

BOOST Camp (Experimental)

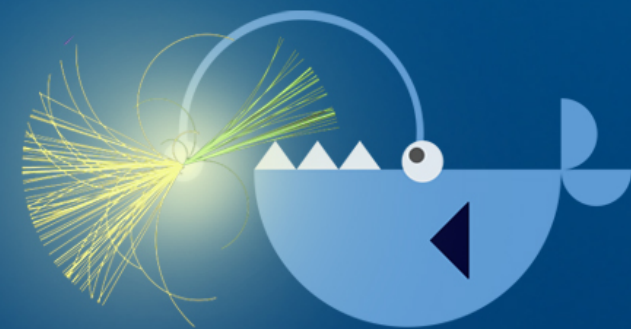
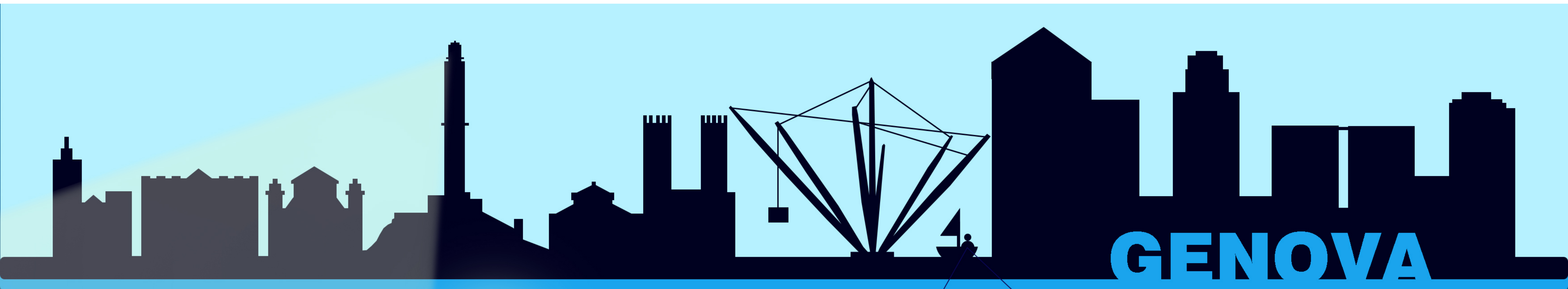
*16th International Workshop on Boosted Object Phenomenology,
Reconstruction, Measurements, and Searches at Colliders (BOOST 2024)*

Genova, Italy, 29 July - 2 August 2024

Matt LeBlanc (Brown University)



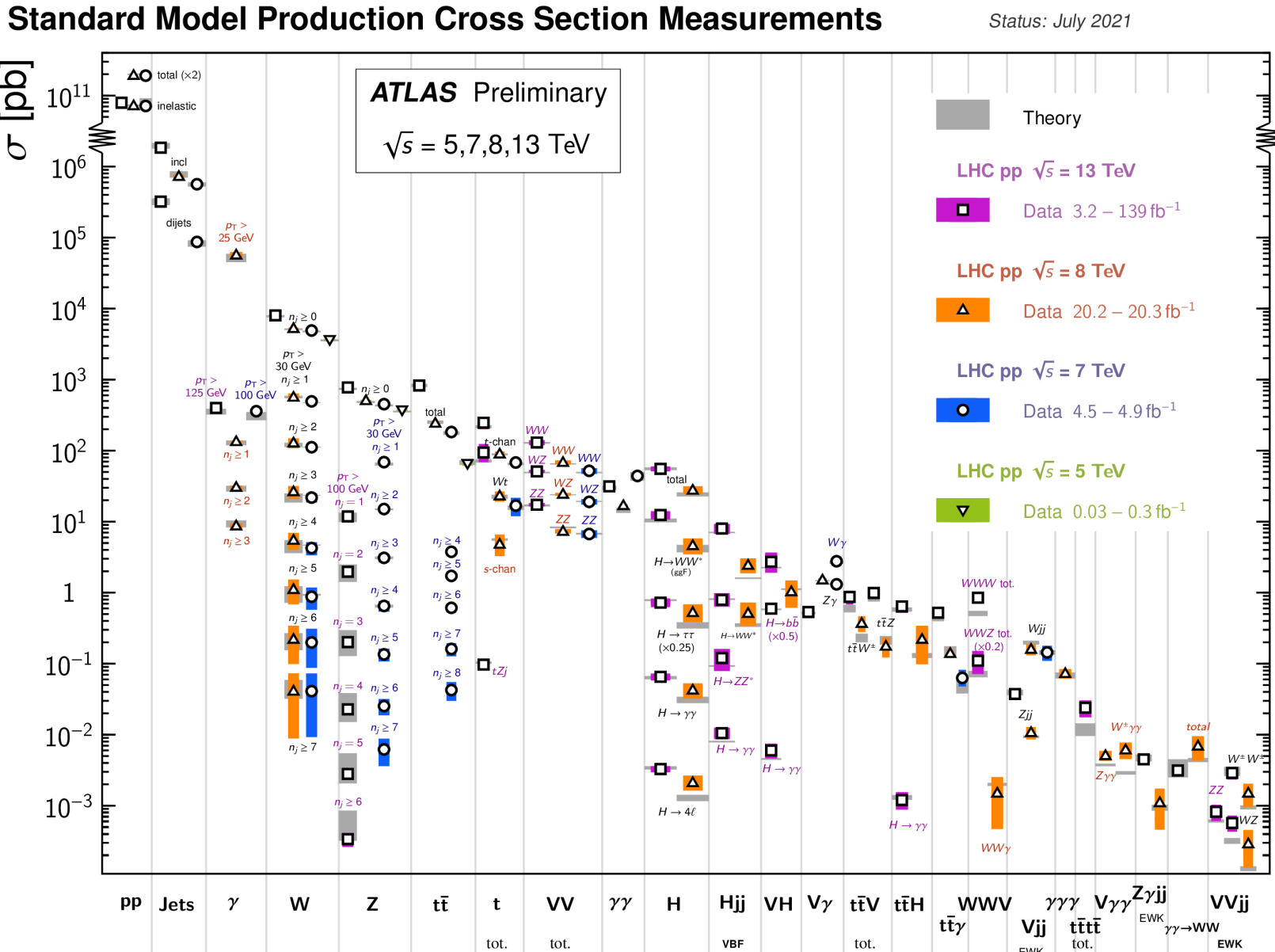
BROWN



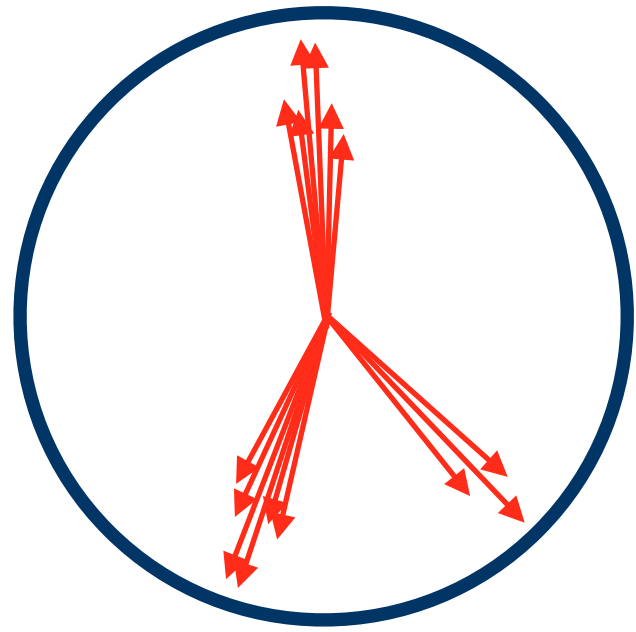
16th International Workshop on
Boosted Objects Phenomenology



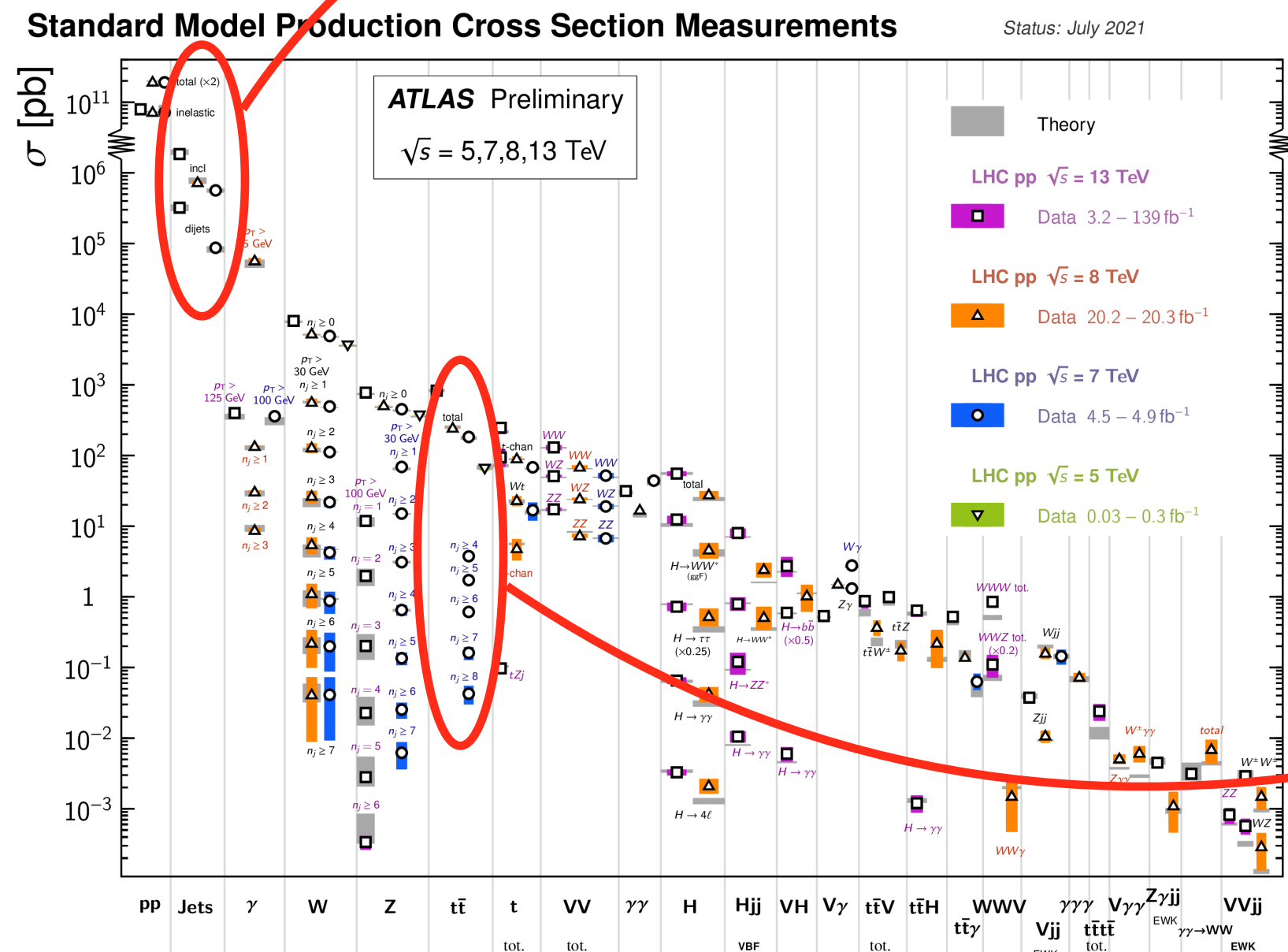
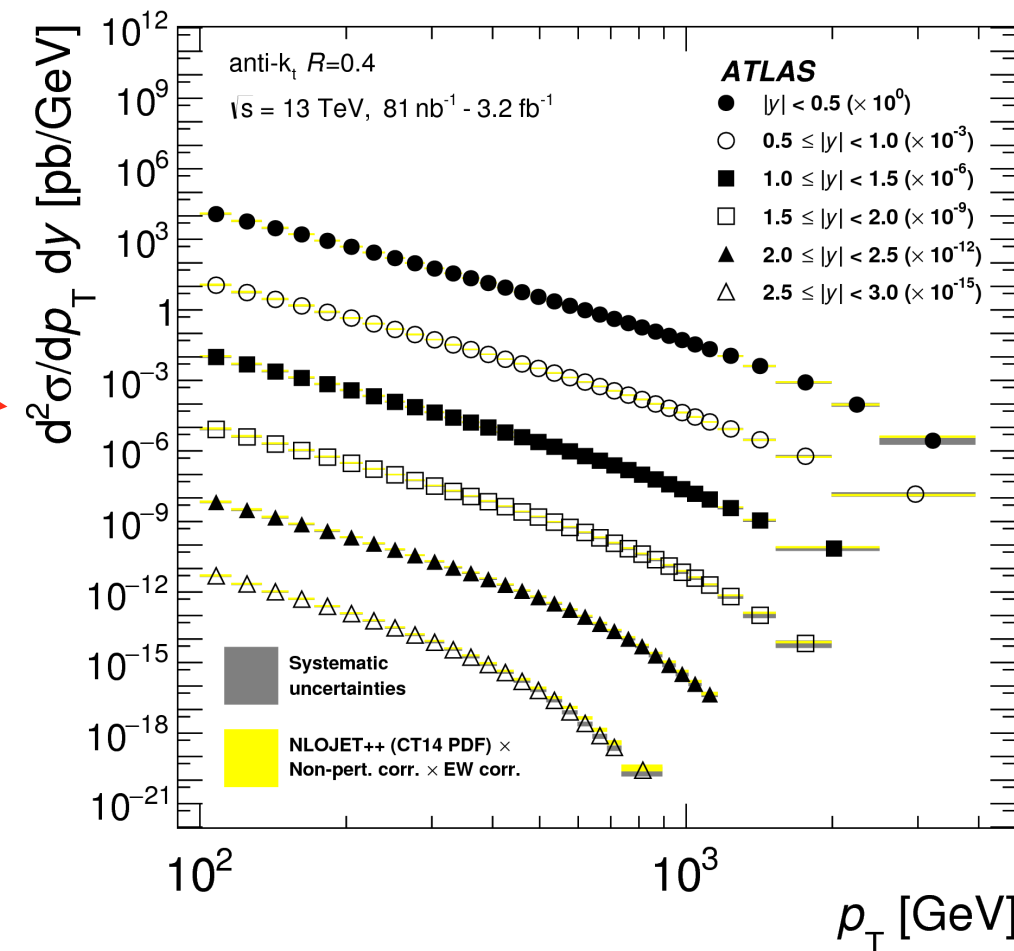
At the LHC, we can study **many processes**



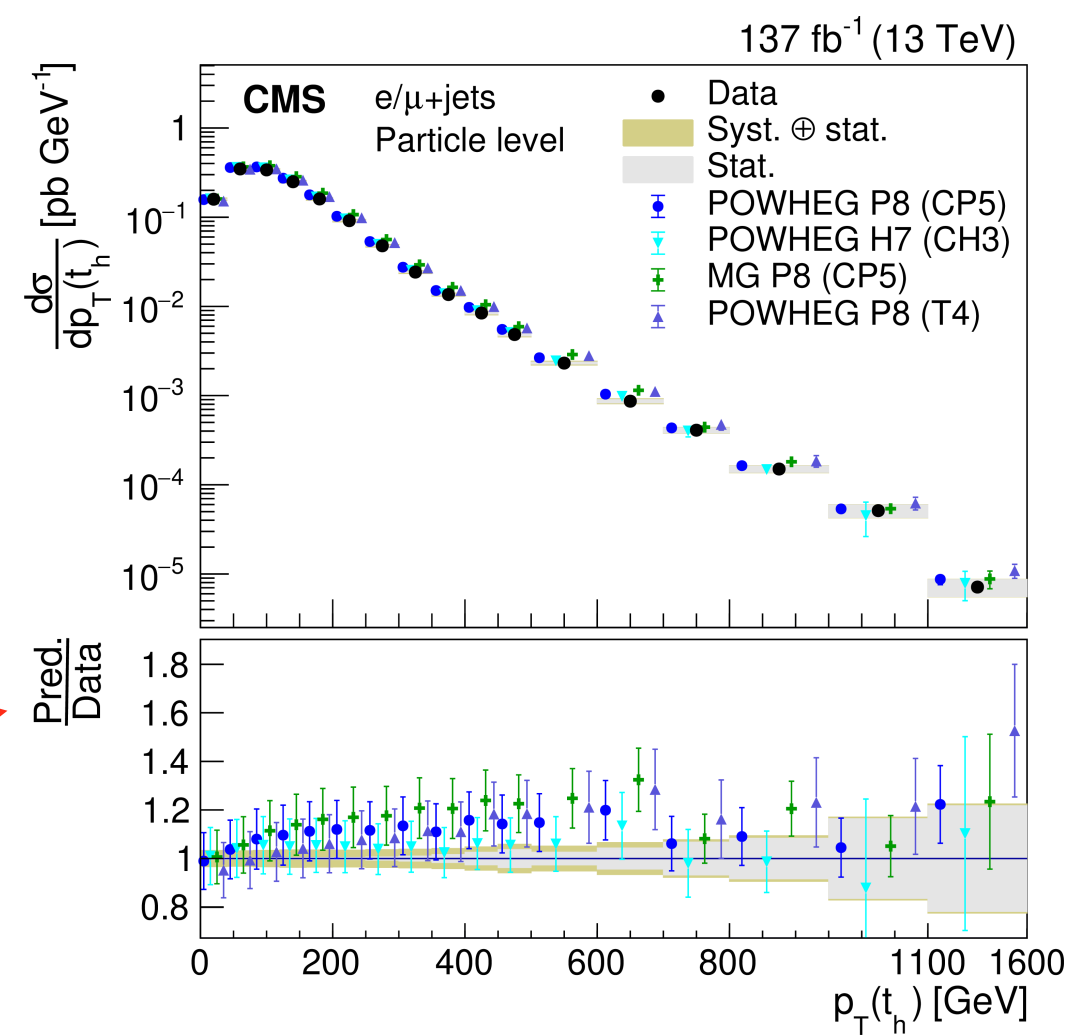
At the LHC, we can study **many processes** at **many scales** ...



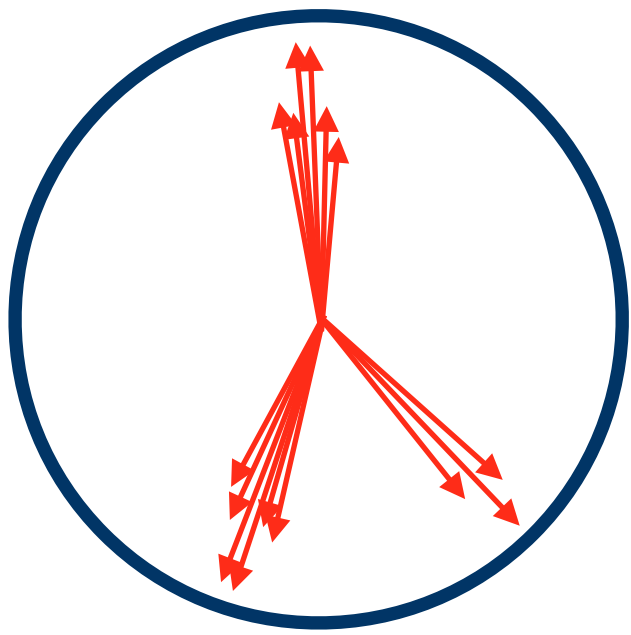
e.g. multijets



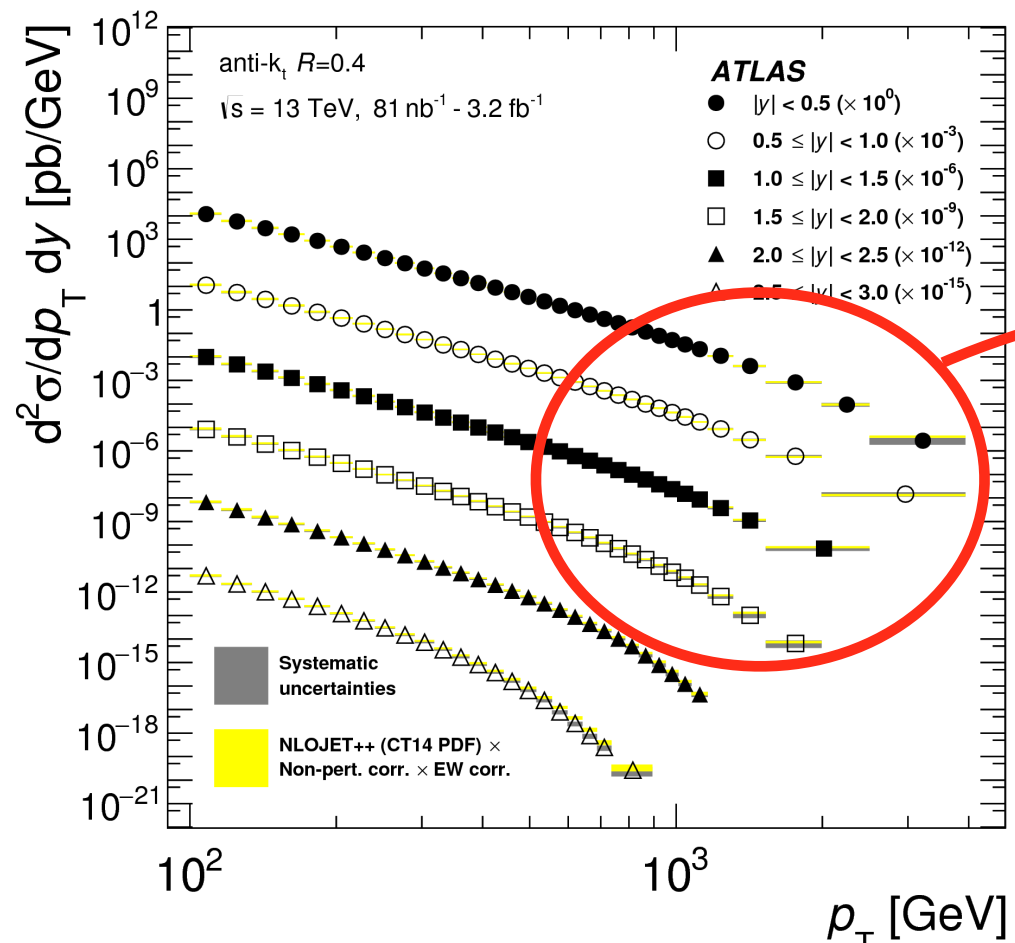
e.g. top pairs



At the LHC, we can study **many processes** at **many scales** ...

