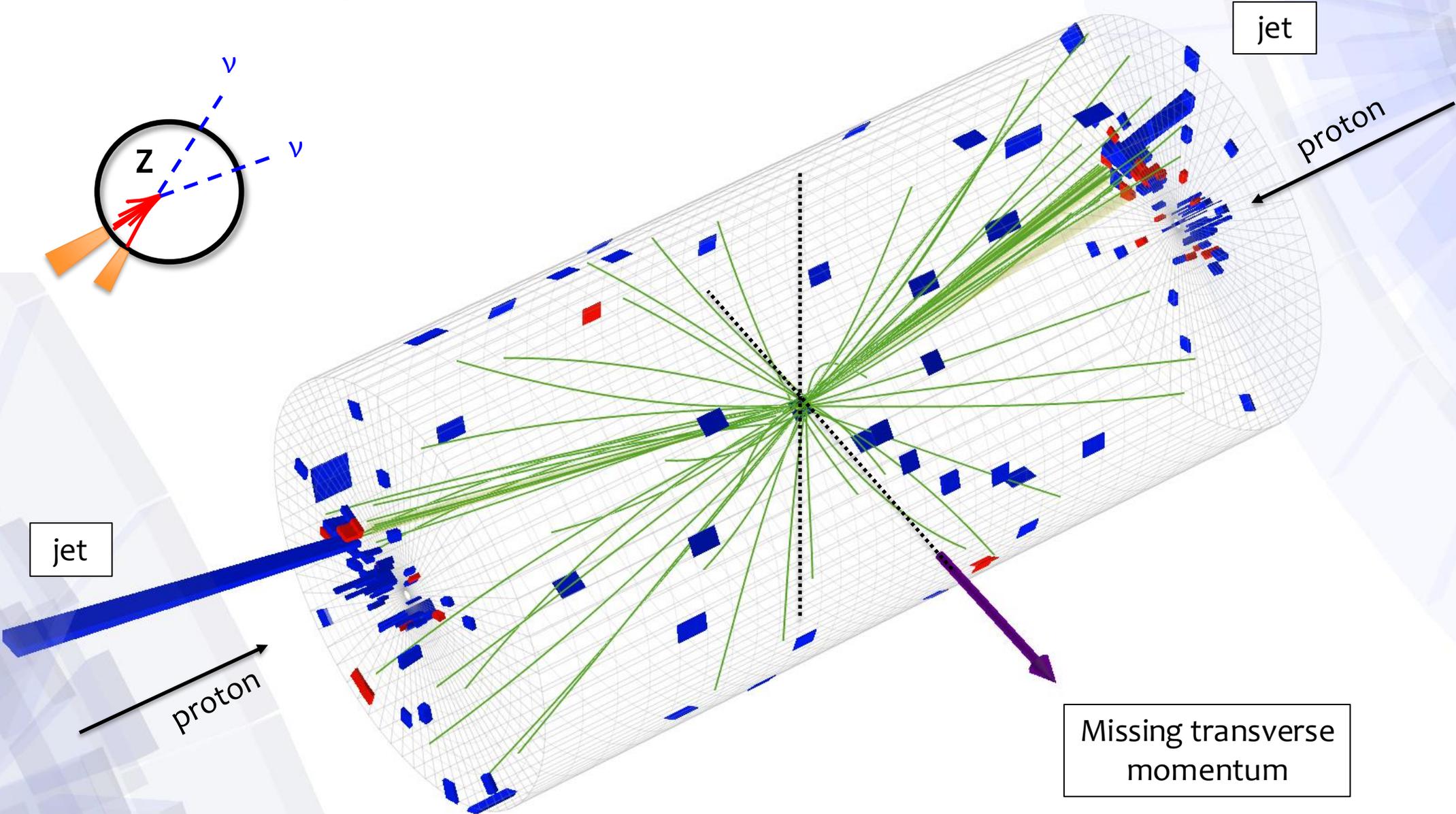


Simulated events must be **weighted** to get the correct predicted yield for a given dataset

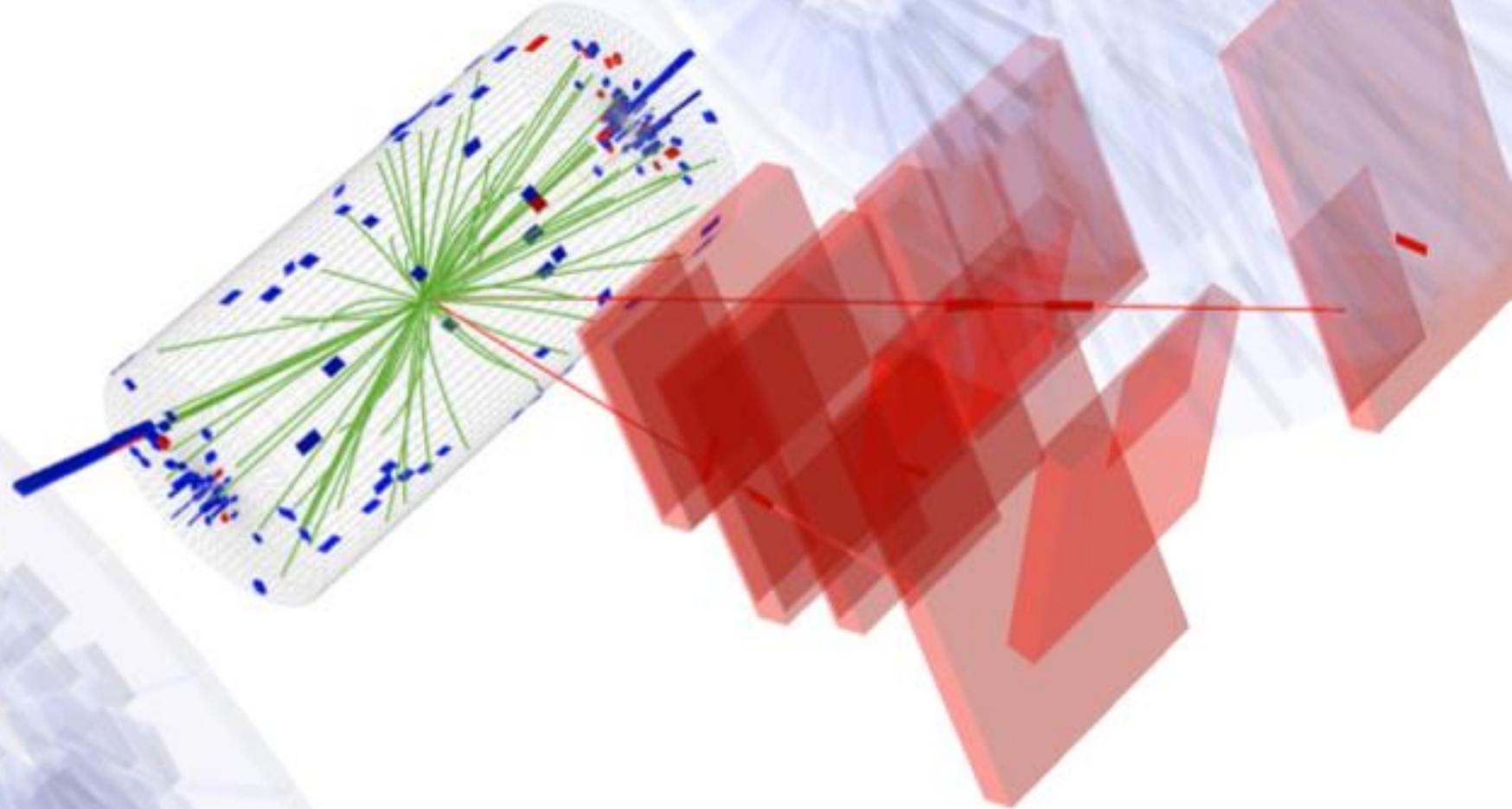
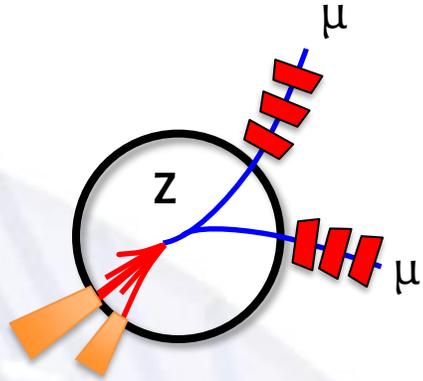
$$L_{\text{eff}} = \frac{N_{\text{gen}}}{\sigma} \rightarrow \text{weight} = \frac{L}{L_{\text{eff}}}$$