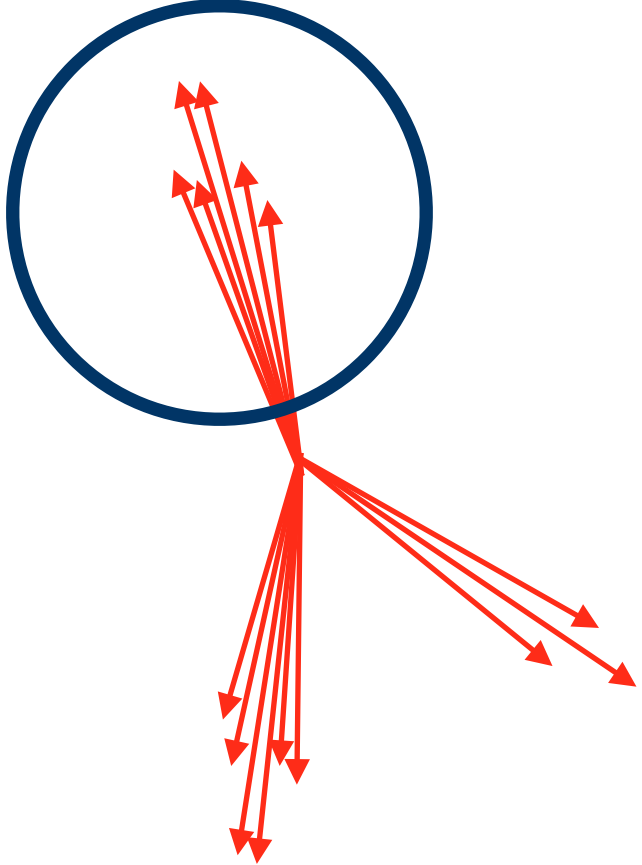


e.g. mutijets

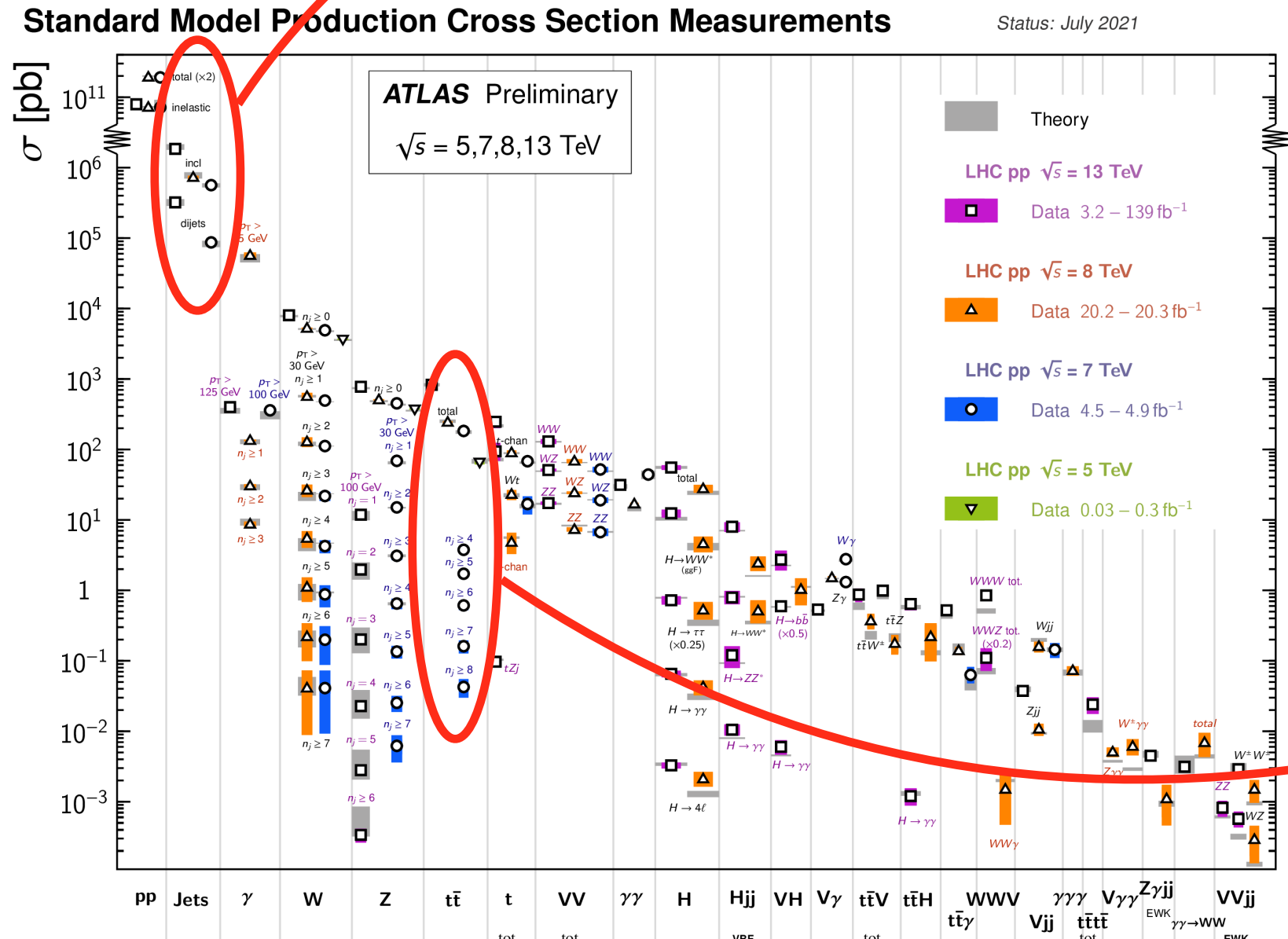


e.g. jet mass (groomed)

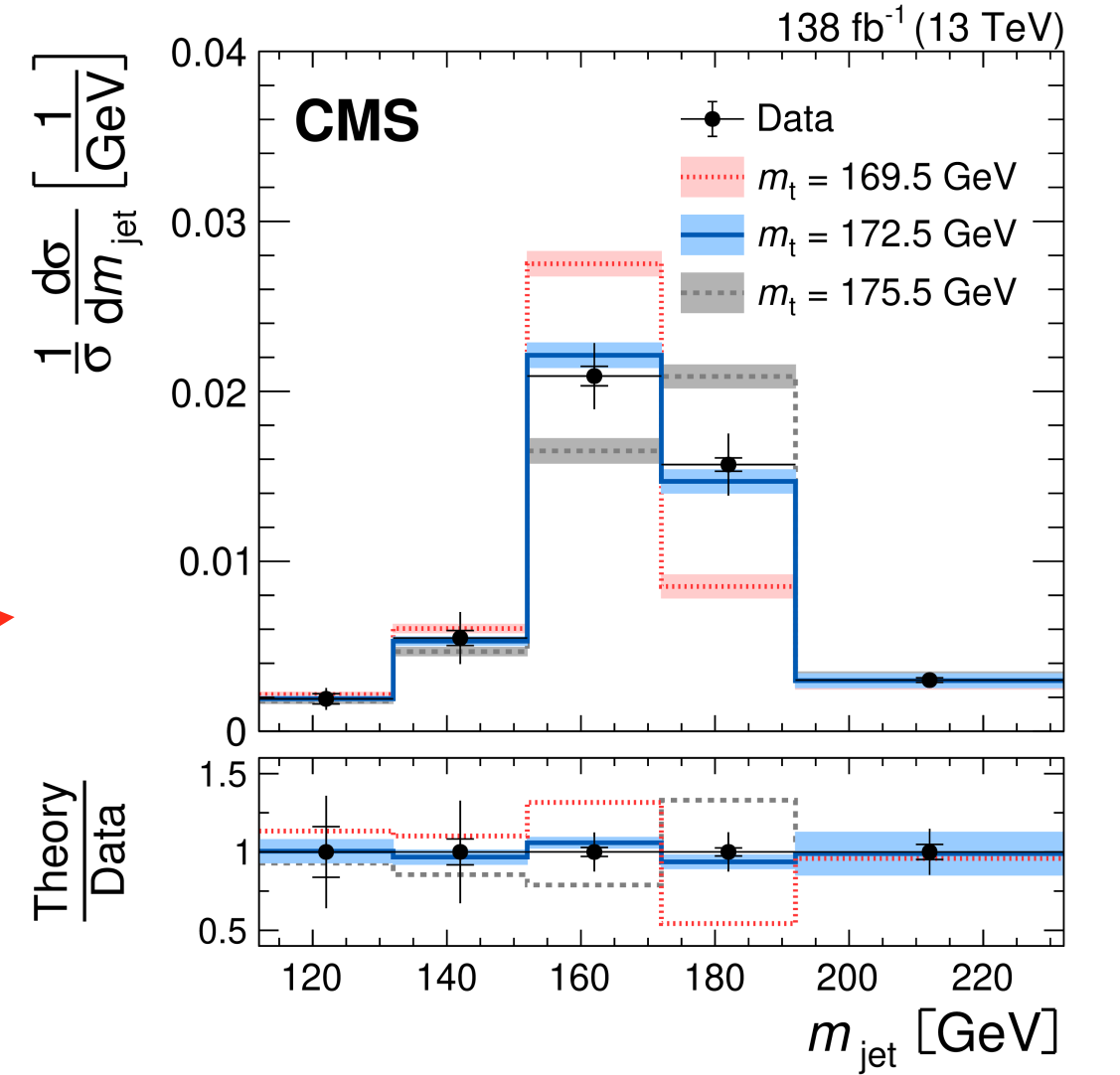
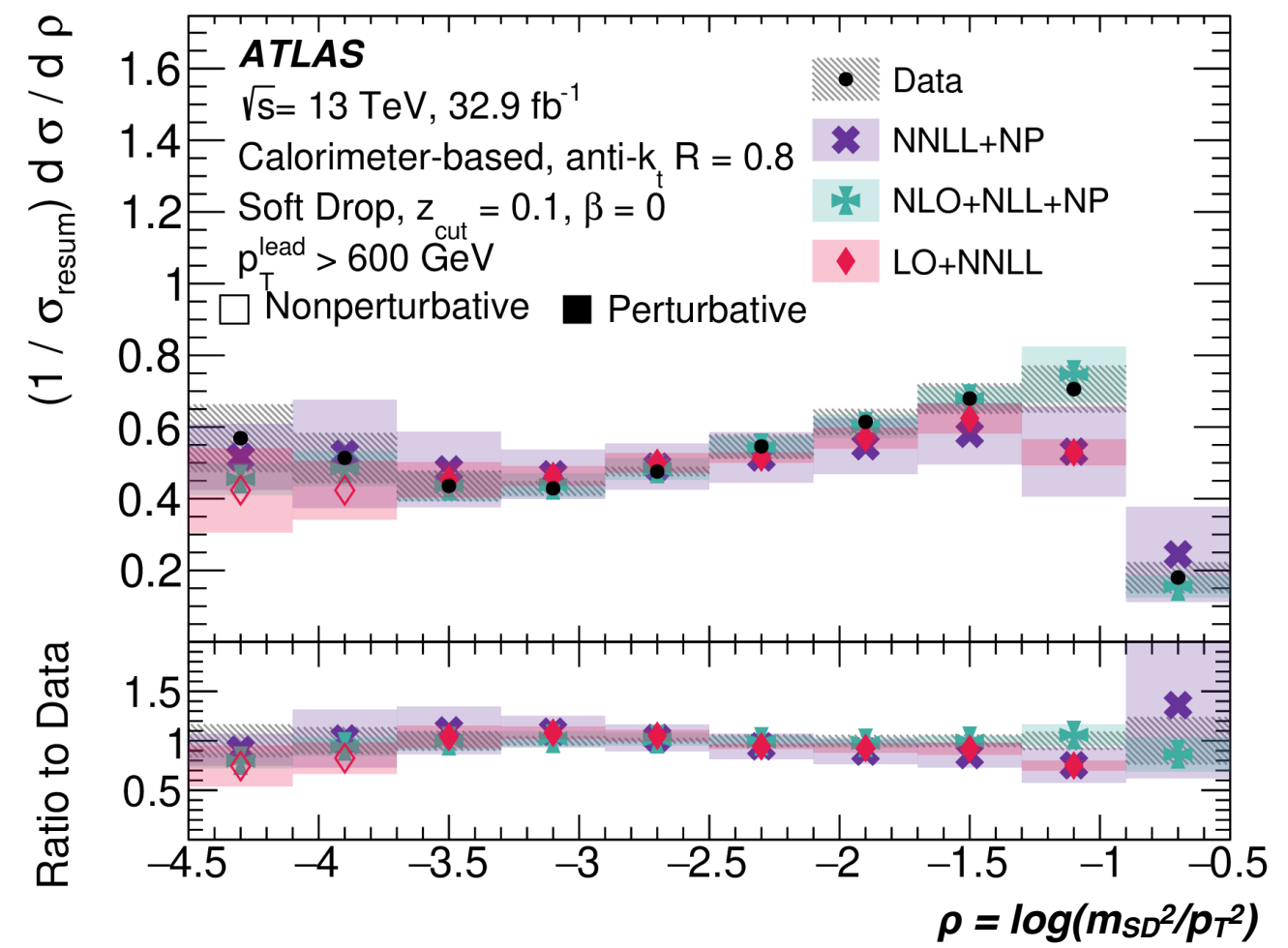
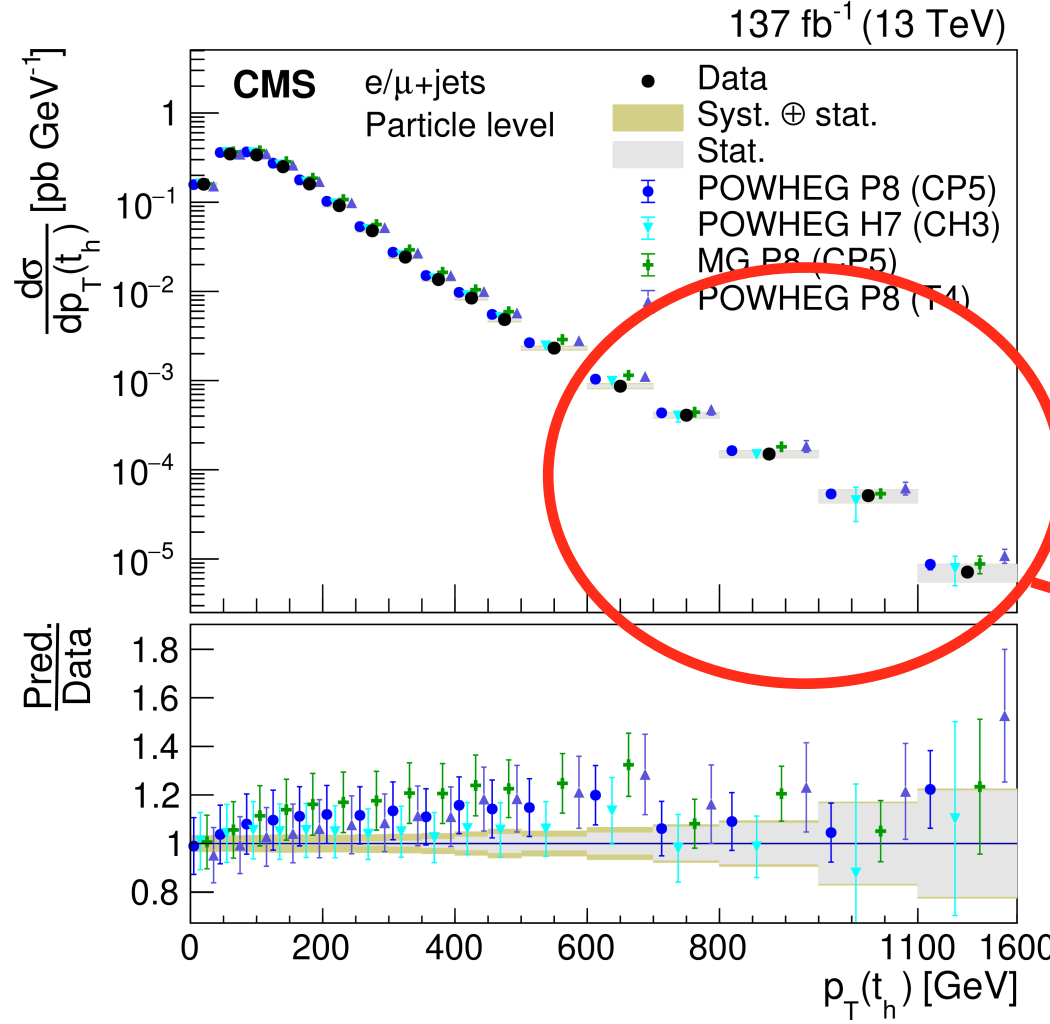
JSS



e.g. top mass (boosted)



e.g. top pairs



... and at the highest scales: **boostemos!**

Today's map

- **Detectors**

- ATLAS, CMS
 - Tracking detectors, calorimeters
 - Pile-up
 - Particle Flow
- ALICE
 - Heavy ions concepts (Centrality, RAA)
- DIS: H1 (HERA), EPIC (EIC)

- **Analysis techniques**

- Tagging
- Unfolding
- *Selected topics, some points might seem obvious to you — goal is for everyone to learn something.*



Genoese map circa 1457, pre-Columbus, Caboto

Today's map

- **Detectors**

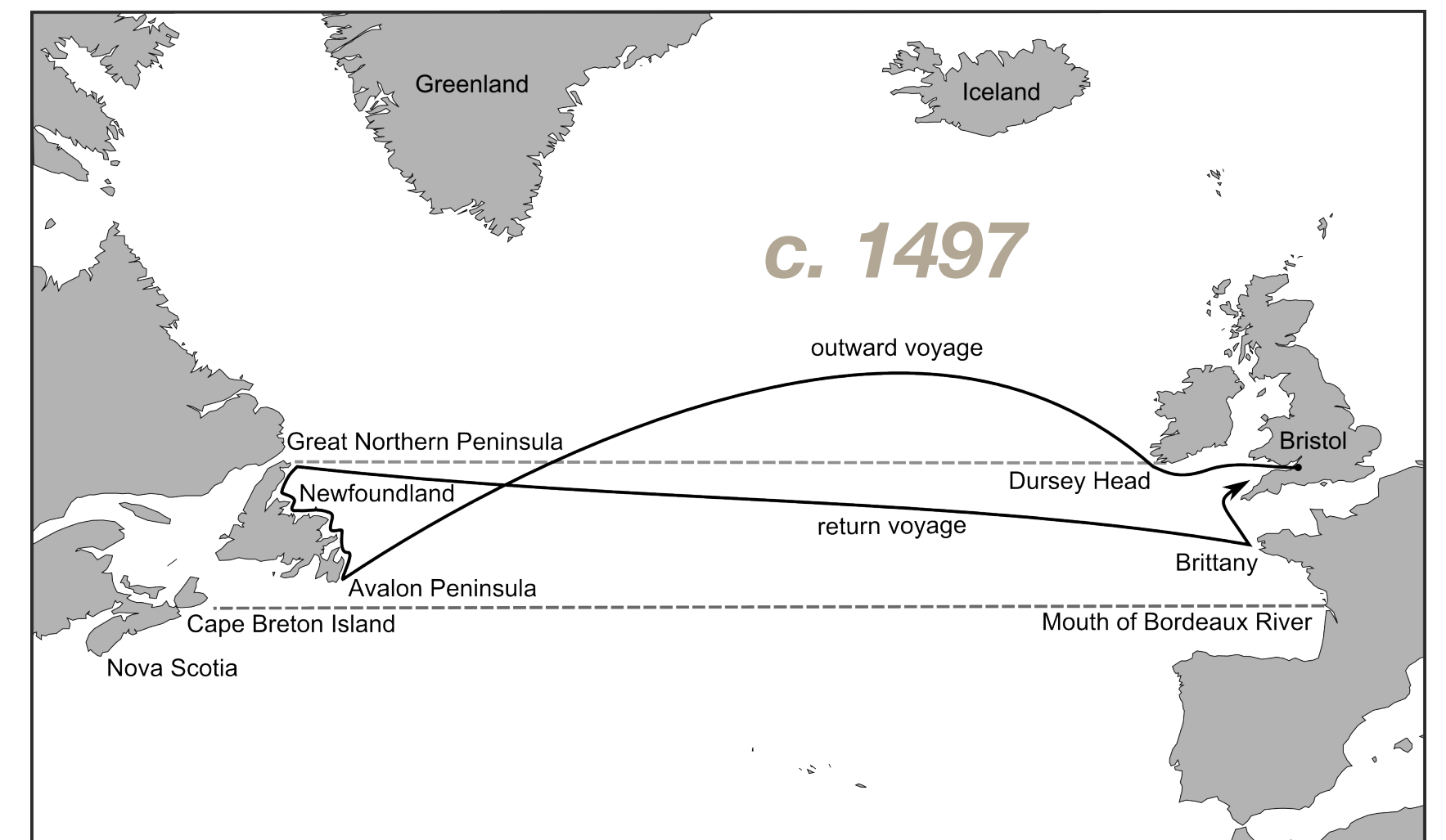
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Genoese map circa 1457, pre-Columbus, **Caboto**





me & Dad watching the
Matthew Aug 24/97

(Halifax 1997)



Monument in Fleming Park (Halifax, Nova Scotia)

Detectors + Reconstruction

Detectors + Reconstruction



Theory Intro
Berkeley, 2023

(Aiming to clarify some fundamentals earlier in the conference this year.)