Data-driven background



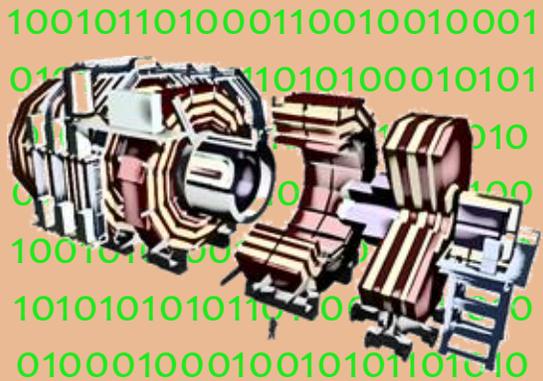
Data-driven background



Estimate the normalization of the $Z \rightarrow \text{neutrinos}$ background using data!

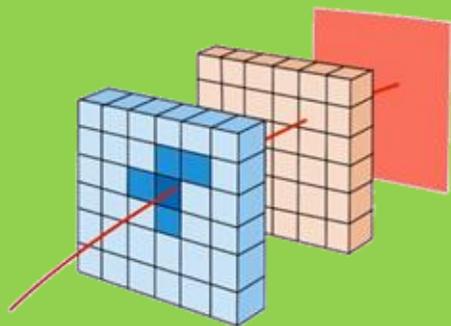
$$N_{Z(\rightarrow \nu\nu)} \approx N_{Z(\rightarrow \mu\mu)} \frac{B(Z \rightarrow \nu\nu)}{B(Z \rightarrow \mu\mu)} A(\mu)\epsilon(\mu)$$