SUISSE
FRANCE

CMS

LHCb

CERN Prévessin

ATLAS

CERN Meyrin

SPS 7 km

PS 6.28 m

ALICE

LHC 27 km



13.6 TeV Run 3 *pp* centre-of-mass energy
5.02 TeV Run 3 Pb+Pb sqrt(s)
1 billion collisions per-second (**40 MHz**)
1.2E11 nominal protons per-bunch
2808 proton bunches per-beam
1232 superconducting dipole magnets
392 superconducting quadrupoles

LHCb

ATLAS

CERN Meyrin

PS 628 m

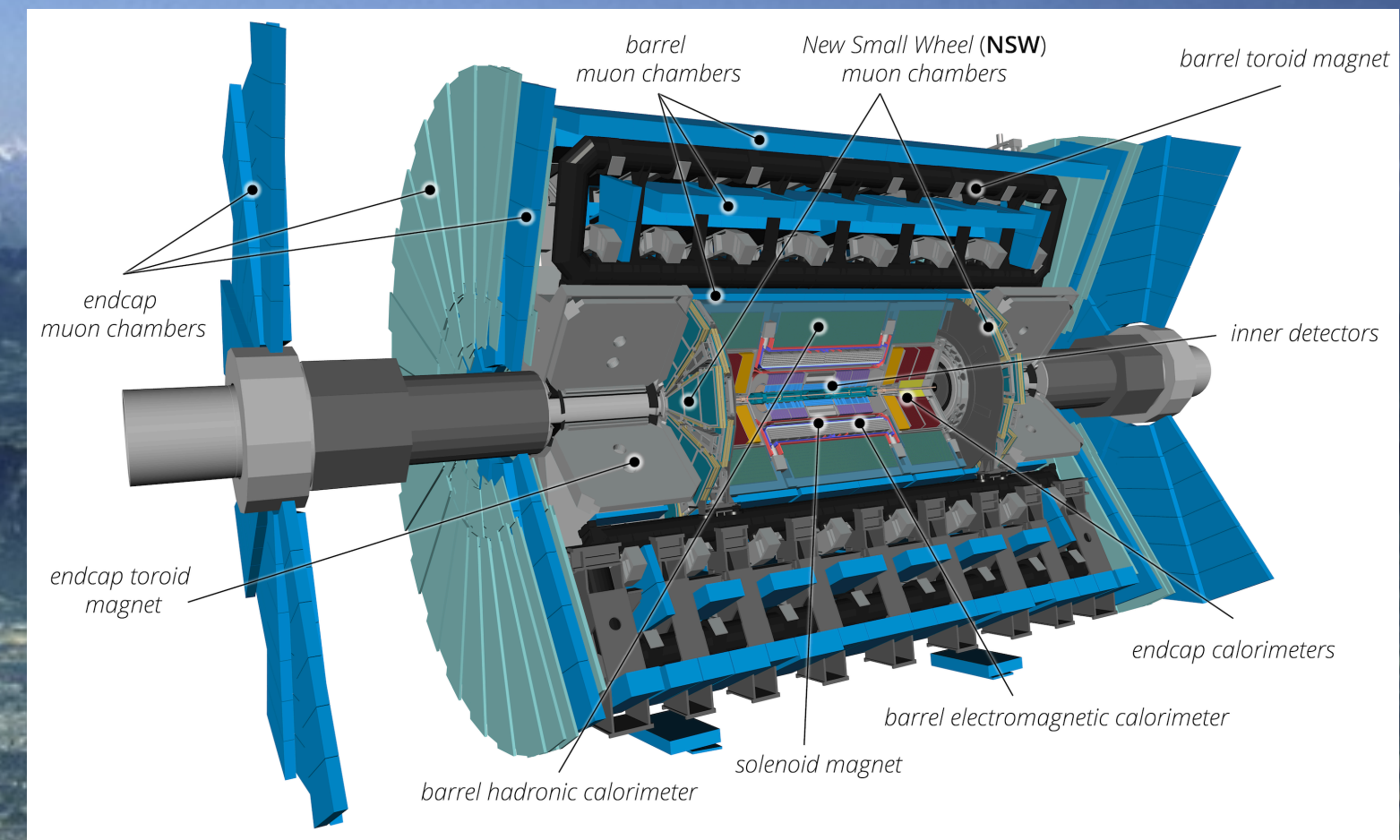
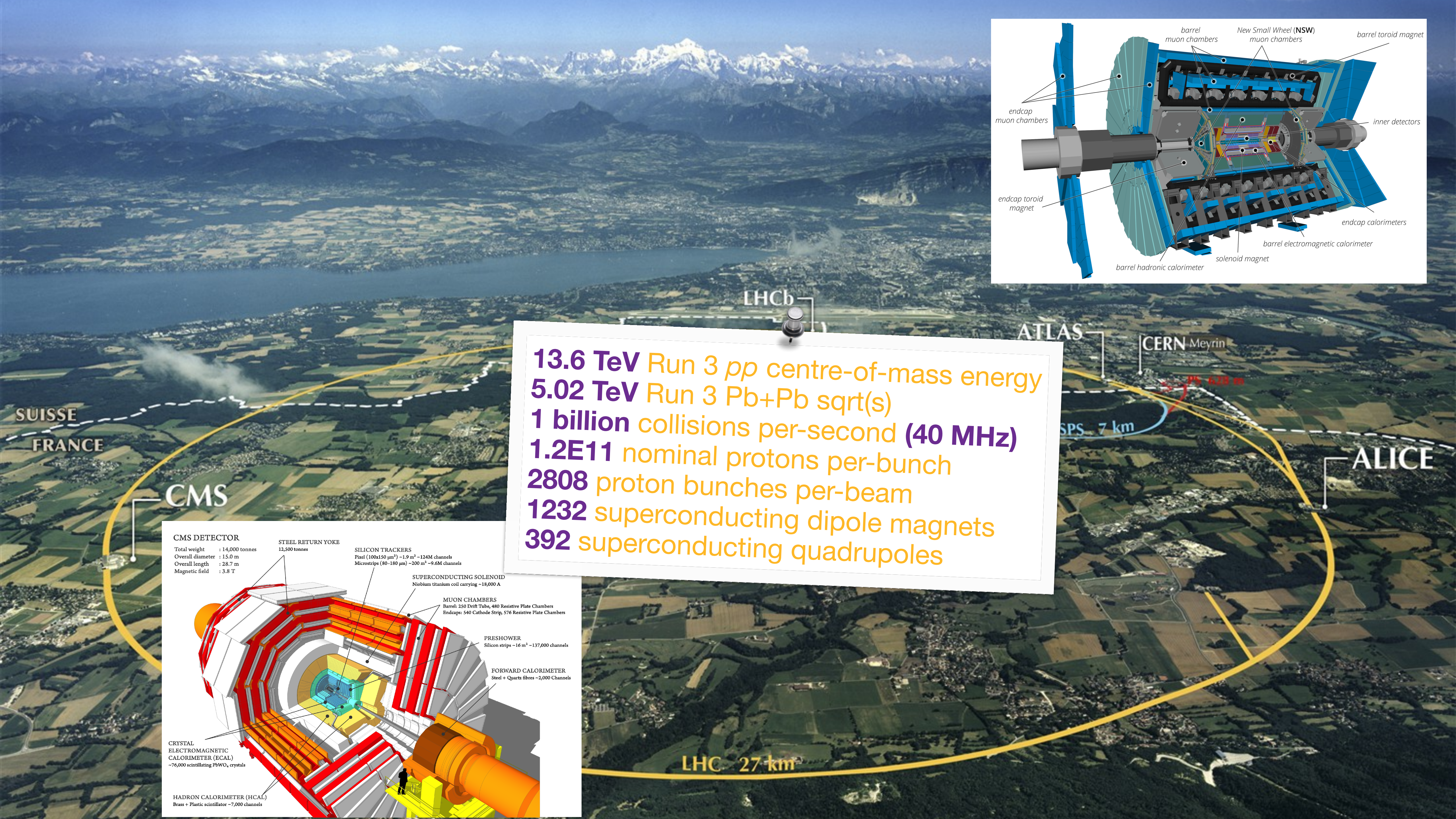
SPS 7 km

ALICE

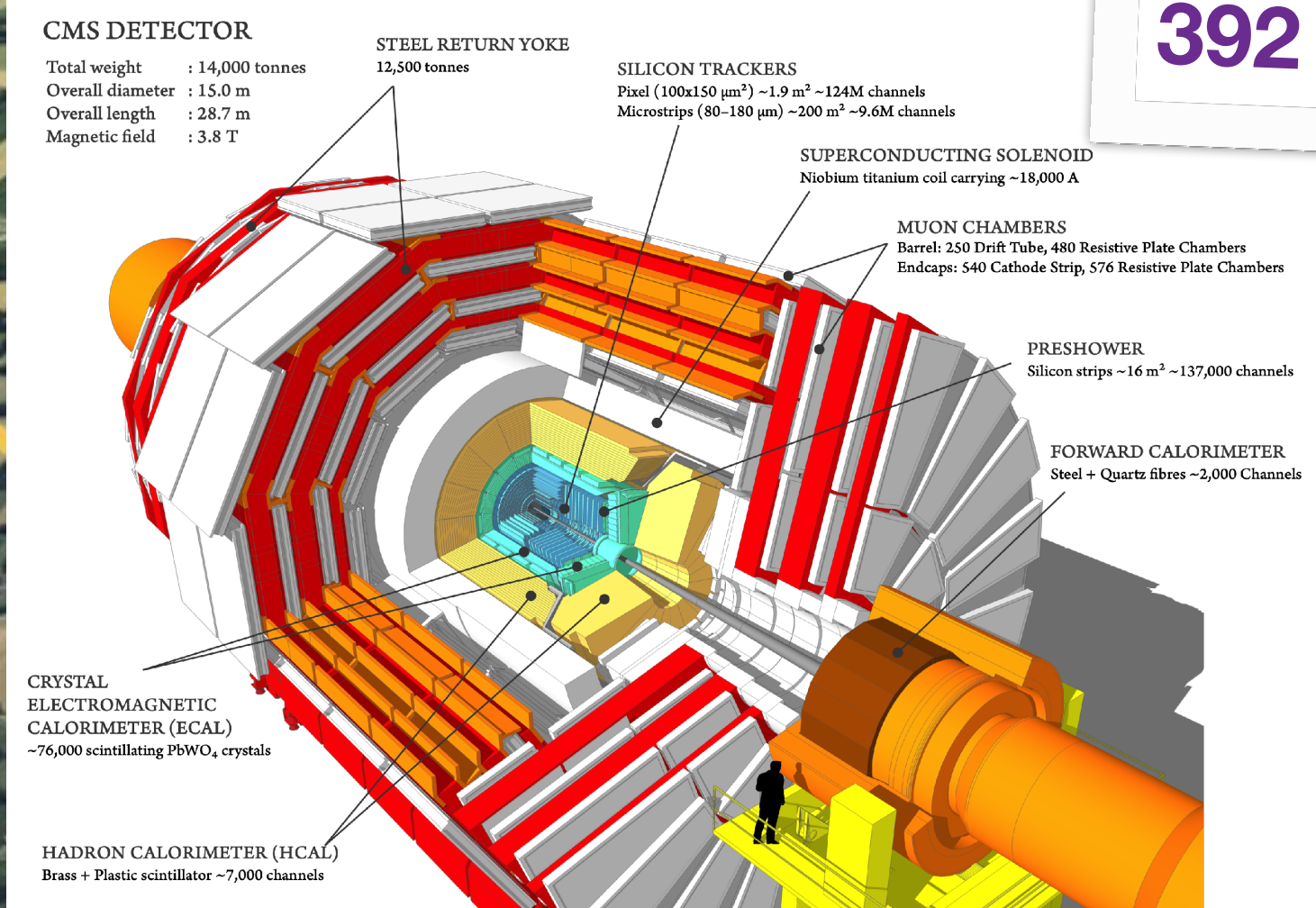
CMS

LHC 27 km

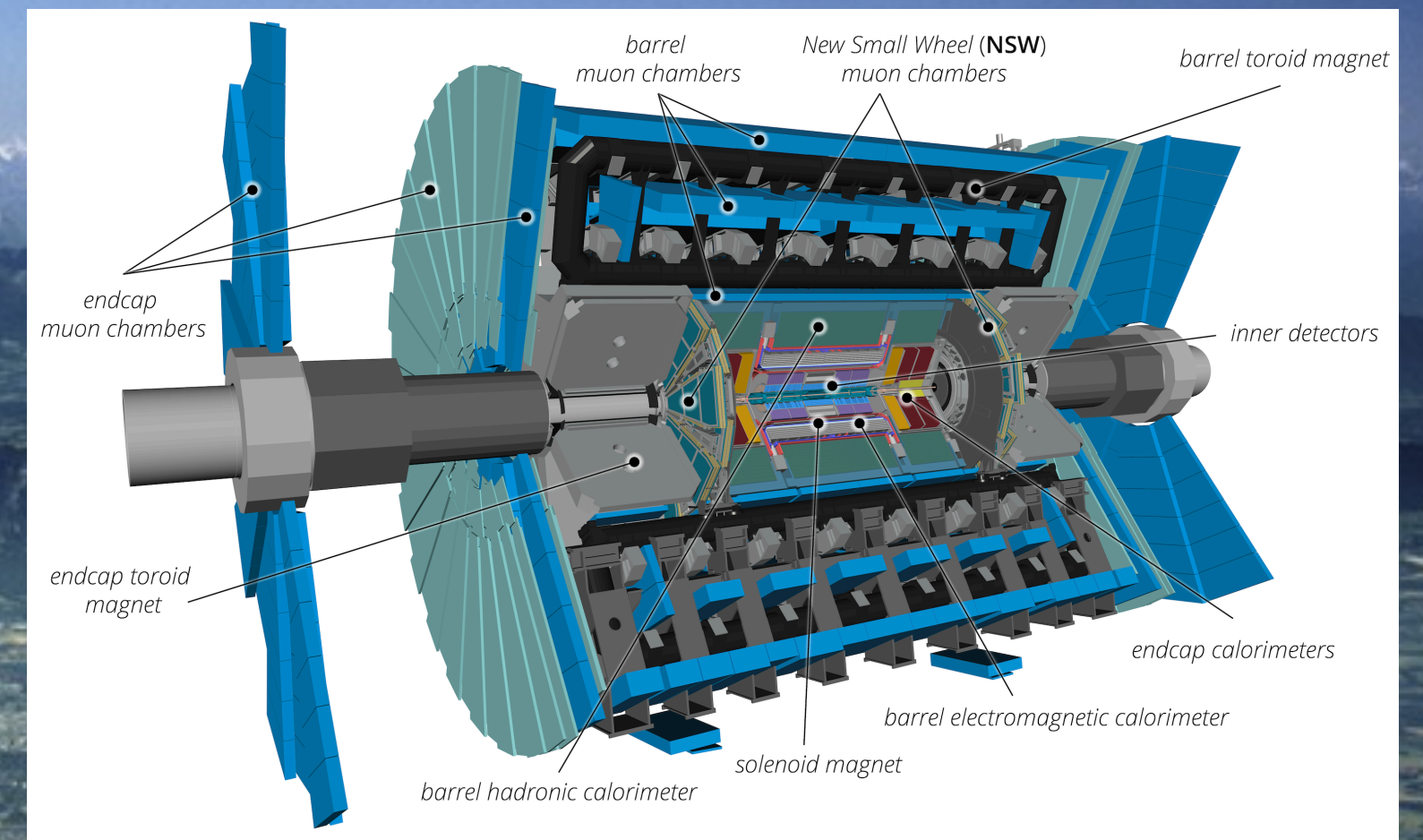
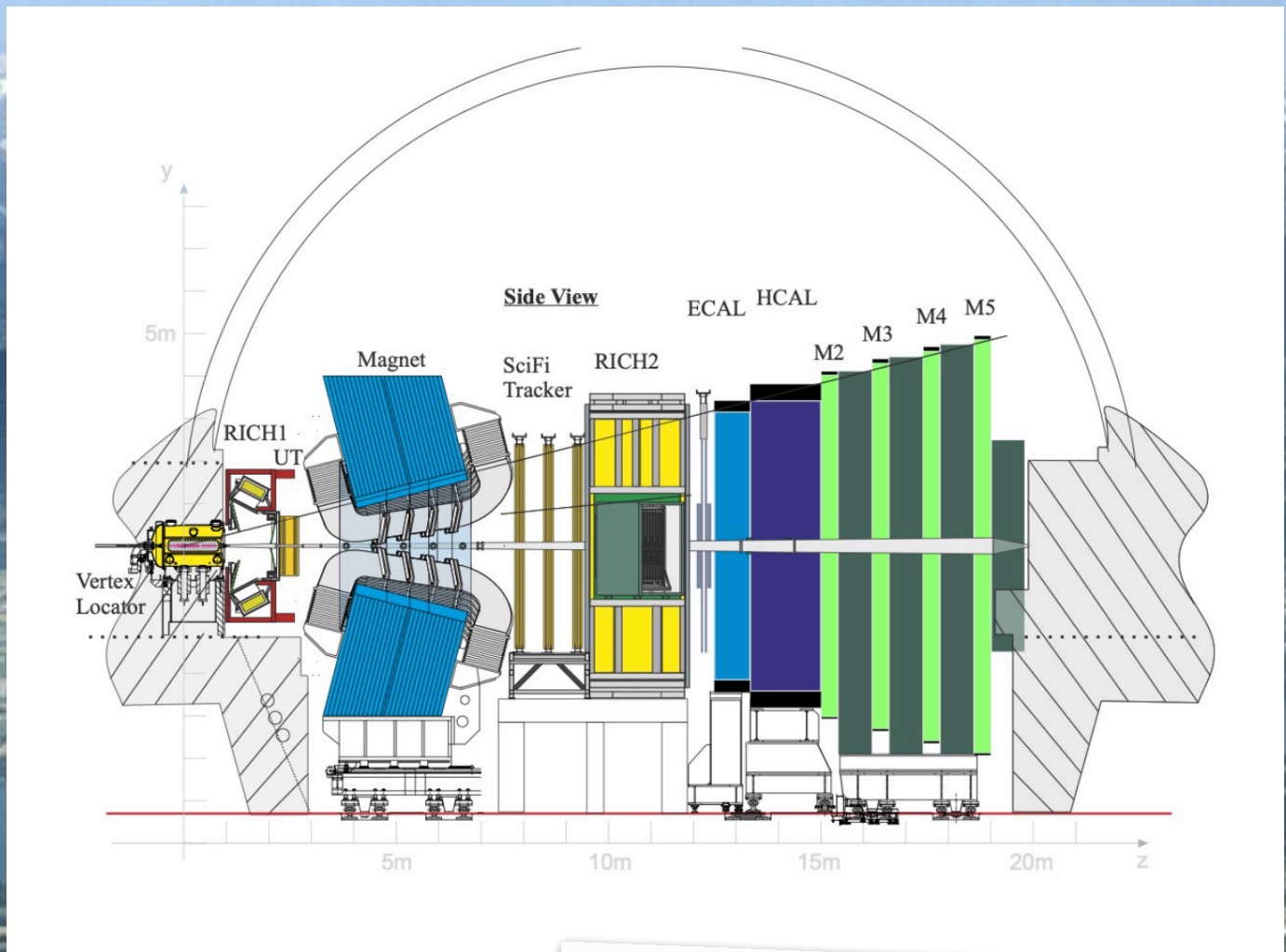
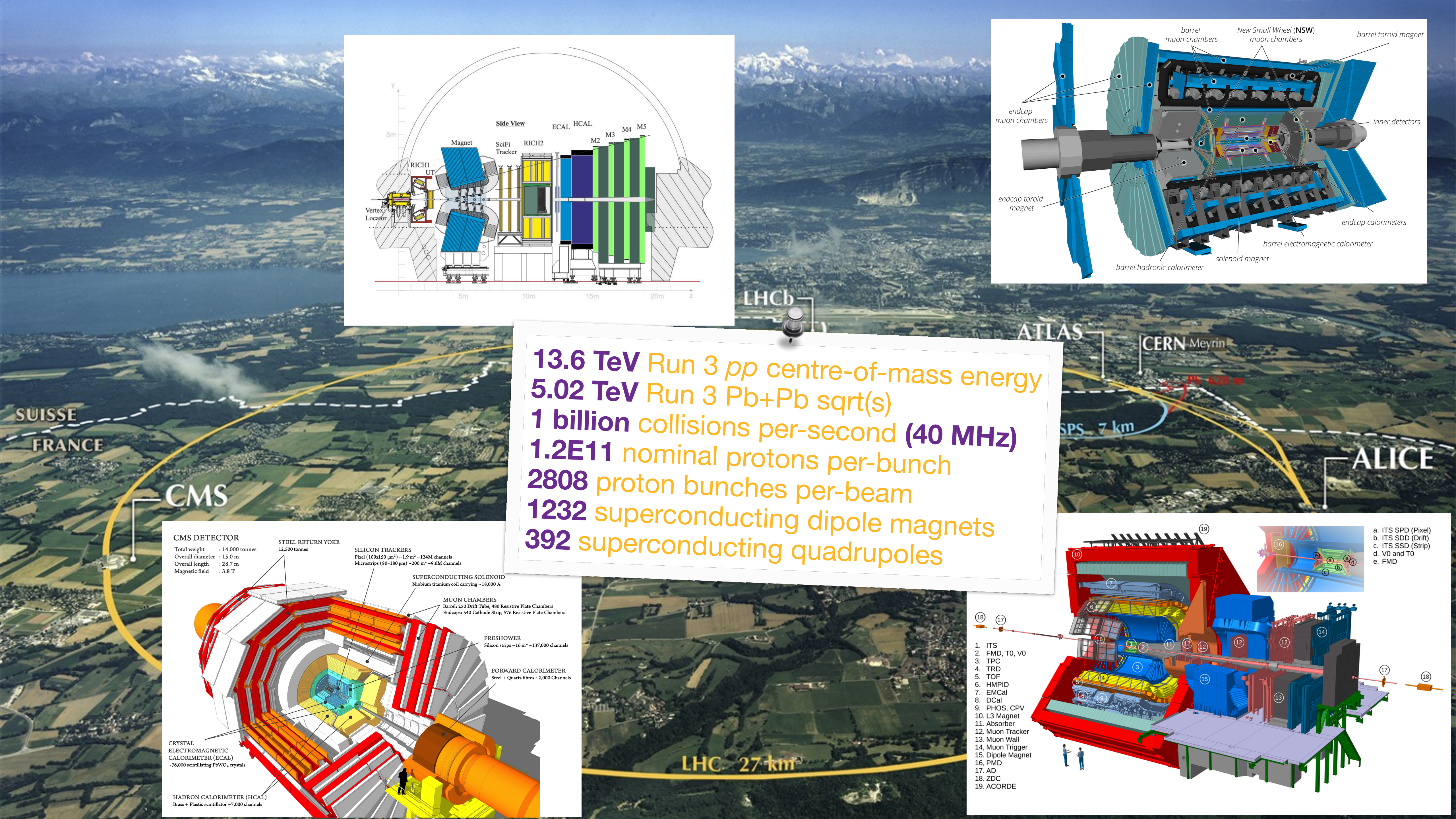
SUISSE
FRANCE



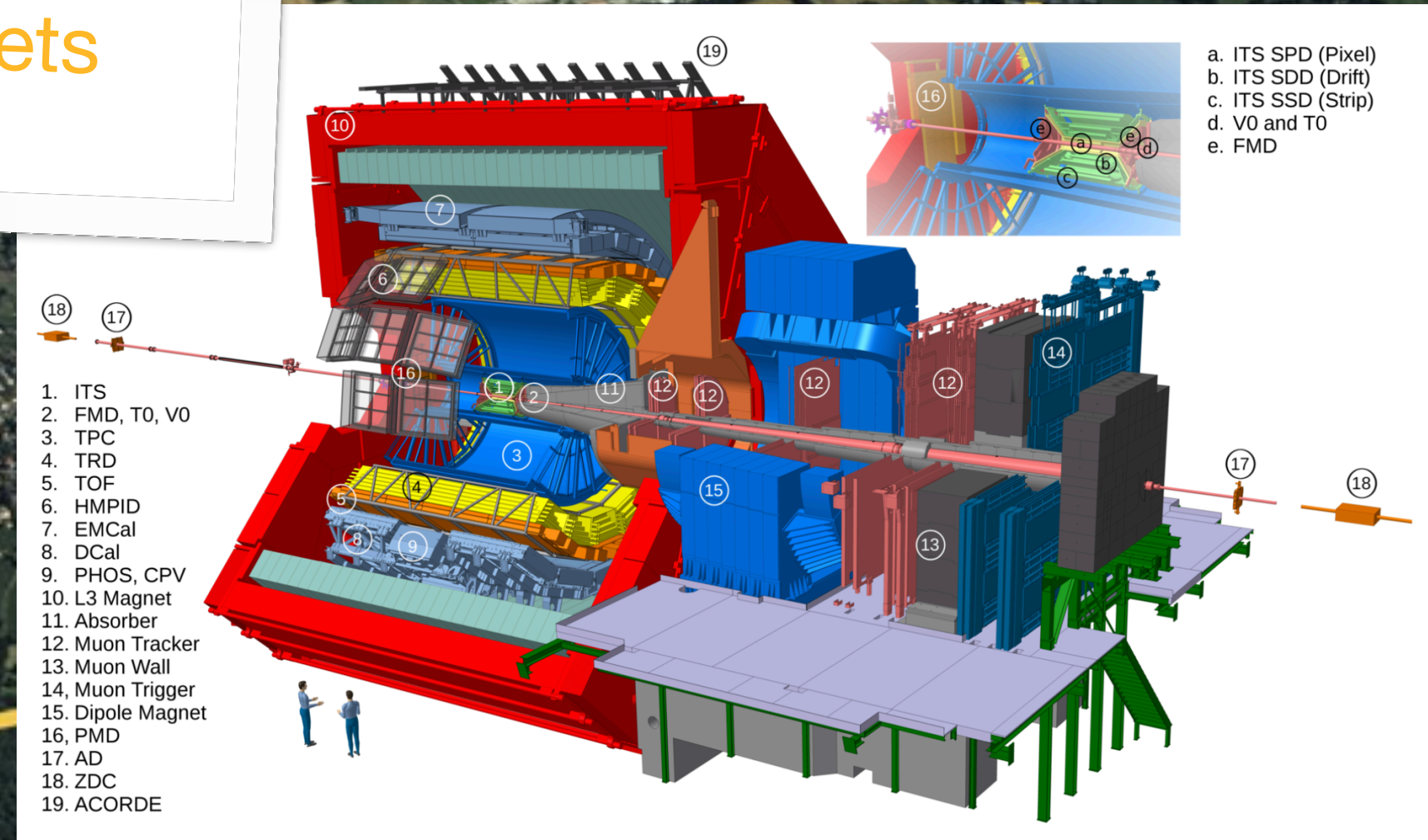
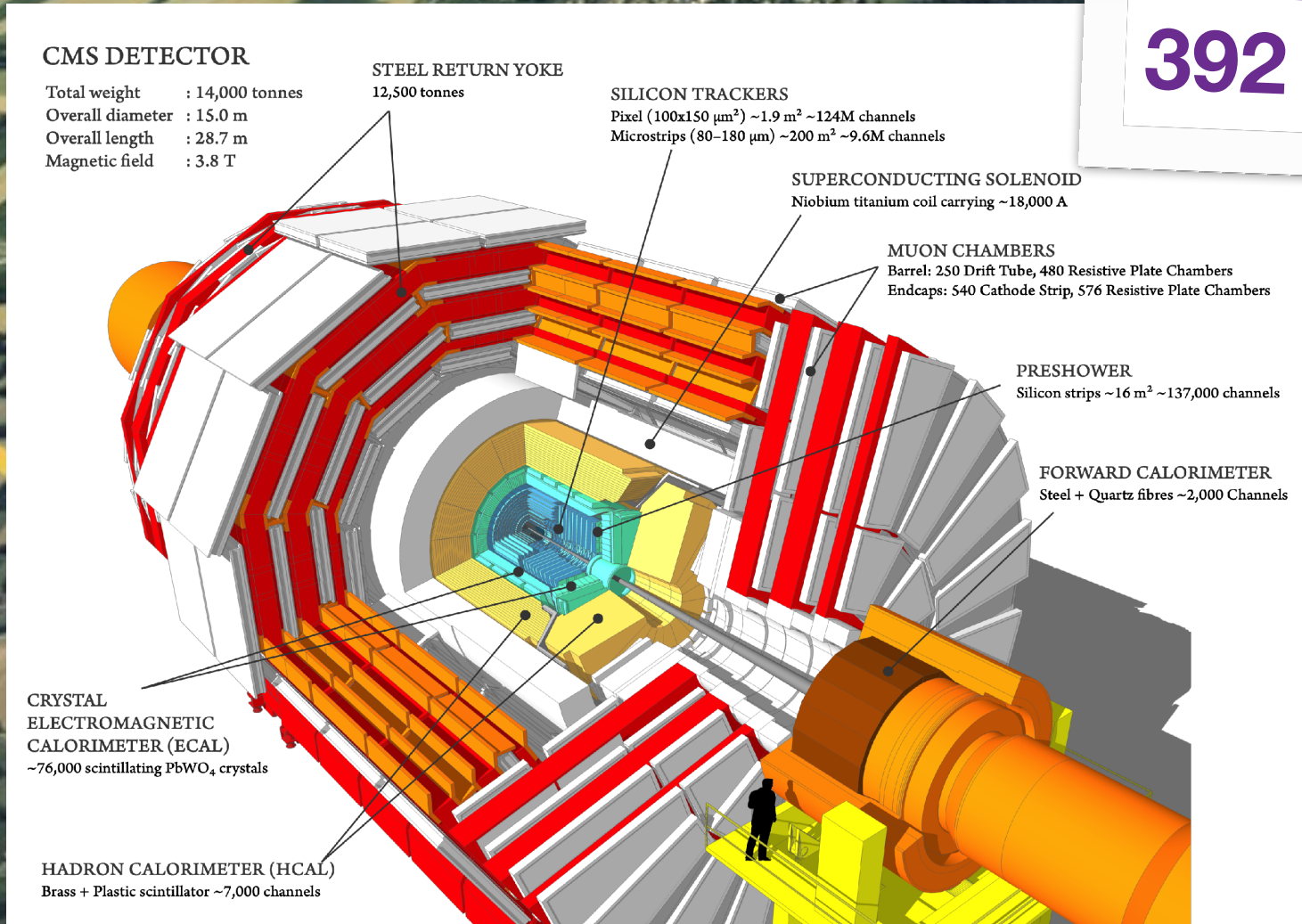
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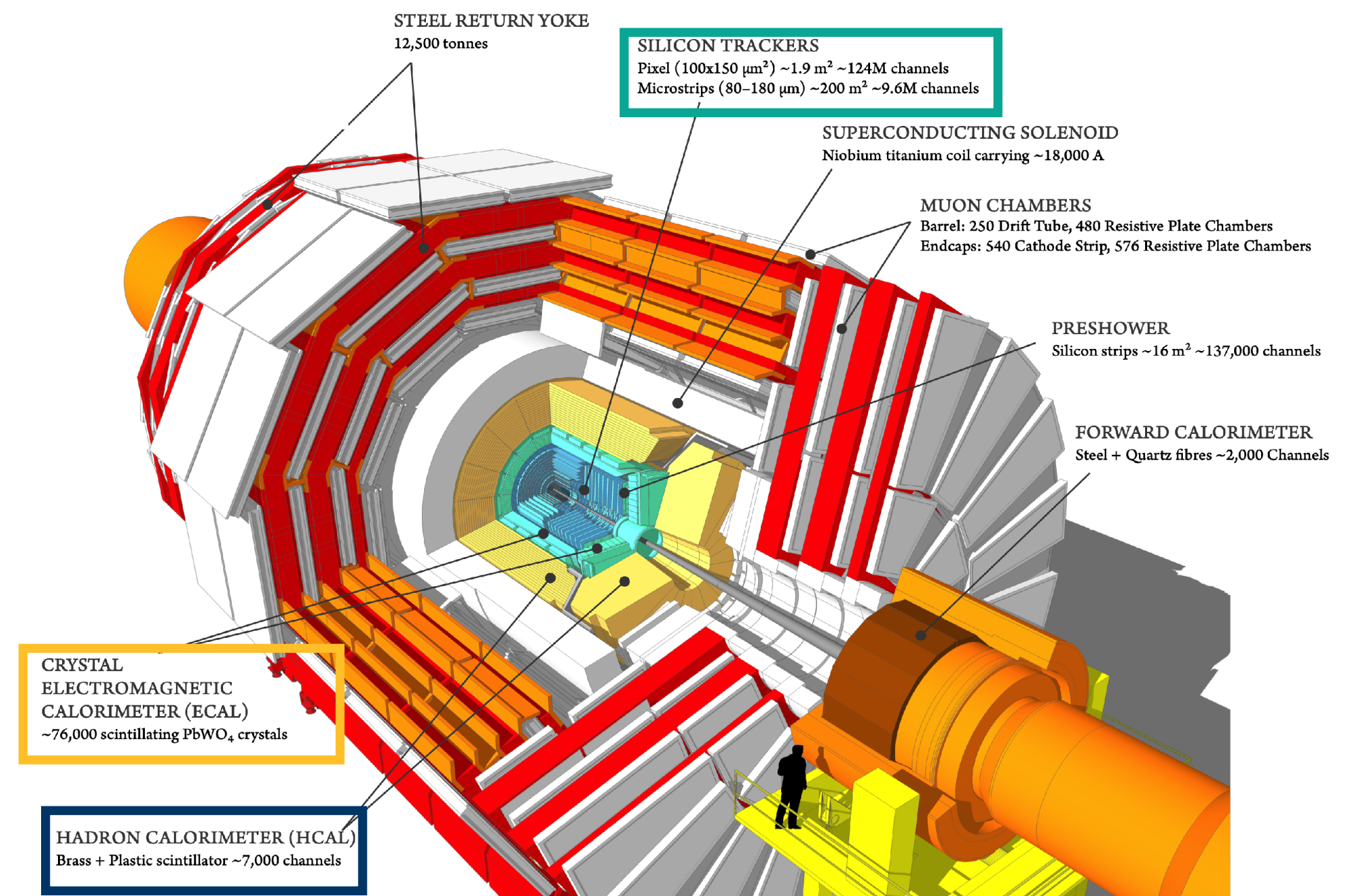
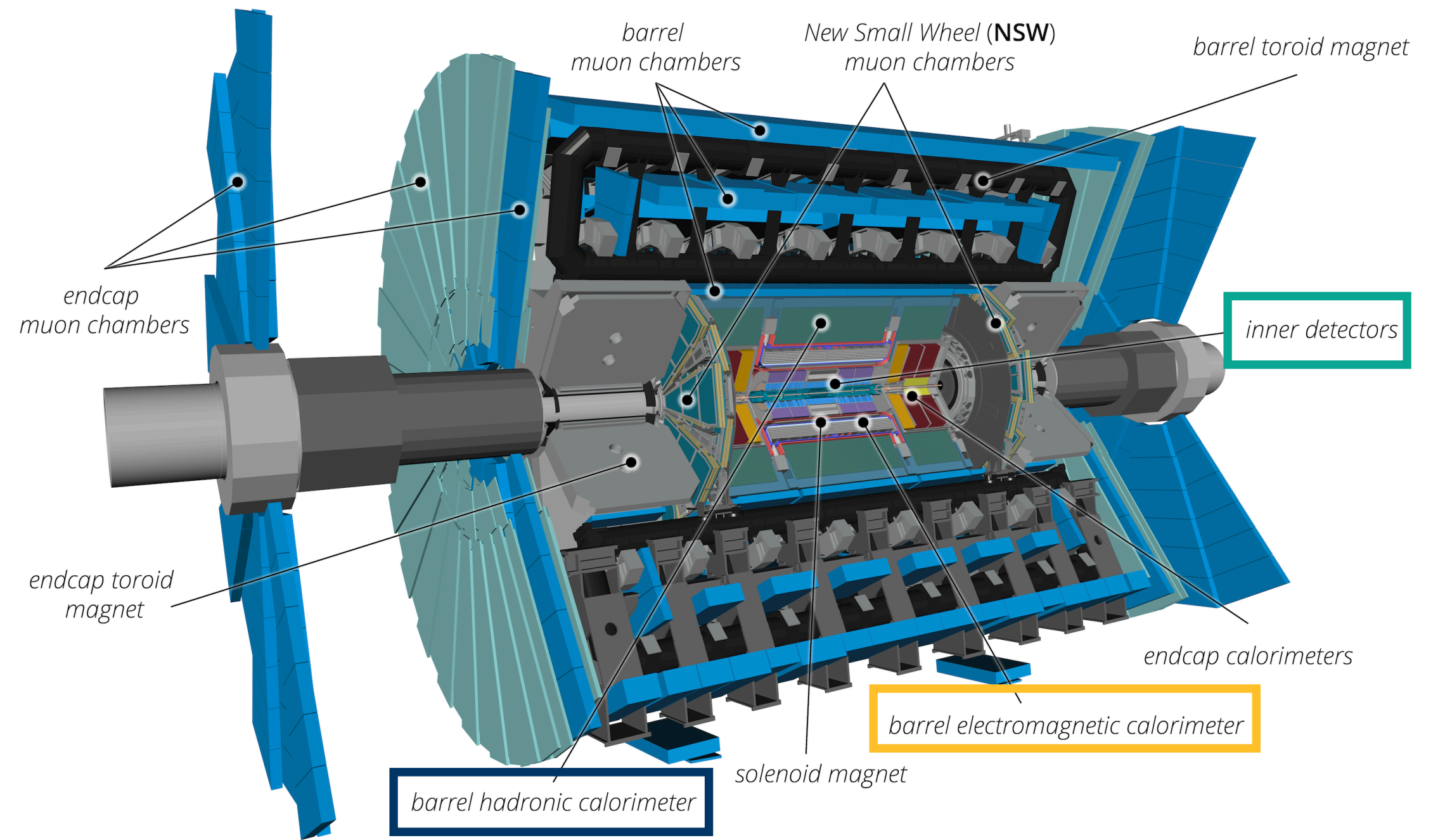
LHC 27 km

CMS & ATLAS

♪ ... are two of a kind / looking for whatever new particles they can find. ♪

- General purpose detectors, similar design goals (discover Higgs + broad BSM sensitivity).
- Implementations quite different in terms of technologies:

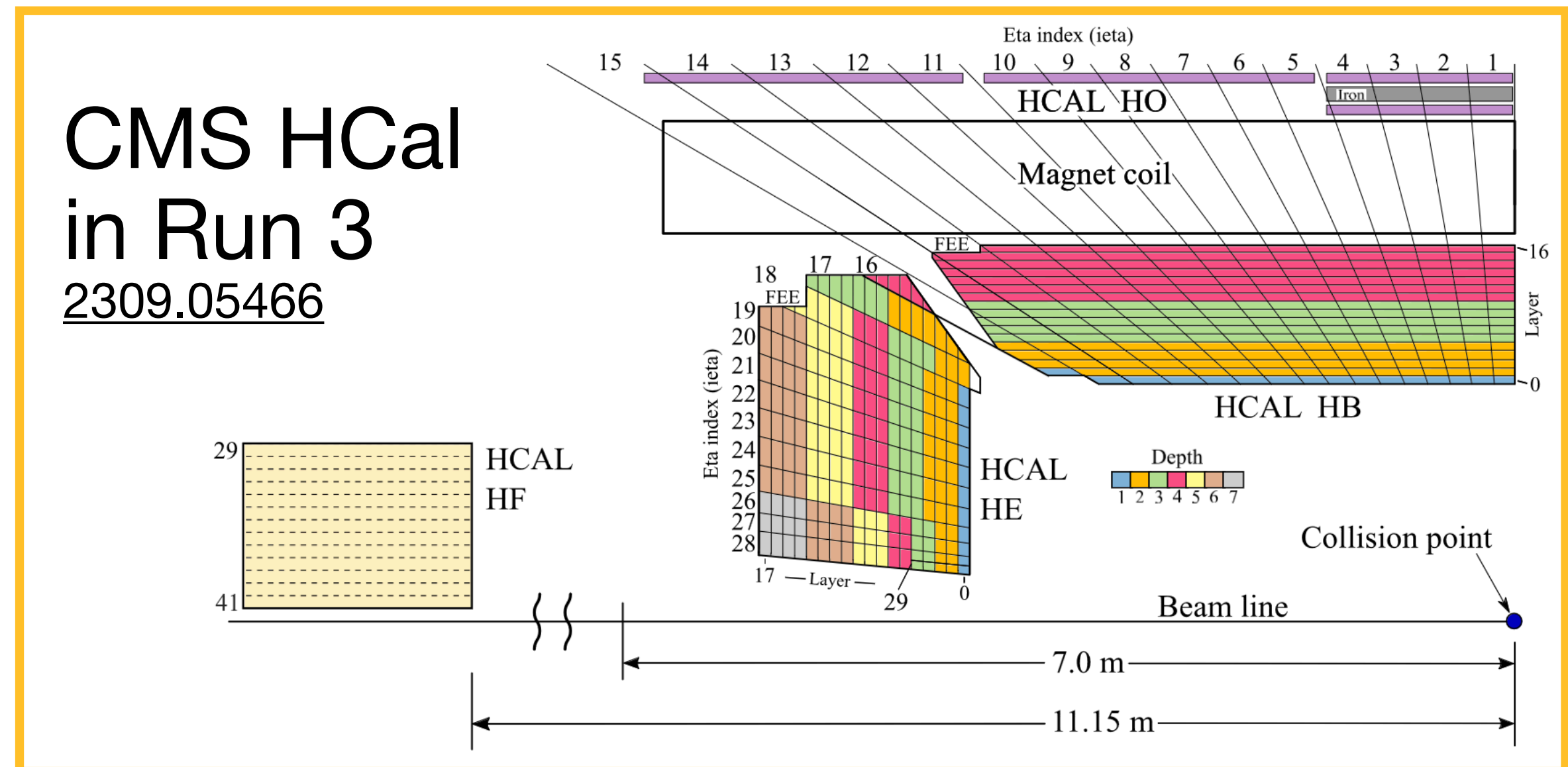
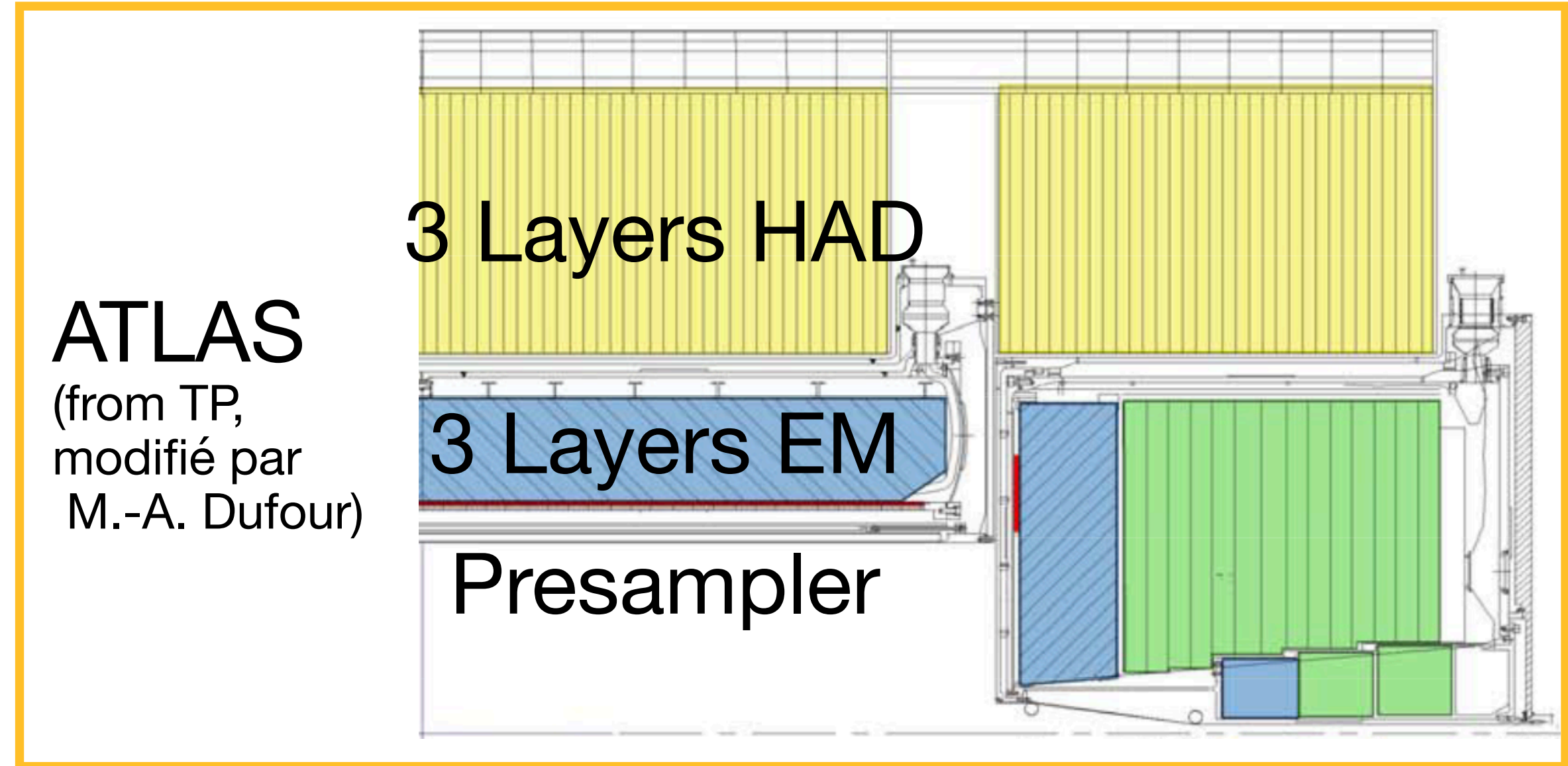
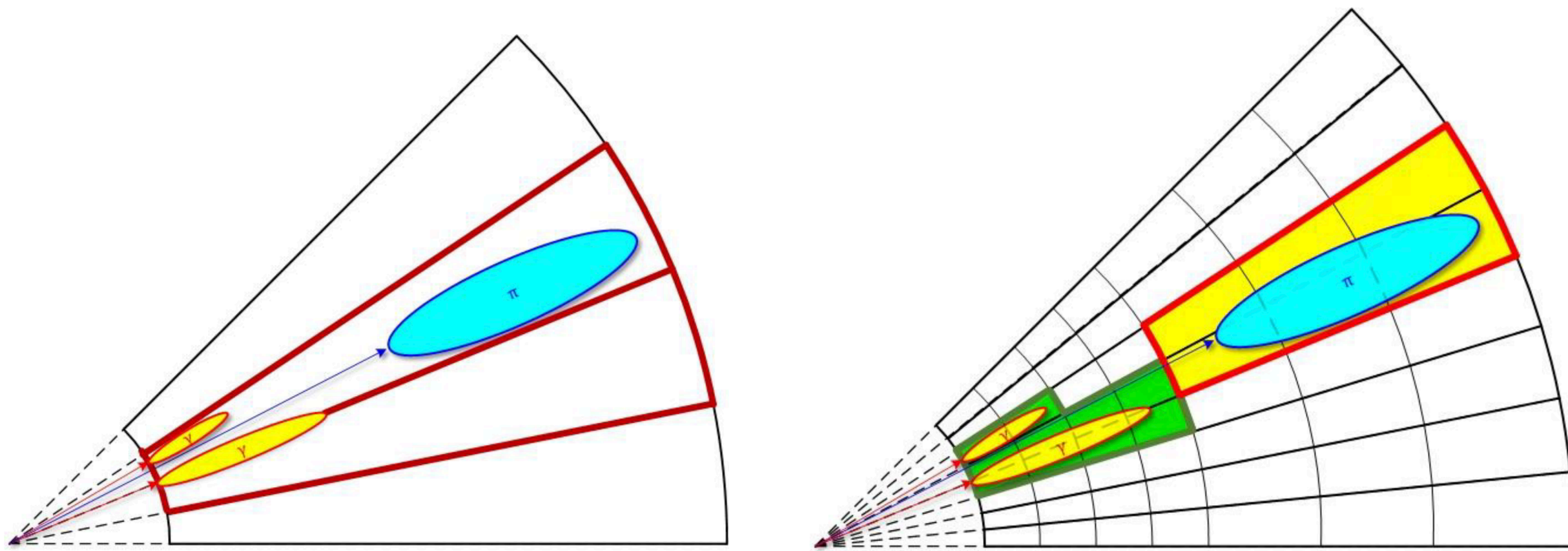
	ATLAS	CMS
Tracker	Silicon+Straw Tube	All-Silicon
ECal (Barrel)	Lead+LAr Sampling, 3 longitudinal segments + Presampler	PbWO4 Crystal Homogeneous + Preshower
HCal (Barrel)	Fe+Scintillators Sampling, 3 readout depths	Brass+Scintillators Sampling, Run 2: 3 readout depths Run 3+: 7 readout depths
Muons	Drift tubes, RPCs, sTGCs, MMs	Drift tubes, RPCs, CSCs, GEMs
Magnetic fields	2 T Solenoidal 2-6 Tm Toroidal	3.8 T Solenoid