Data

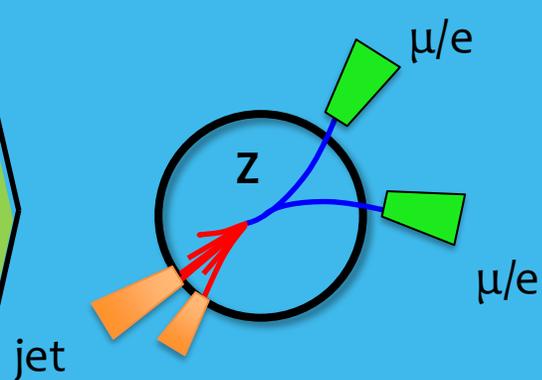


Simulation

Reconstruction & Particle ID



Calibrations



Event Selections & Distributions

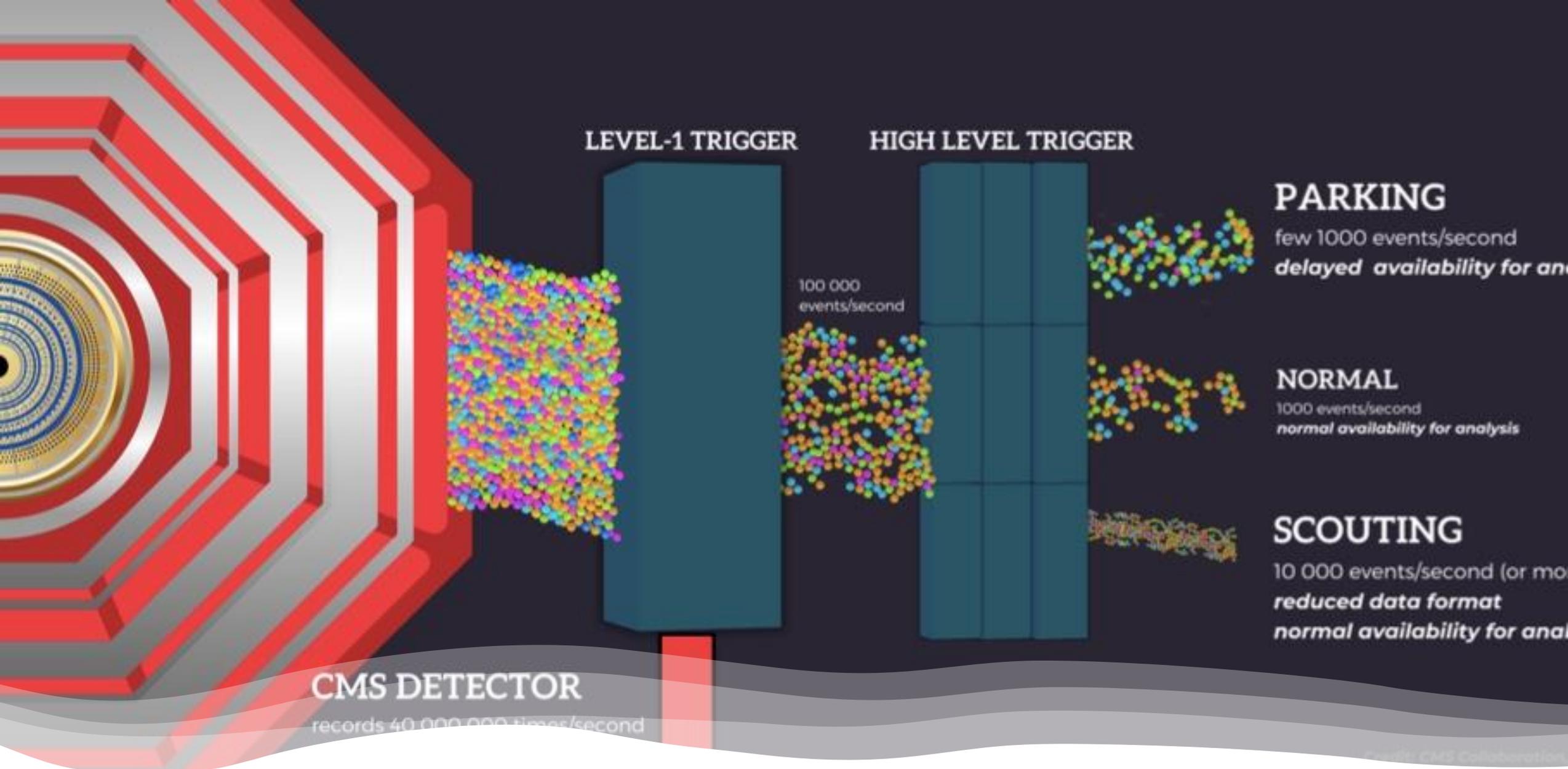


RESULTS



Huge computing power
required to acquire and
analyse LHC data





Online Selection of events “Trigger” determines which events to keep in around 4 micro-seconds!

A large, multi-level atrium with a central circular structure and many people on balconies. The central structure is a complex, multi-layered circular arrangement of blue and red components, resembling a particle detector. The balconies are filled with people of various ages and ethnicities, looking towards the center. The architecture is modern, with white walls and glass railings. The lighting is bright, highlighting the intricate details of the central structure and the diverse group of people.

Huge collaborations of
people required for Data
Analysis at the LHC

Now it's your turn!

This afternoon, we are going to have a go at doing a data analysis with some real CMS proton-proton collision data!

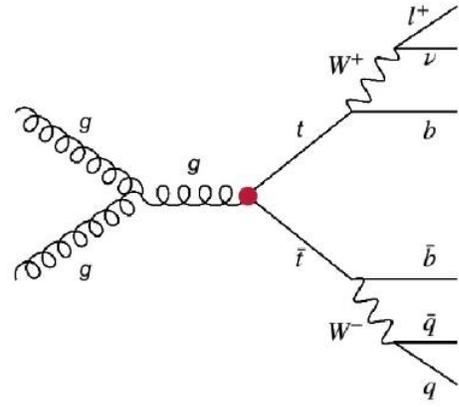
All of the instructions for getting setup and the exercises are available here:

<https://nucleosynthesis.github.io/LHCDataStatisticsICISE2024/>

You will also see links to the lecture slides (password VSOP LHC2024)

If you haven't already done so, please go through the "Getting started" section before this afternoon's session!

This afternoon, we will be working through Exercise 1



LHCDataStatisticsICISE2024

Home

Getting started

Exercises

Getting Started

To complete these exercises, we will be using two container images, with the software installed for you. In the examples here, we will use Docker to run the images. The Docker desktop is available for mac, windows and linux so follow the link and download the right installation for your personal laptop. You should start by downloading the Docker desktop for your laptop (click [here](#) and follow the instructions). You will need to setup an account to do so.

Once you have the Docker desktop installed, make sure it is running and download the two containers that we'll need for the exercises using the terminal commands below. Note that the Docker desktop has its own terminal if you prefer to use that. If you are using a linux machine, be sure that you allow all users to access the Docker daemon otherwise you will need to add `sudo` to the start of your `docker run` commands. Below is what the Docker desktop looks like for Mac, where you can see which containers are running.

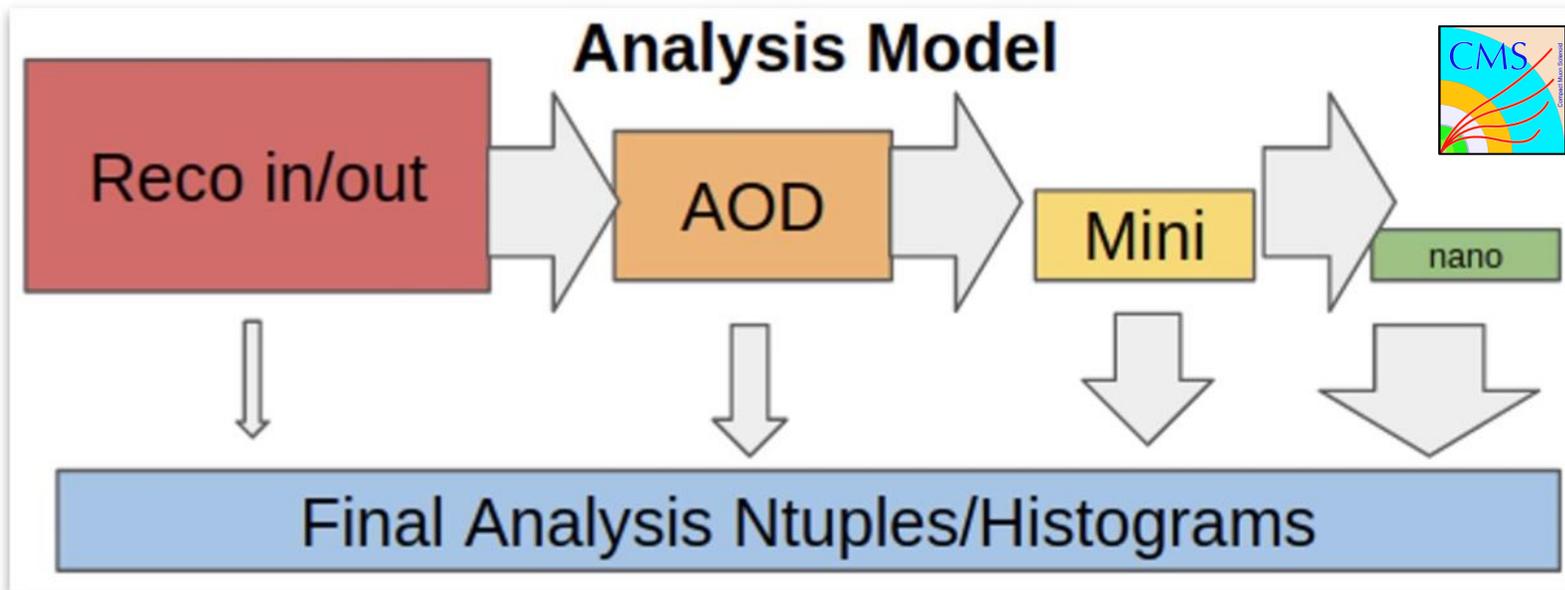
Name	Image	Status	Port(s)	CPU (%)	Last	Actions
romantic_willow	luc0p45002	Running		0%	16 mi	1
villains_fisher	luc0p45001	Exited		0%	17 mi	1
objective_morse	luc0p45001	Exited		0%	19 mi	1
morally_ambiguous	luc0p45001	Exited		0%	23 mi	1
combined_1.1.1	luc0p45001	Running	8888:8888	0.0%	1 day	1
cms_python	luc0p45001	Exited (255)	5901-5901	0%	6 day	1