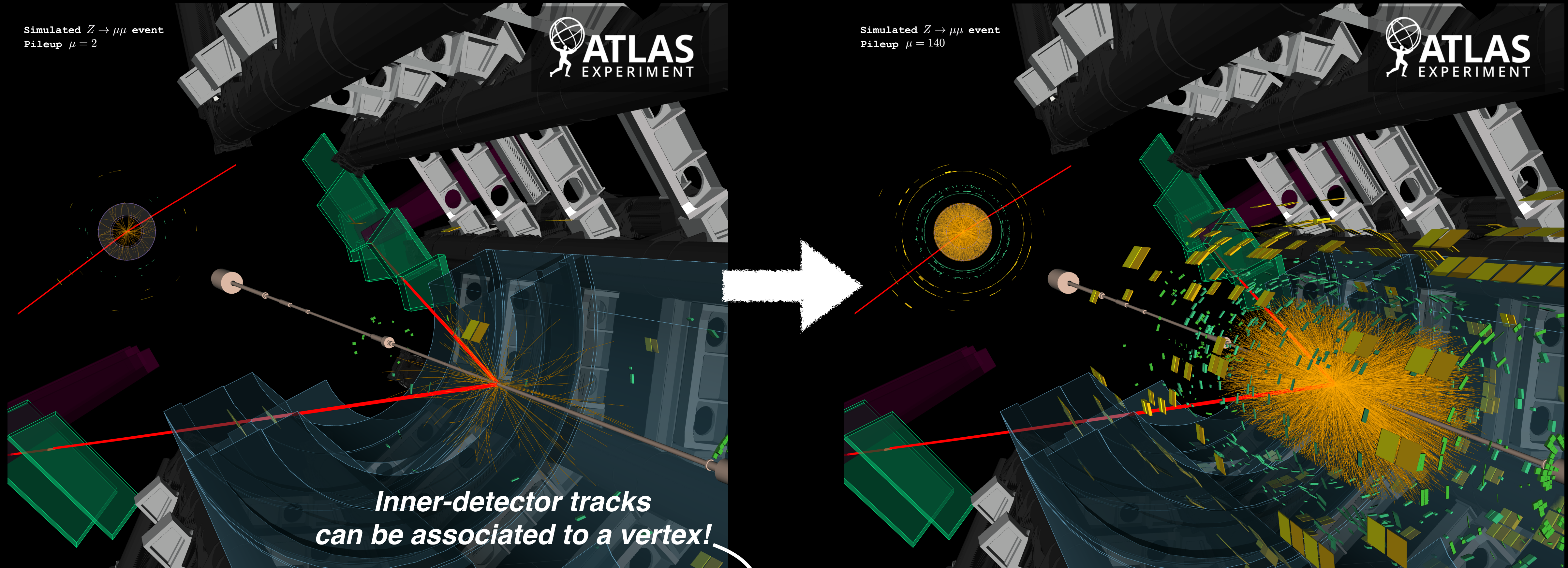
CMS & ATLAS: Segmentation

P. Loch @ UniGe 2017, <https://dpnc.unige.ch/seminaire/talks/loch.pdf>

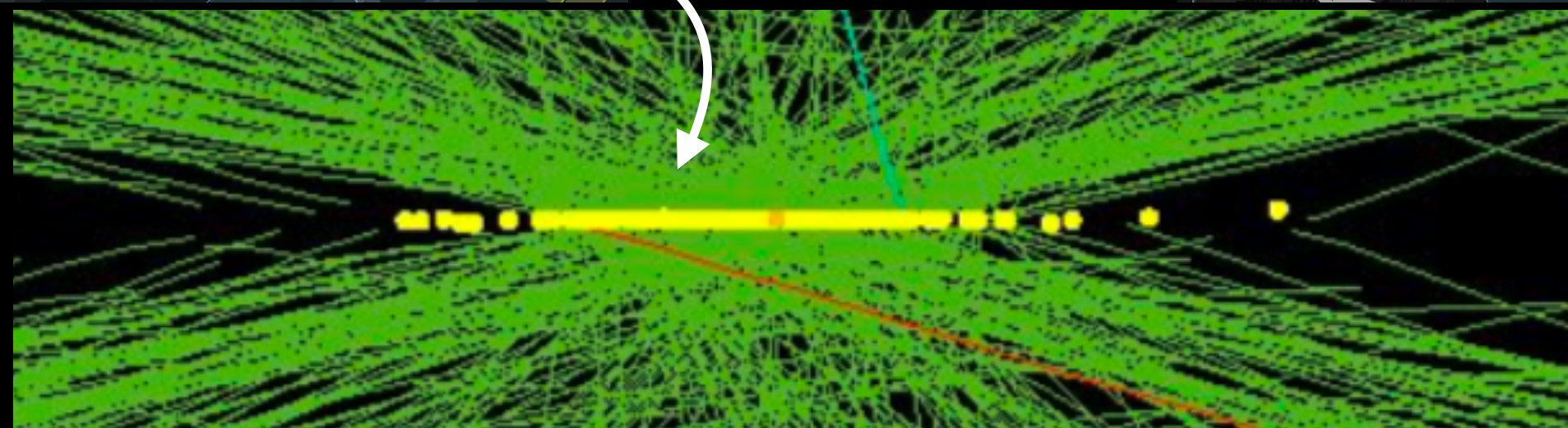
- Longitudinal segmentation of calorimeters important for JSS reco at LHC.
 - Reconstruct kinematics & structure while reading out less volume:
 - Less electronics/detector noise
 - Less **pile-up** (next slide)
 - Better signal calibration shapes (EM/HAD)
 - ATLAS *Eur. Phys. J. C* 77 (2017) 490,
 - P. Loch @ H1, 1992



“Pile-up” → uncorrelated, additional pp collisions in same bunch-crossing as signal process.



2010 pile-up
($\mu \sim 2$)



Run 4 pile-up
($\mu \sim 140$)

Pileup

CMS, JINST 15 (2020) P09018

ATLAS, Eur. Phys. J. C 81 (2021) 334

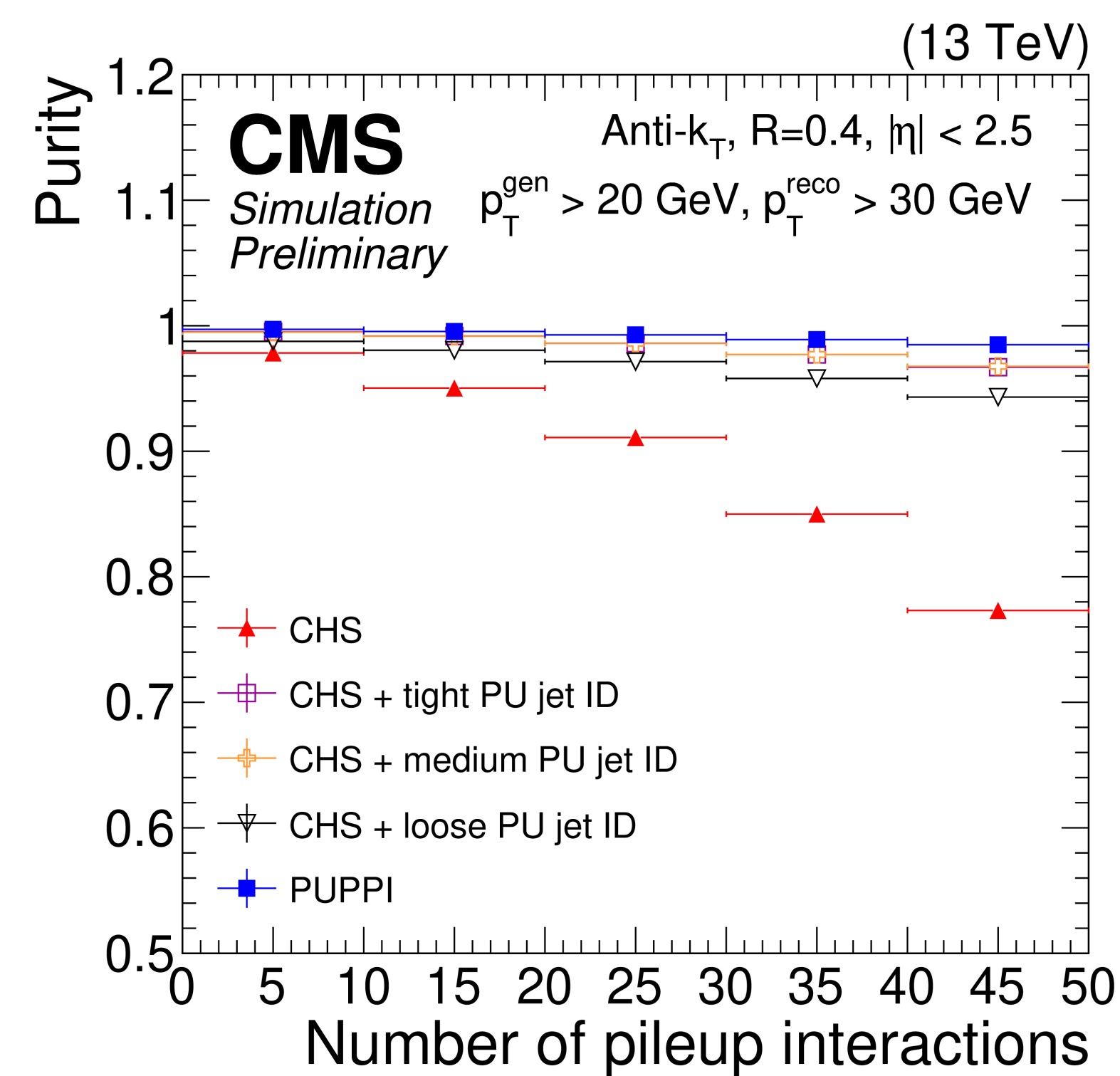
- Uncorrelated pp interactions that occur in the same bunch-crossing as the 'hard scattering' of interest
 - Effect on hadronic objects pronounced:
 - Jets have large catchment area
 - JSS sensitive to soft radiation
 - MET requires balanced energy flow
 - Spurious jets!
(uncorrelated w/ signal of interest)
- Modern approach: **mitigate bit-by-bit, at each step** of reconstruction.
 - Constituent level (e.g. PUPPI, CS+SK)
 - Jet level (e.g. Soft-Drop grooming)

Pileup

CMS, *JINST* 15 (2020) P09018

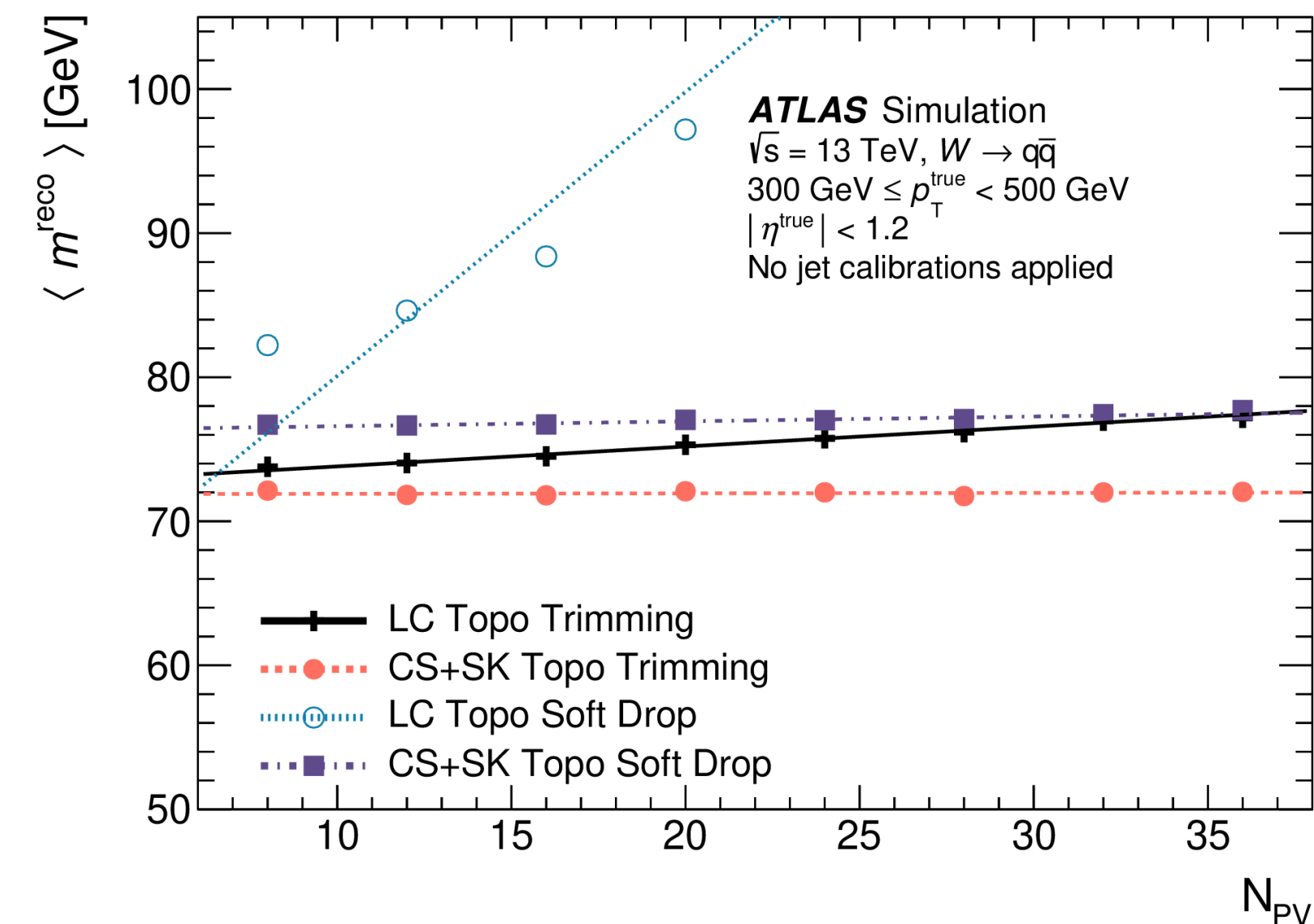
ATLAS, *Eur. Phys. J. C* 81 (2021) 334

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PUPPI improves on CHS+PU jet ID rejection of spurious jets.

(Closer to 1 is better!)



Average jet mass increases with pileup (Flat is good!)

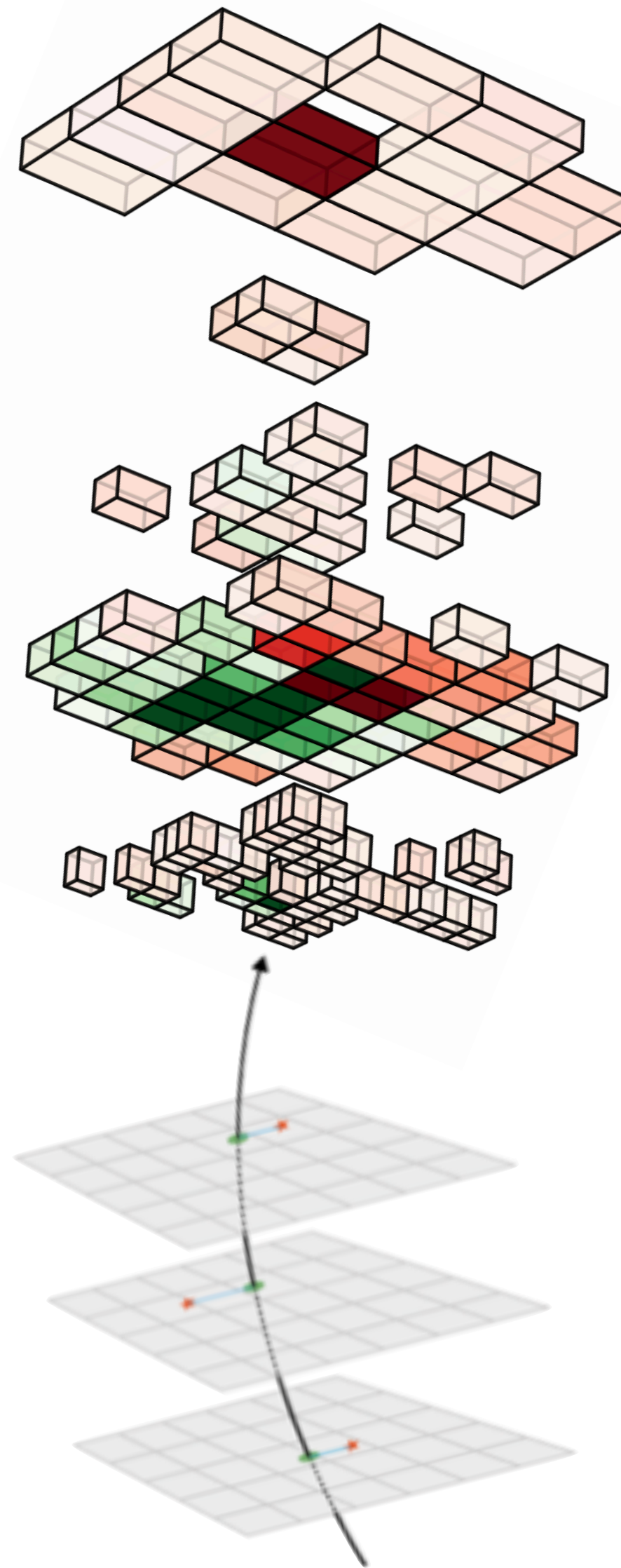
CMS & ATLAS : Particle Flow

Calorimeter

- Good energy resolution (**high- p_T** particles)
- Coarse angular resolution
- Susceptible to pile-up noise
- Retains neutral component
- Central+Forward coverage

Tracker

- Good energy resolution (**low- p_T**)
- Fine angular resolution
- Stable w.r.t. pile-up (vertex association)
- Lacks neutral energy flow
- Only central coverage until Run 4



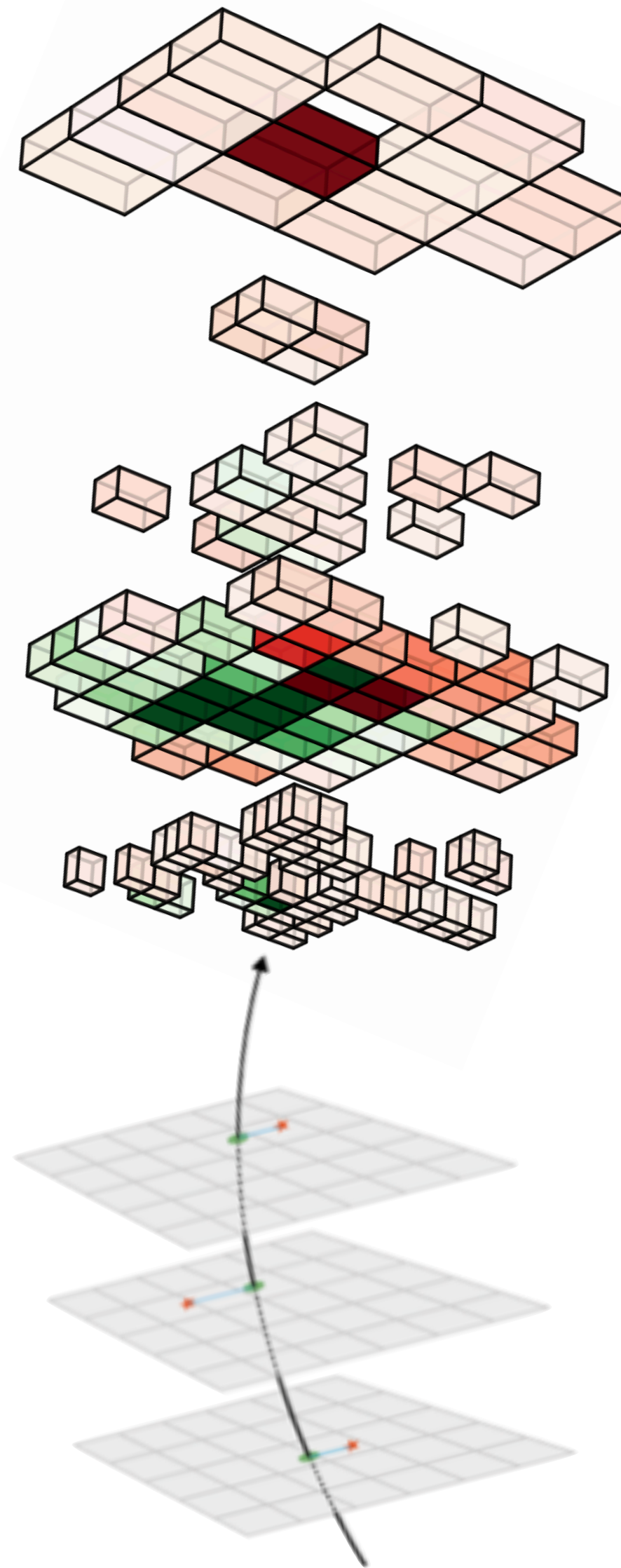
CMS & ATLAS : Particle Flow

Calorimeter

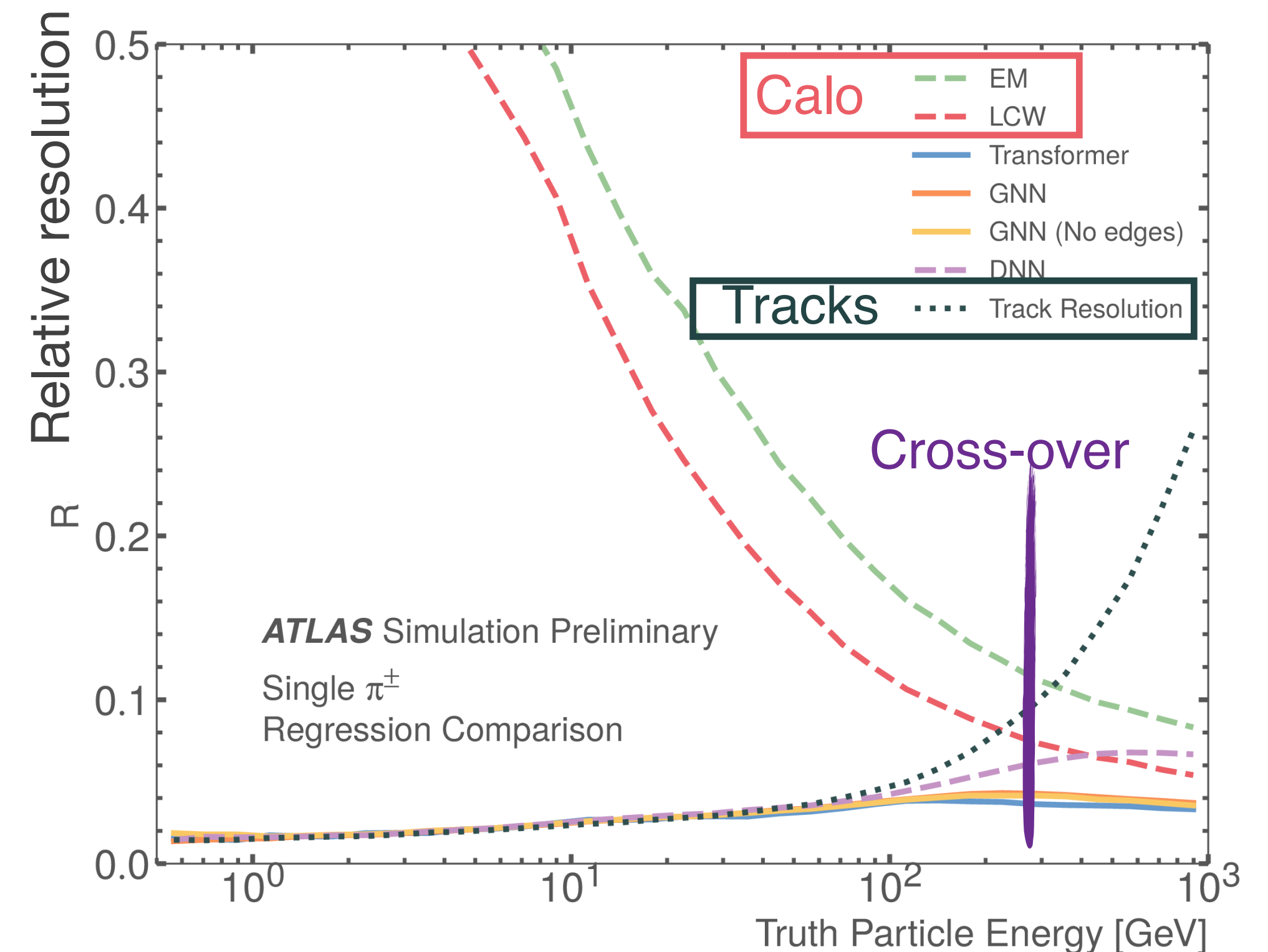
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Both ATLAS & CMS now use Particle Flow algorithms for both small- and large-radius jet reconstruction!



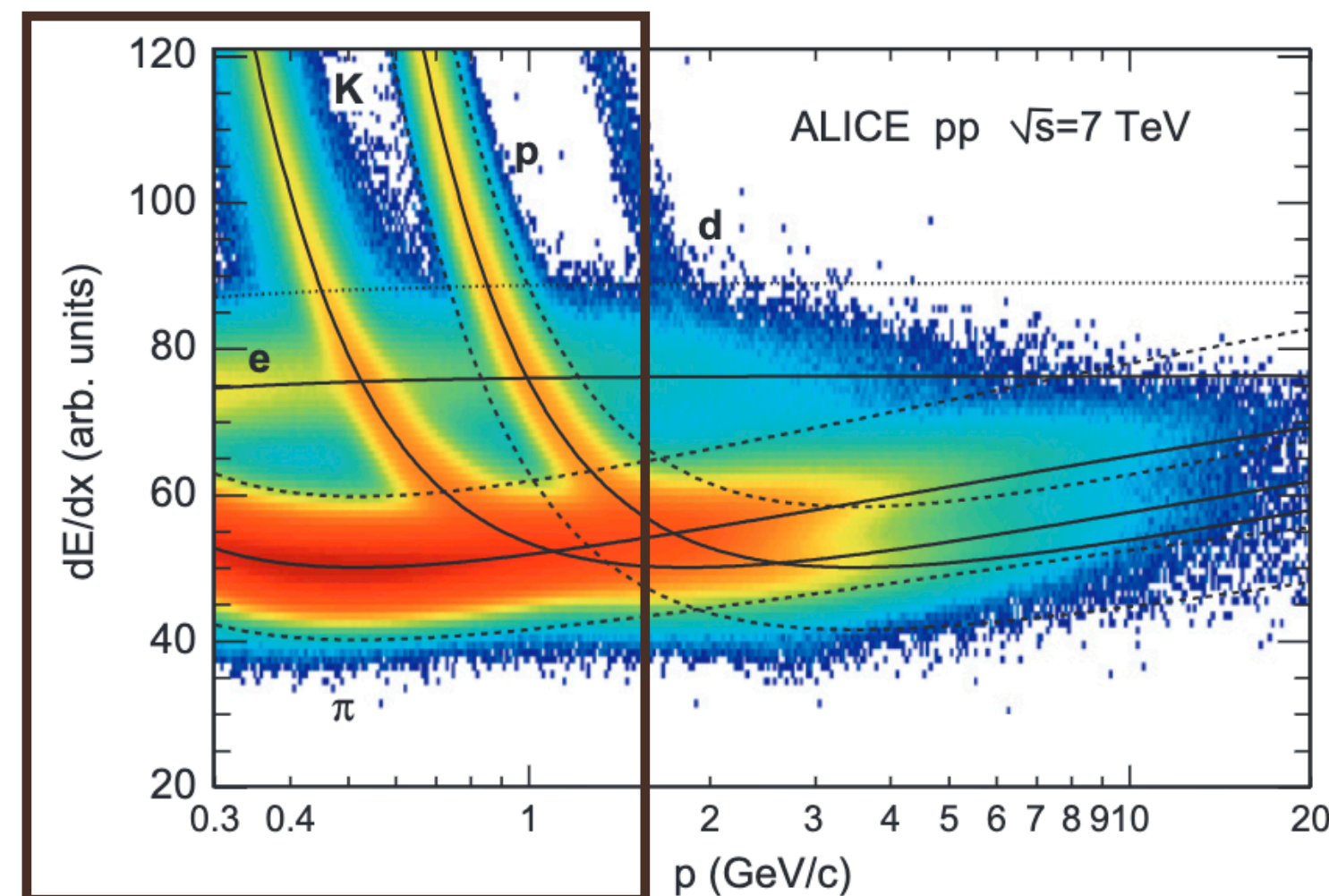
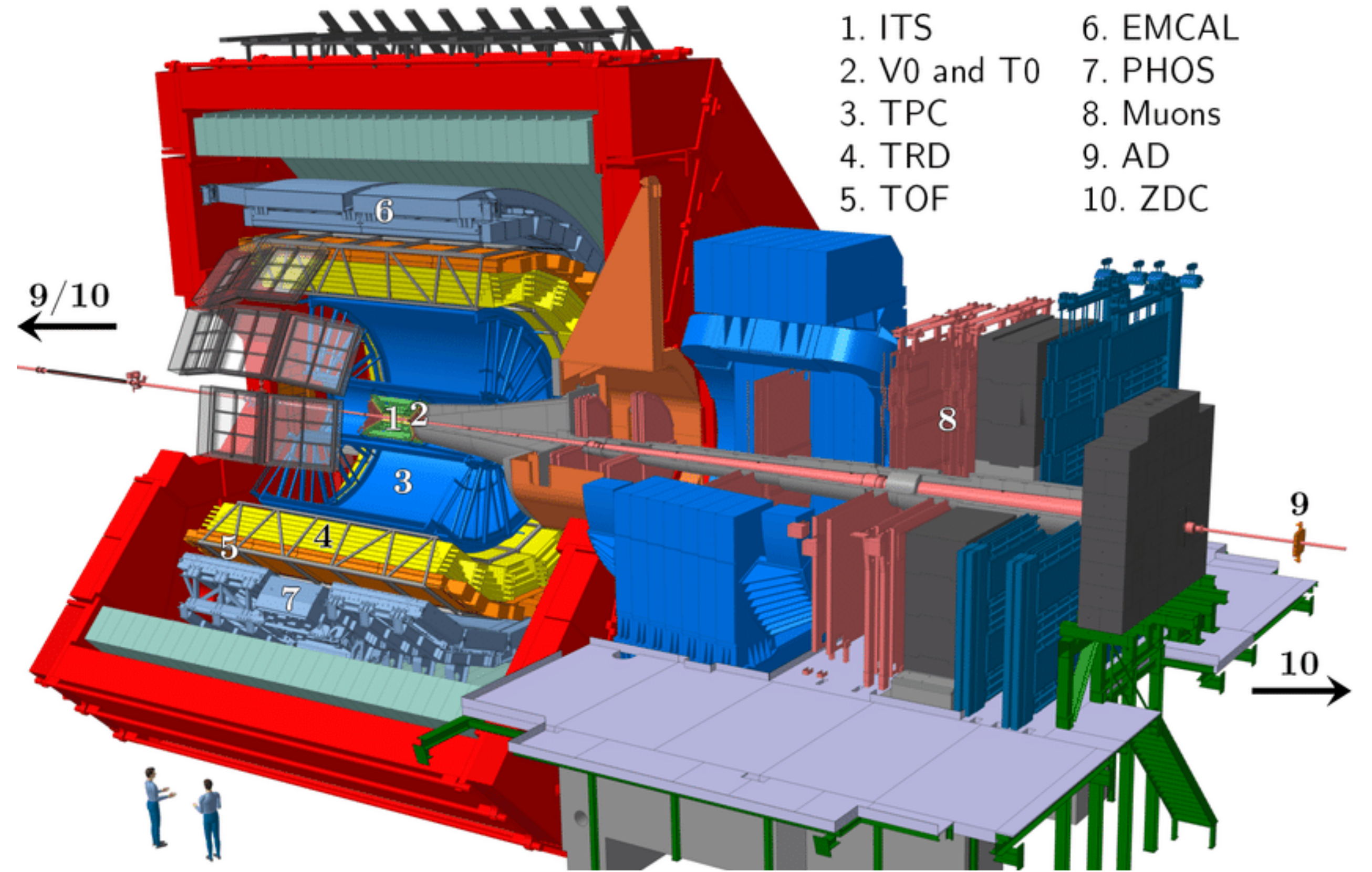
**Cross-over point for CMS somewhere else:
different magnetic field, tracker, calorimeter...**

Heavy ions + DIS

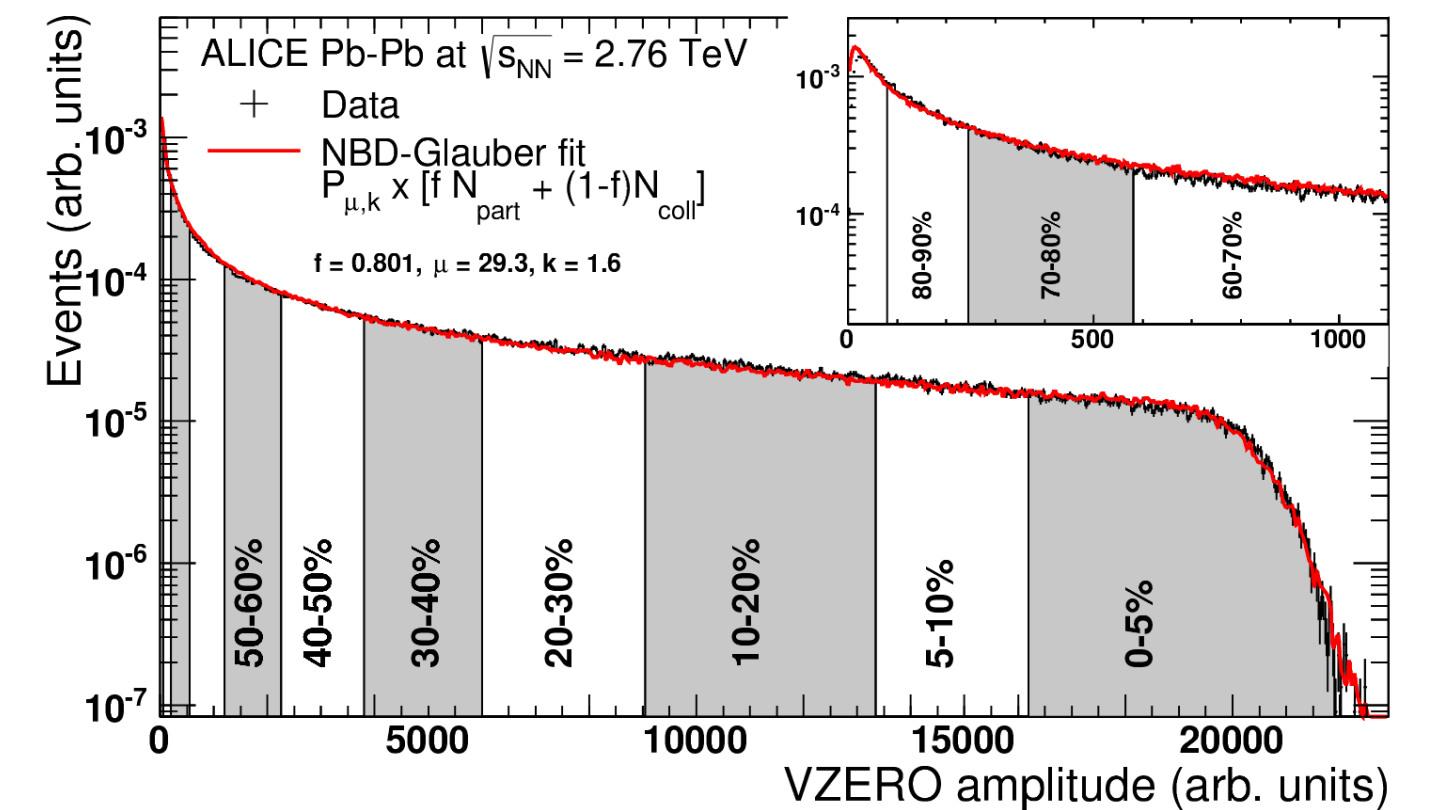
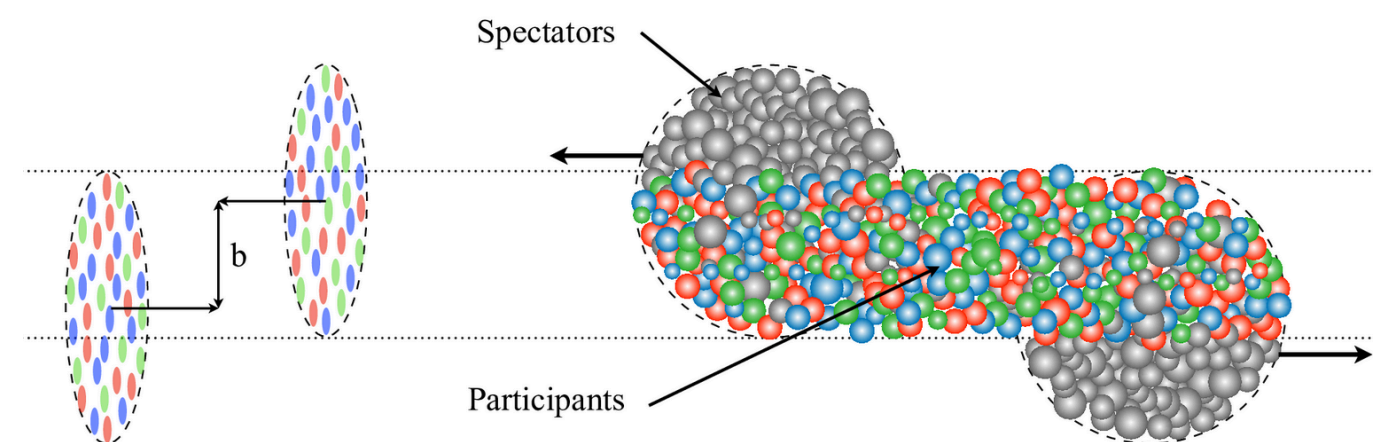
ALICE

♪ ... looks at collisions of lead ions. ♪

- Designed for heavy-ion physics studies.
 - TPC provides excellent dE/dx measurements → enhanced capability of particle ID studies
 - p/Pi/K separation @ 3σ , < 2 GeV
 - Enables D-hadron ID from TVs
 - TOF MRPC system extends PID to higher p_T 's



- **Key concept** in HI collisions: **Centrality** related to overlap region of colliding nuclei.
 - Determined from V0 or ZDC subdetectors.



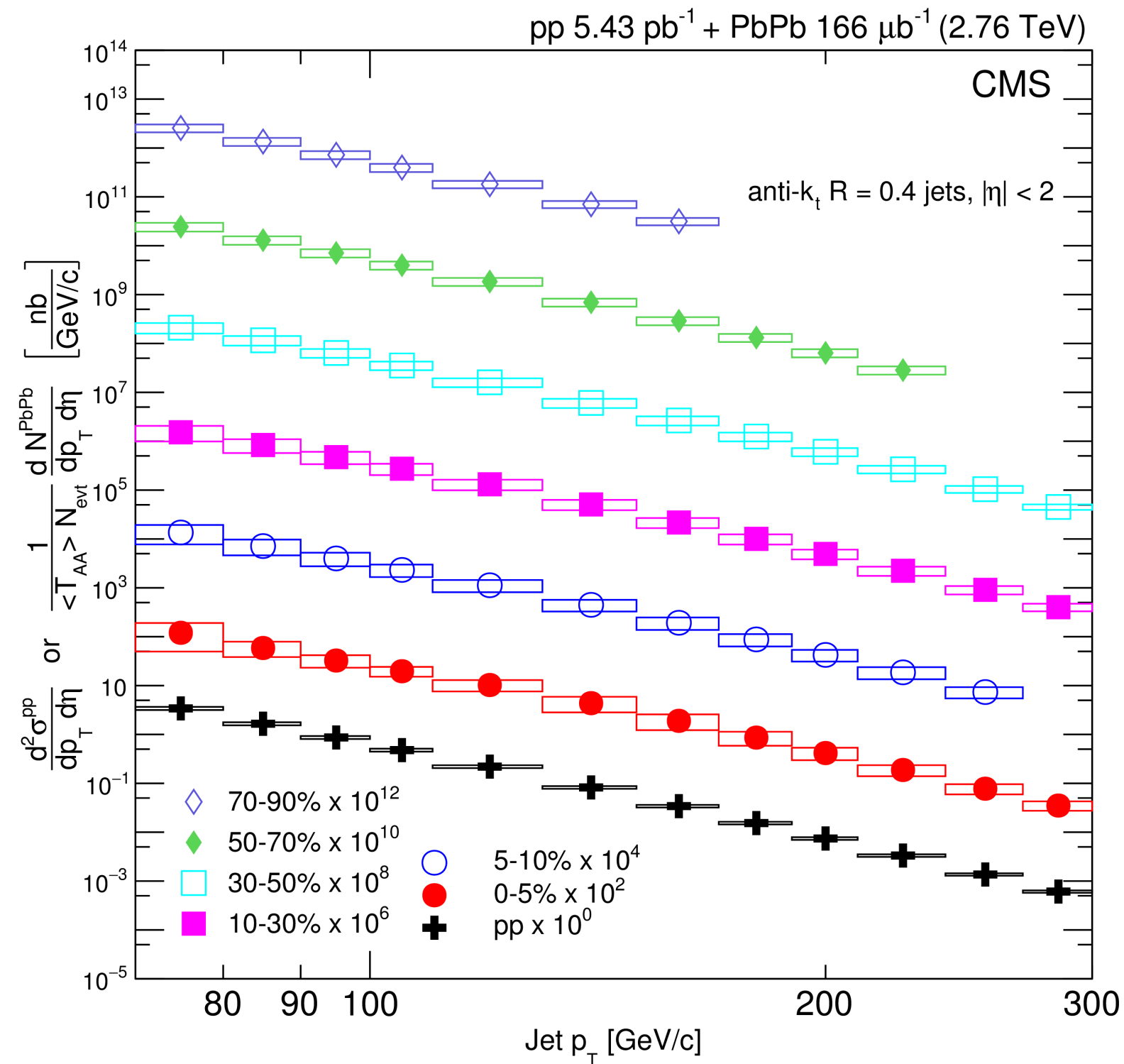
Ratios of Pb+Pb / pp (R_{AA})

Example from CMS, Phys. Rev. C 96 (2017) 015202

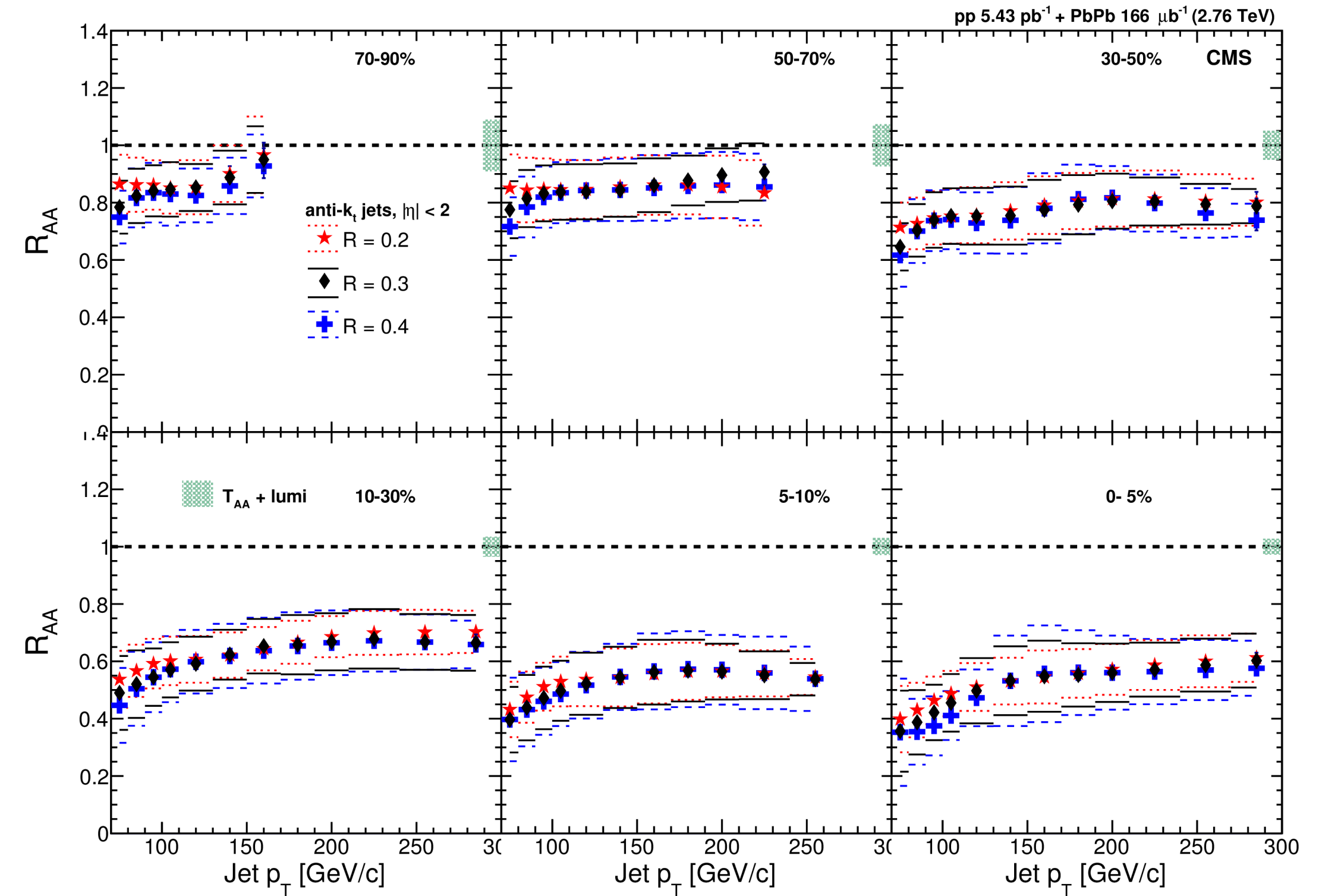
Many HI results study the relative differences of systems
 “in-medium” (w/ QGP) and “in-vacuum” (pp) → Ratios!