(Extra Slide) Data Tiers

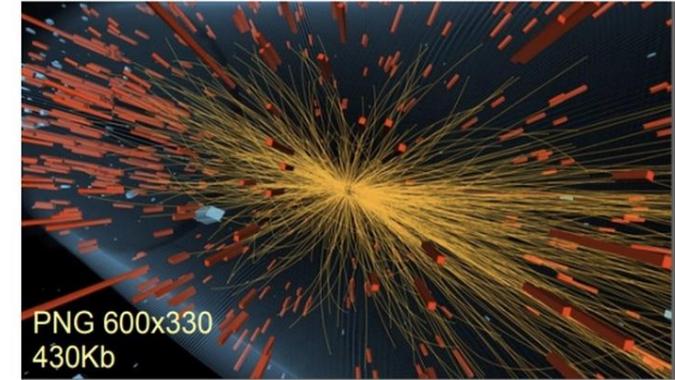
RAW data output from experiments is too large for direct analysis

e.g 2018 data from CMS $O(10)$ PB at RAW data level

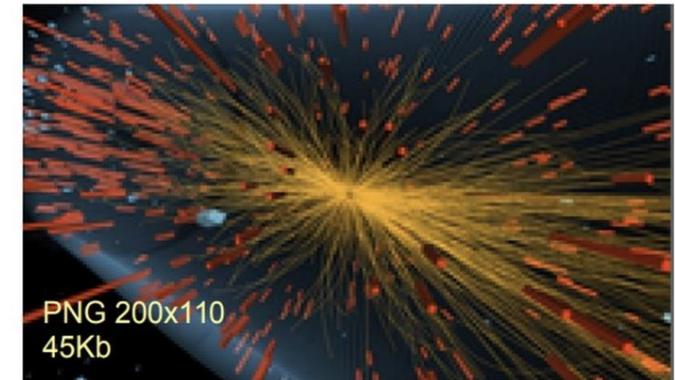
- Reduce content through processing at different data tiers to make analysis manageable
- Less information but content is closer to final analysis objects (hits → particles)



AOD 450kb/ev



MiniAOD 50kb/ev



NanoAOD 1-2kb/ev



(Extra Slide) A Real CMS analysis selection flow