Pb+Pb
 More to
 less central

pp



*Least central:
 least modification*



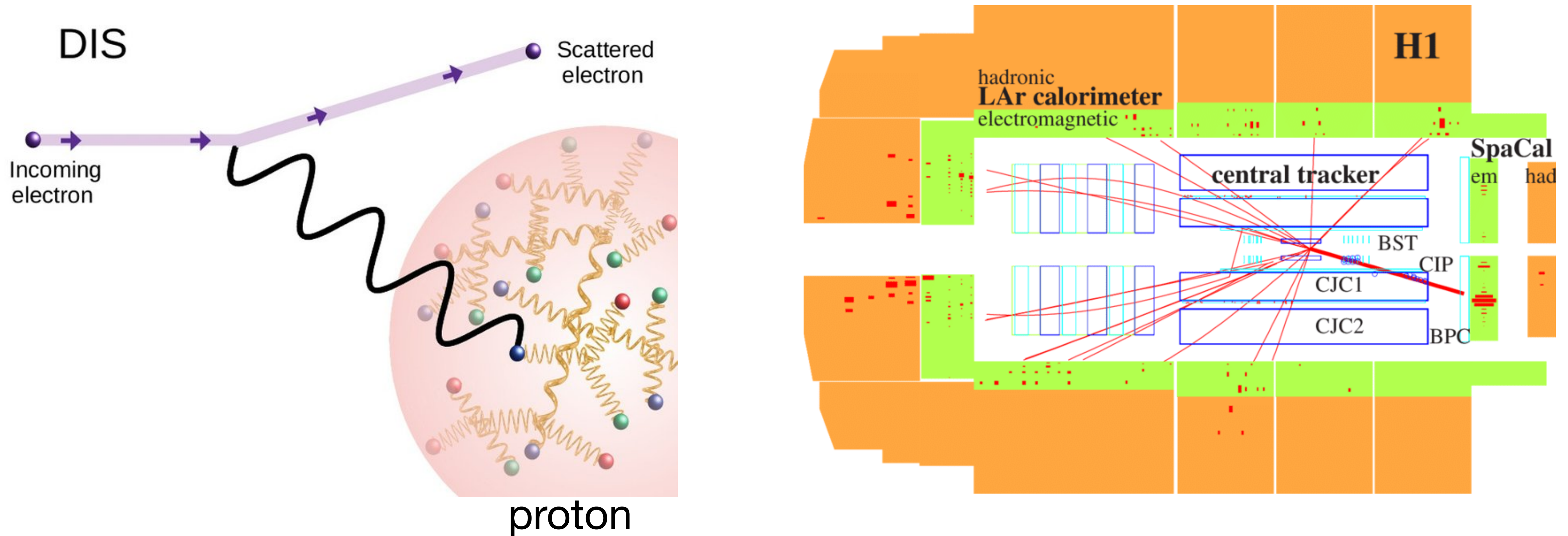
*Most central:
 most modification*

Deep Inelastic Scattering (DIS)

Special “zombie” prize for **H1**:
Bringing jet substructure back in time,
with three talks this year!

M. Światłowski @ BOOST '23

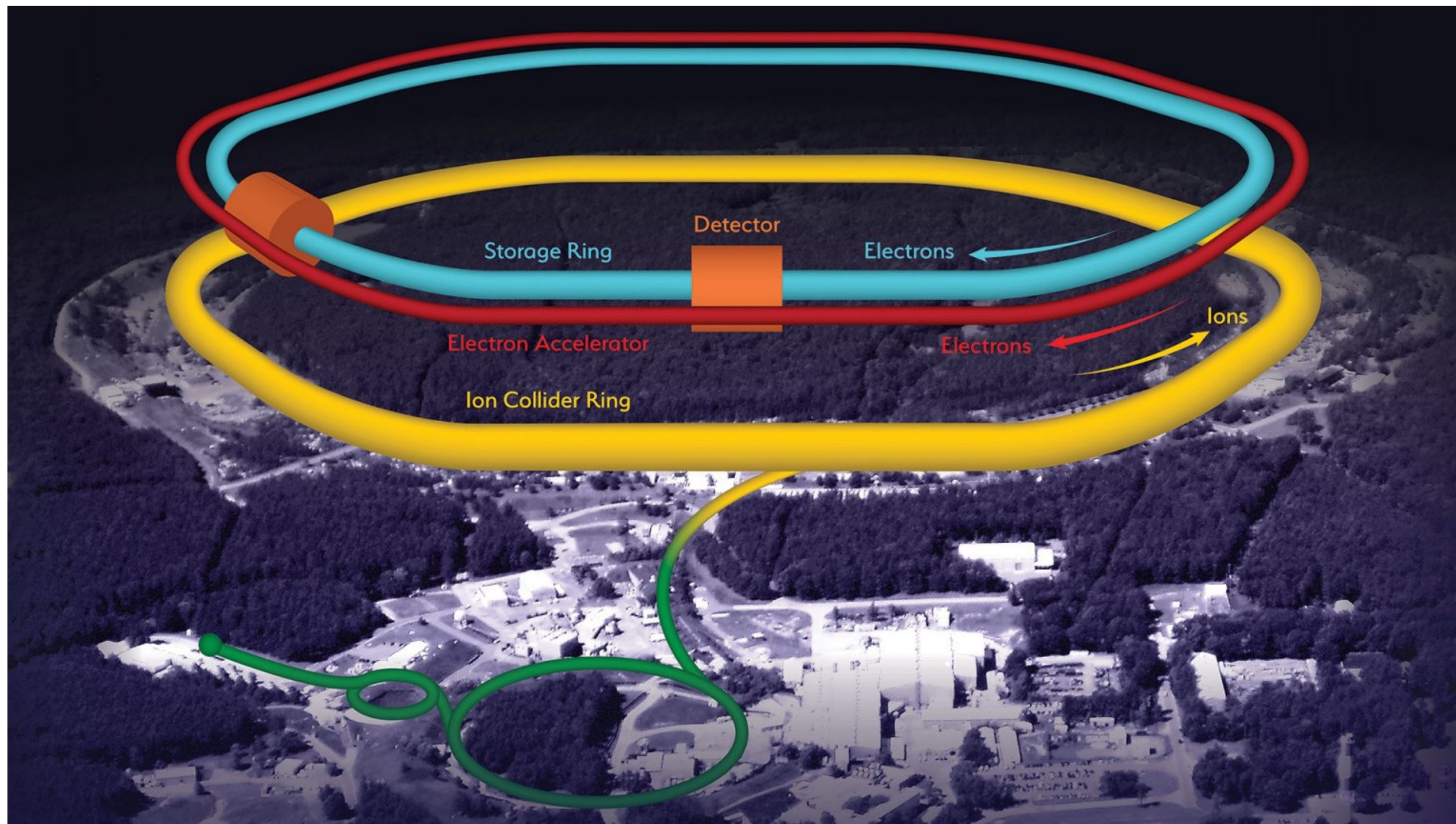
- HERA delivered $\sqrt{s}=320$ GeV electron-proton collisions to H1 and ZEUS detectors (1992-2007).
- Long history of fragmentation measurements continues: 3 new H1 results shown last year!



Electron-Ion Collider

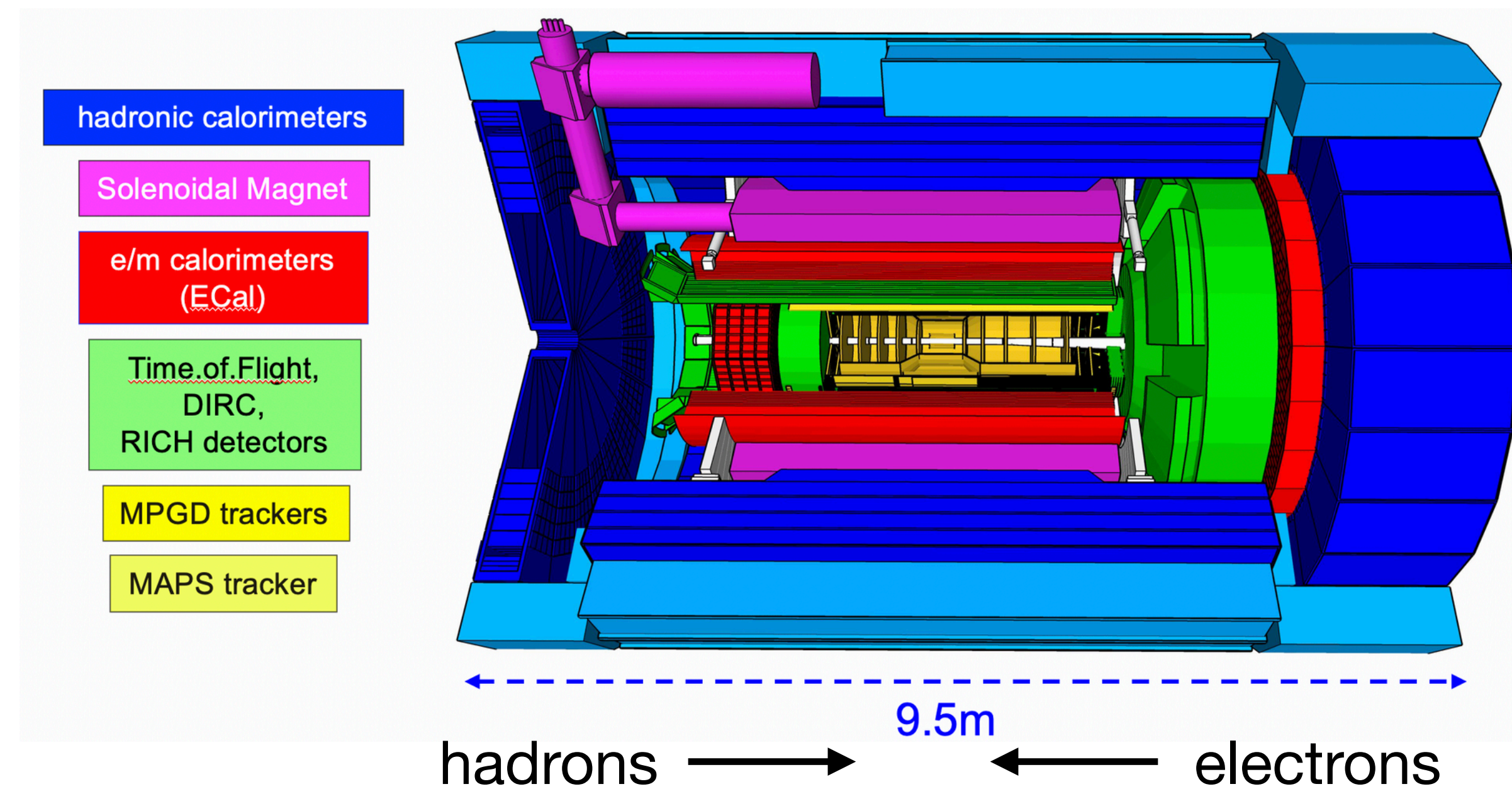
EIC Snowmass White Paper 2203.13199

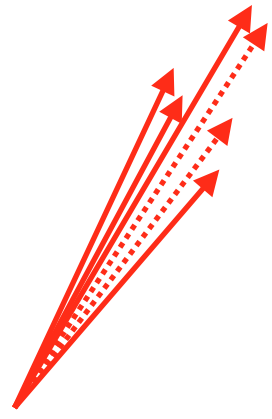
- EIC project is scheduled for construction at Brookhaven this decade.
 - e-p and e-nucleus deep inelastic scattering
 - polarized beams (spin physics studies)



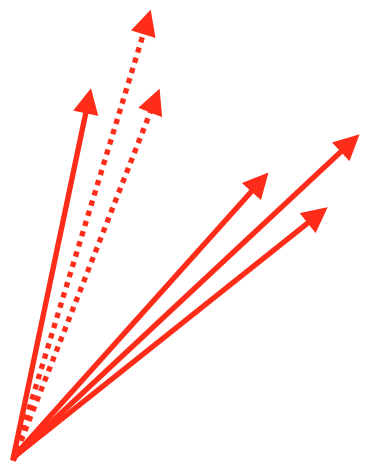
EPIC Detector Concept

- Tracker: MAPS, micro pattern gaseous detectors (MPGDs)
- ECal: Homogeneous PbWO_4
- Far-forward & far-backward roman pots, ZDC along beam line
- PID (\rightarrow p/Pi/K separation @ 3σ)
 - AC-LGAD ToF
 - Barrel: DIRC, <6 GeV
 - Backward: Aerogel RICH, <10 GeV
 - Forward: Gaseous RICH, <50 GeV



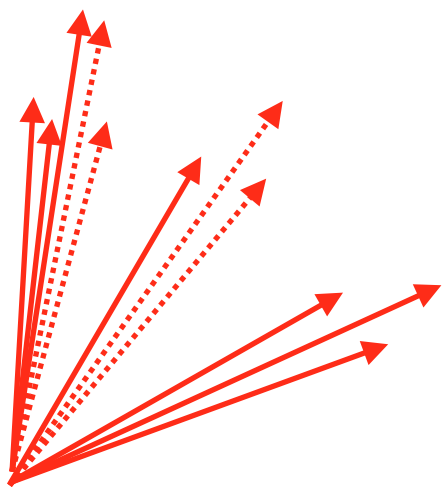


Quark? Gluon?



Higgs boson?

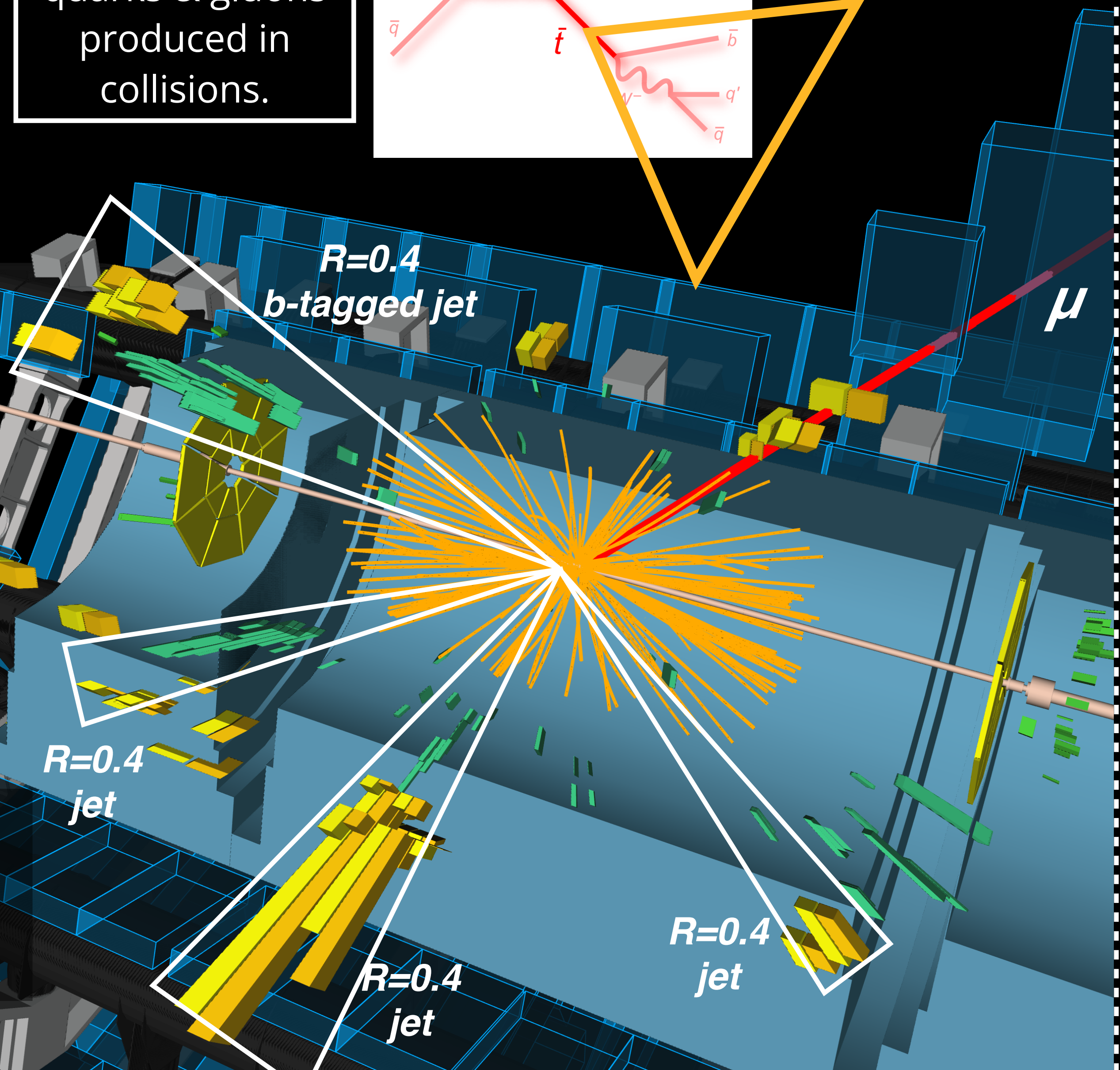
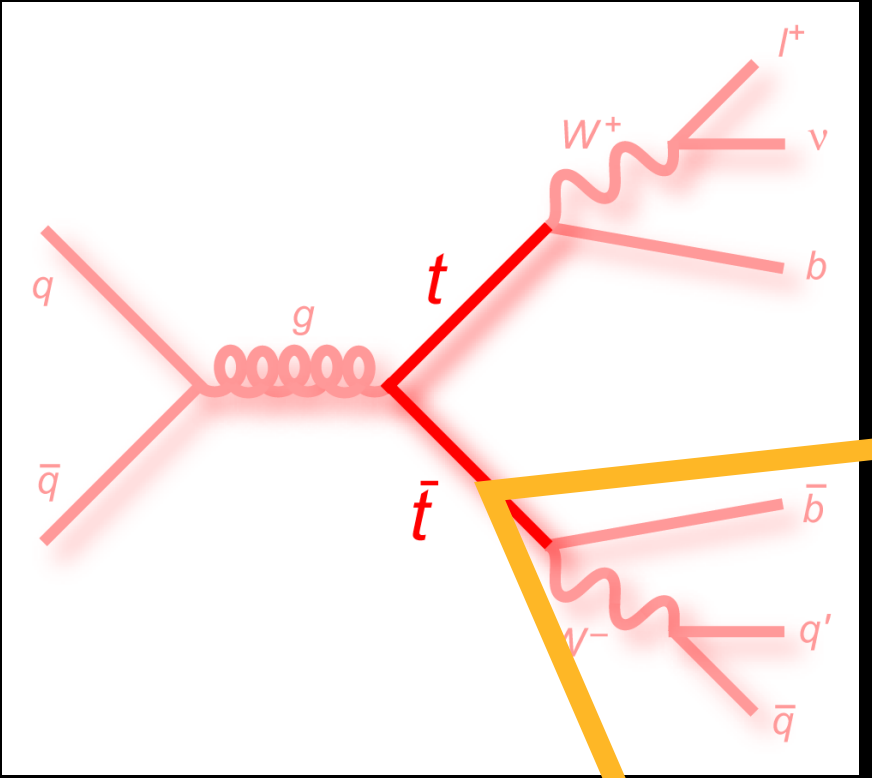
W/Z boson?



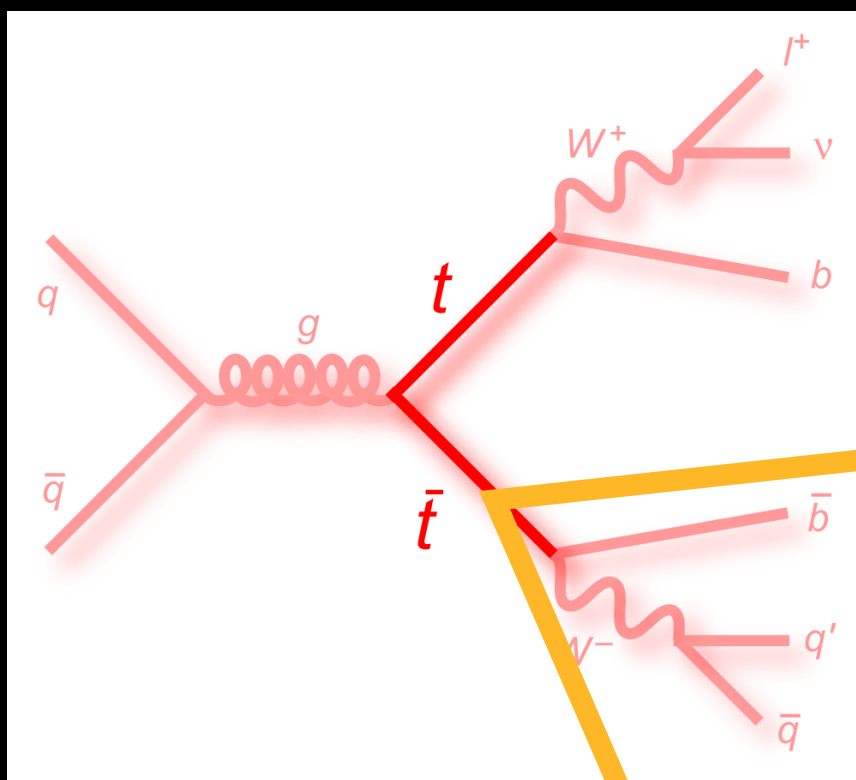
Top quark?

Tagging / Classification

Jets are proxies for high-energy quarks & gluons produced in collisions.

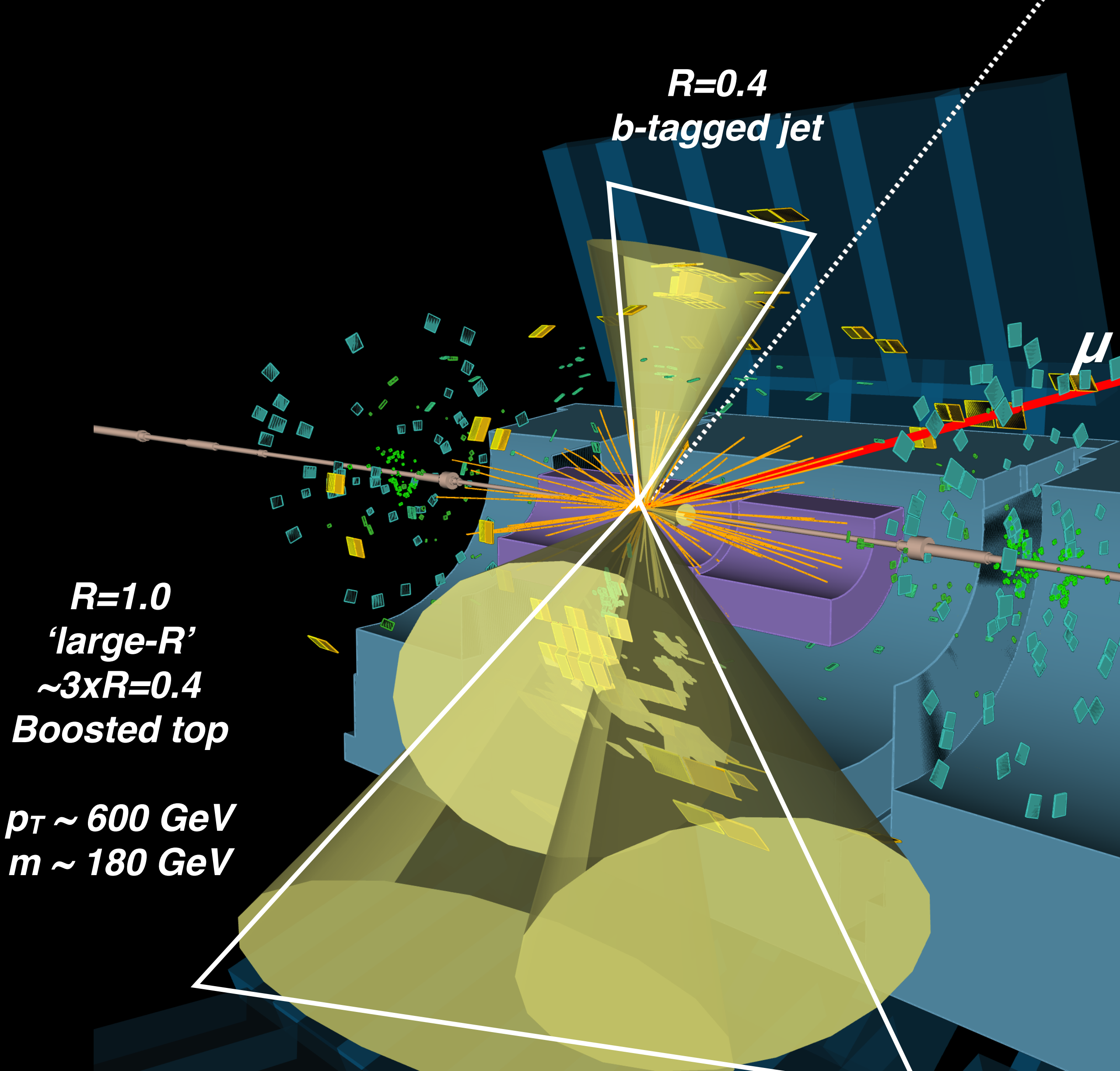
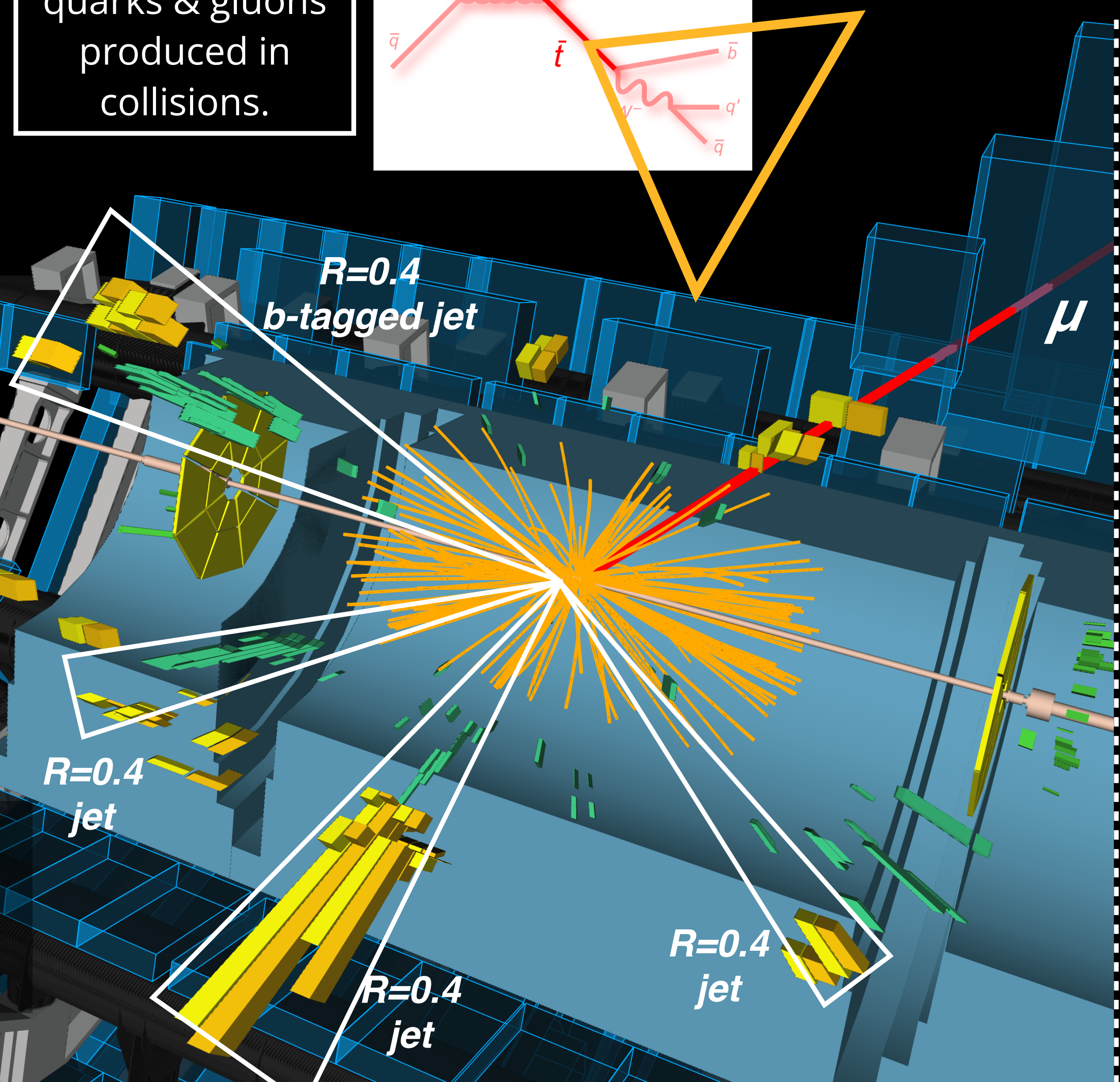


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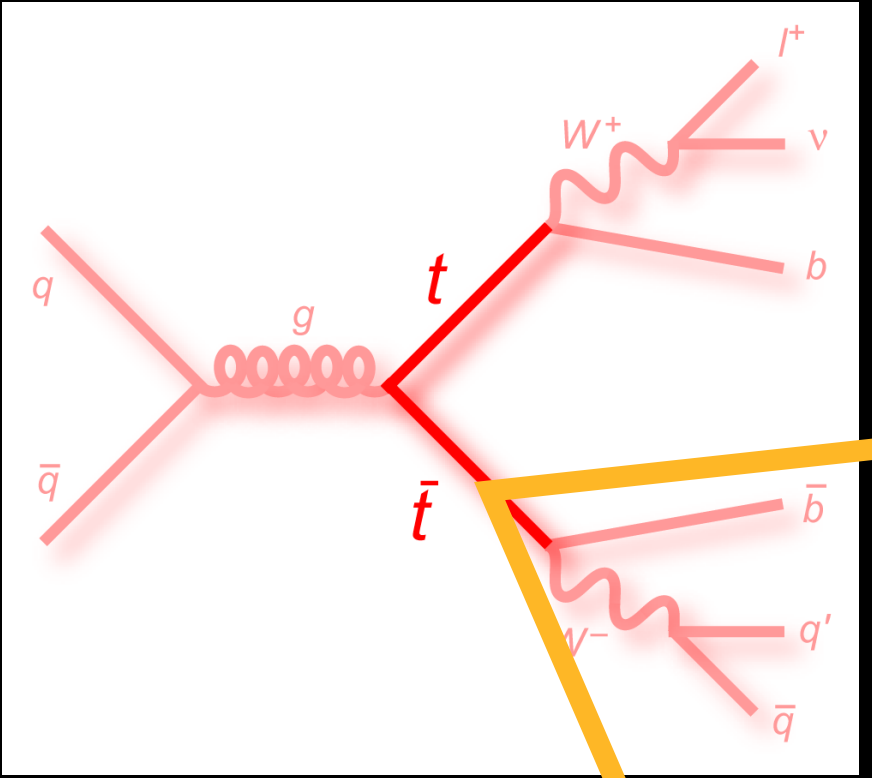


BOOST!

... use "large-R" jets to collect the hadronic decays of heavy particles like top quarks!

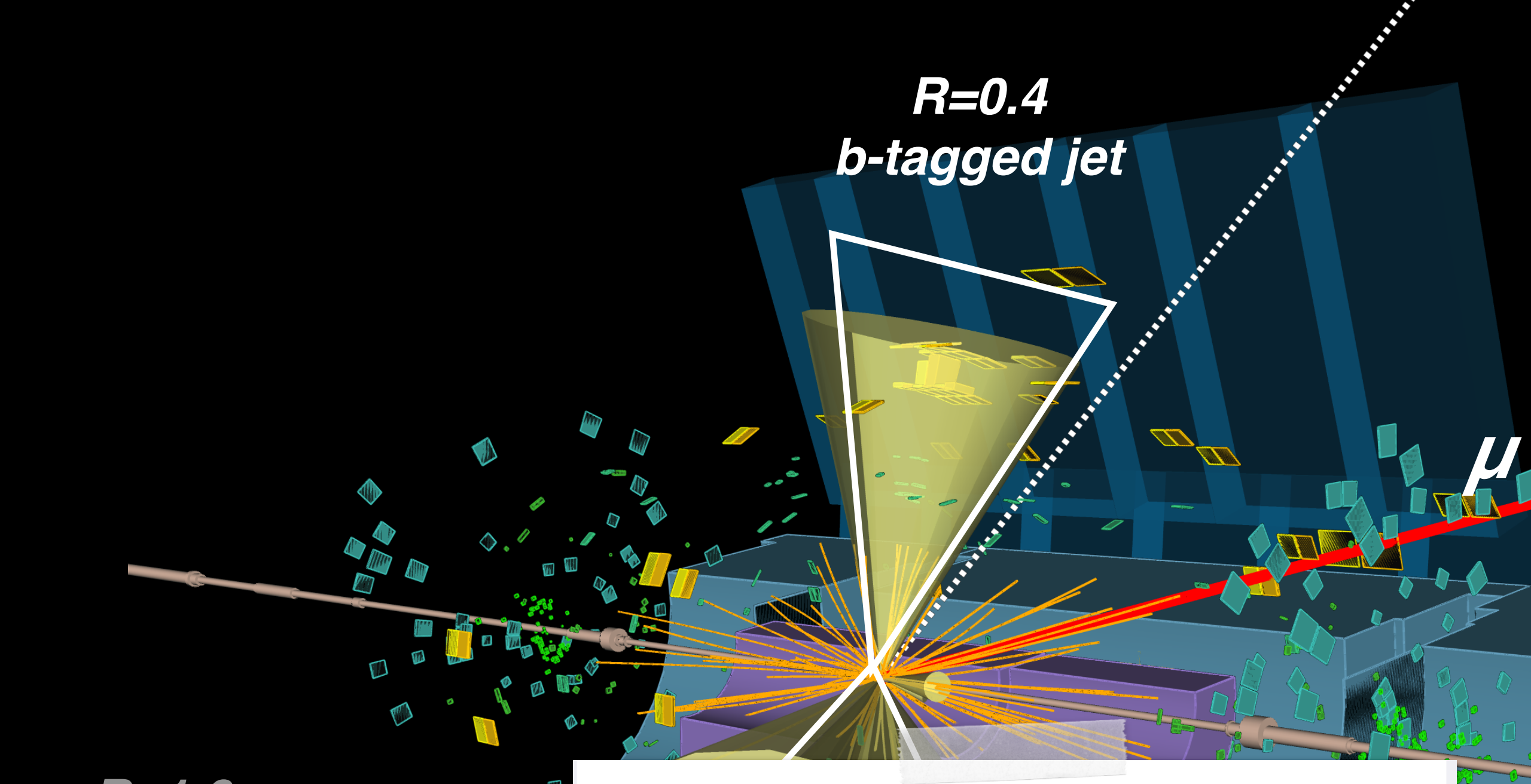
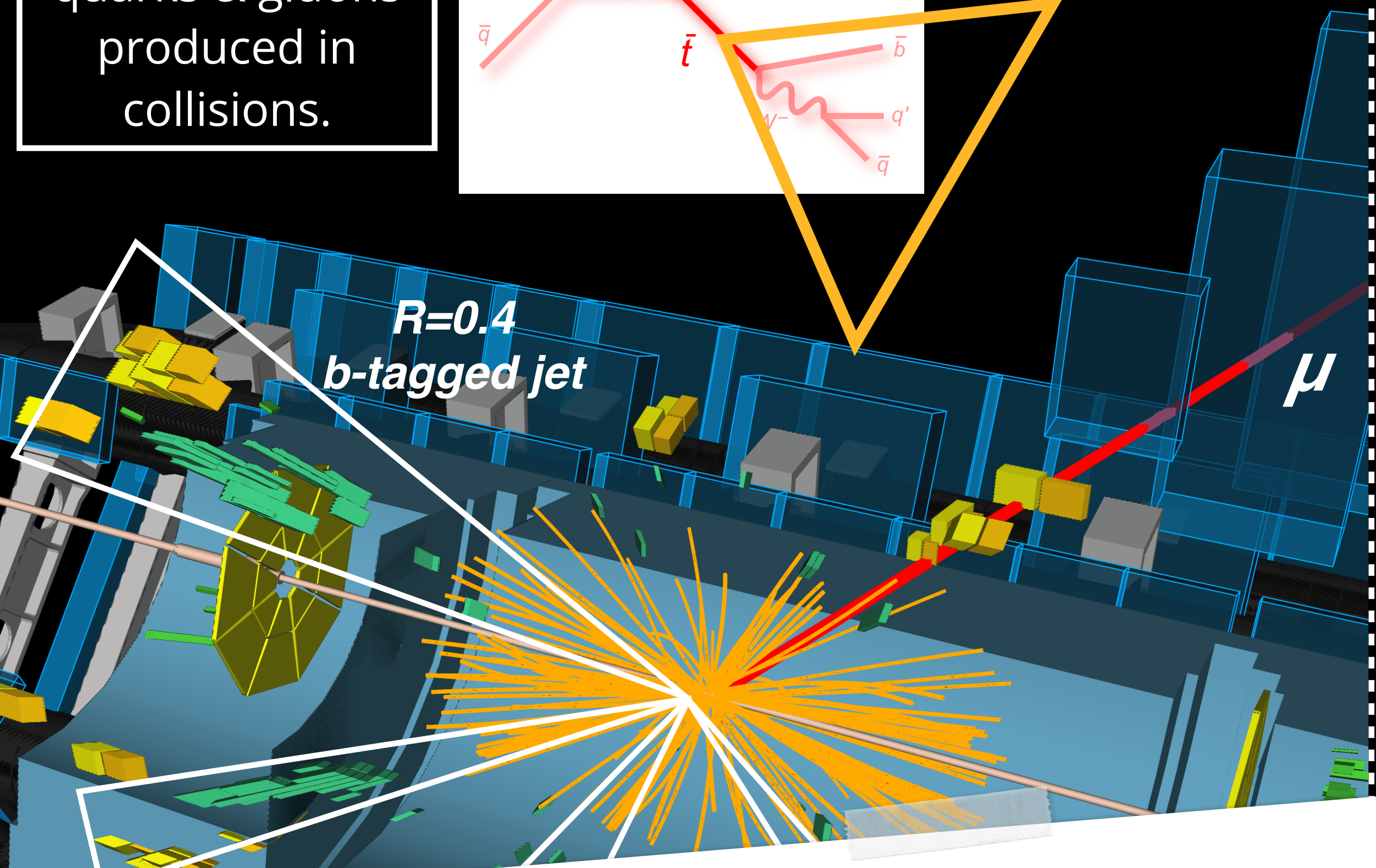


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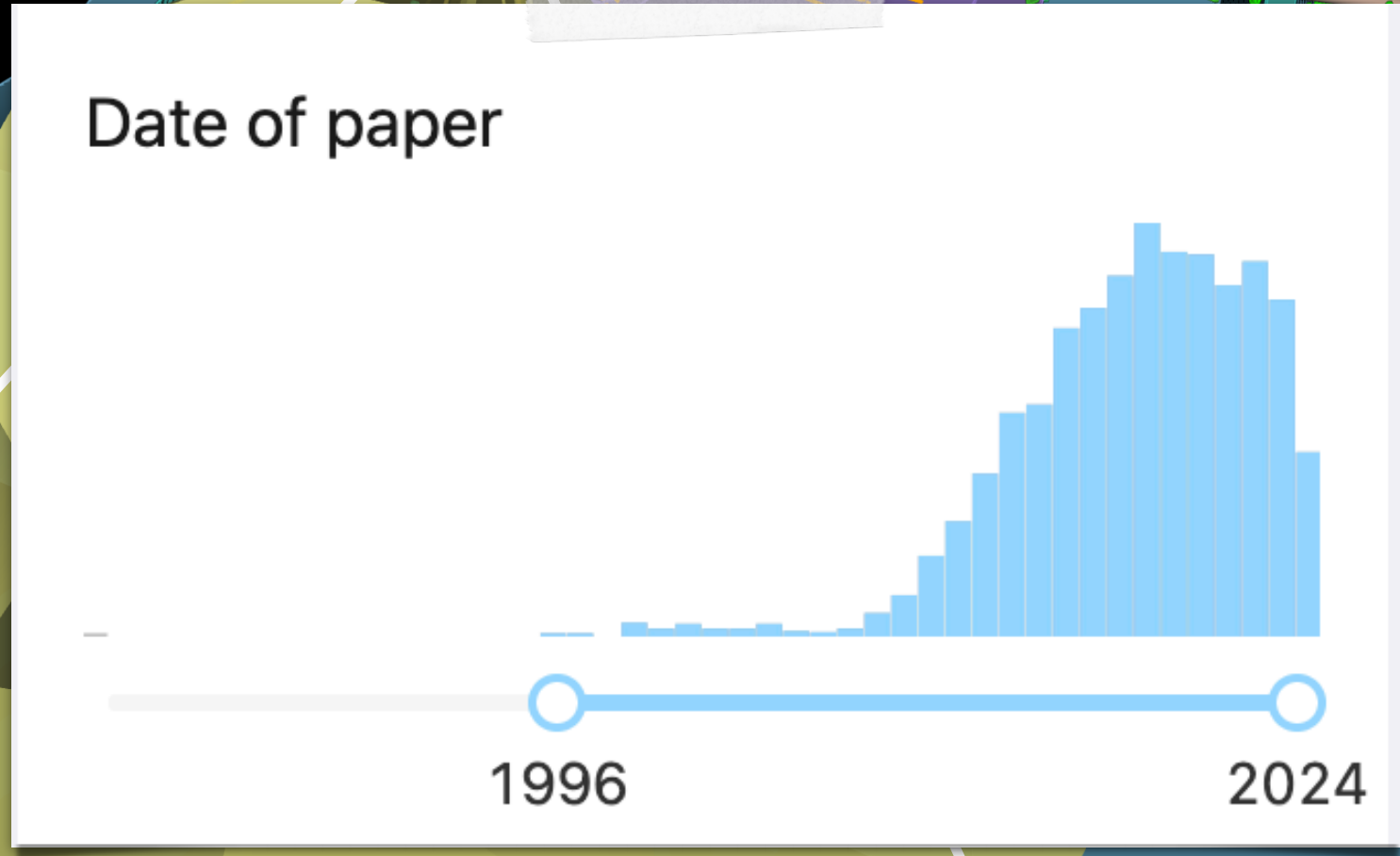


BOOST!

... use "large-R" jets to collect the hadronic decays of heavy particles like top quarks!

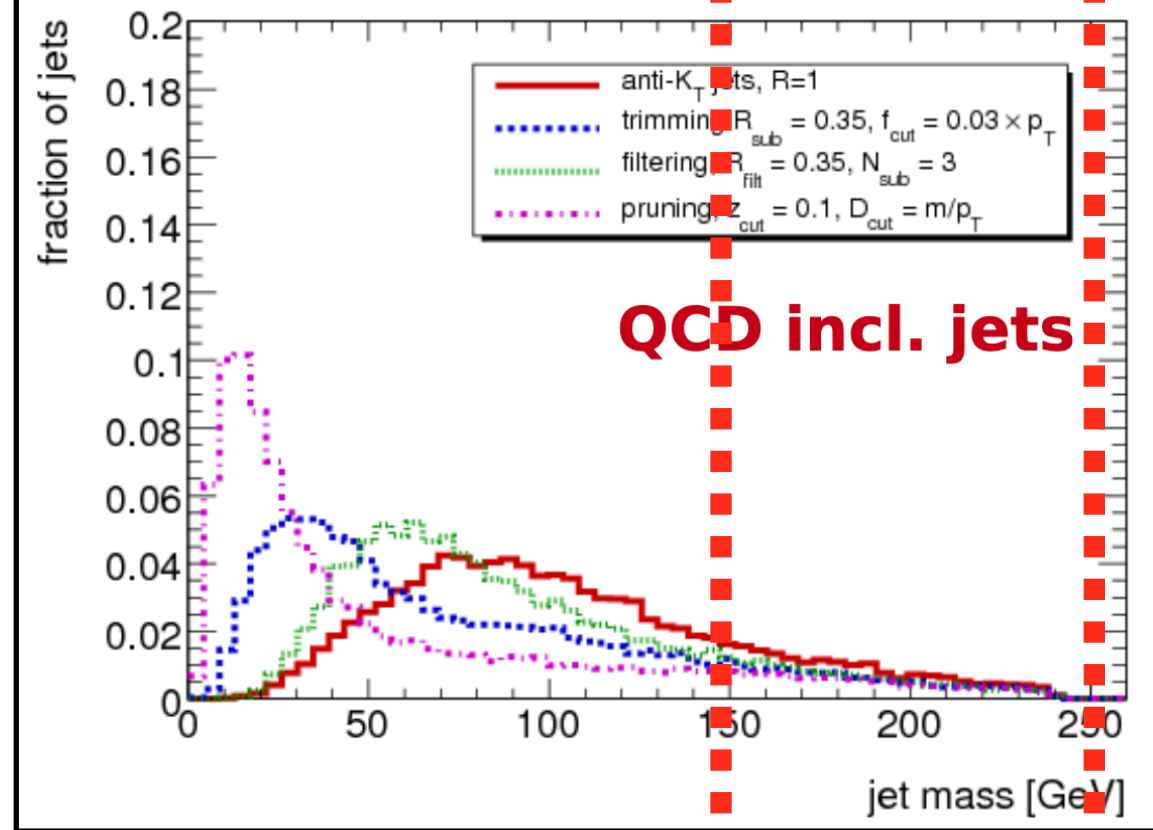
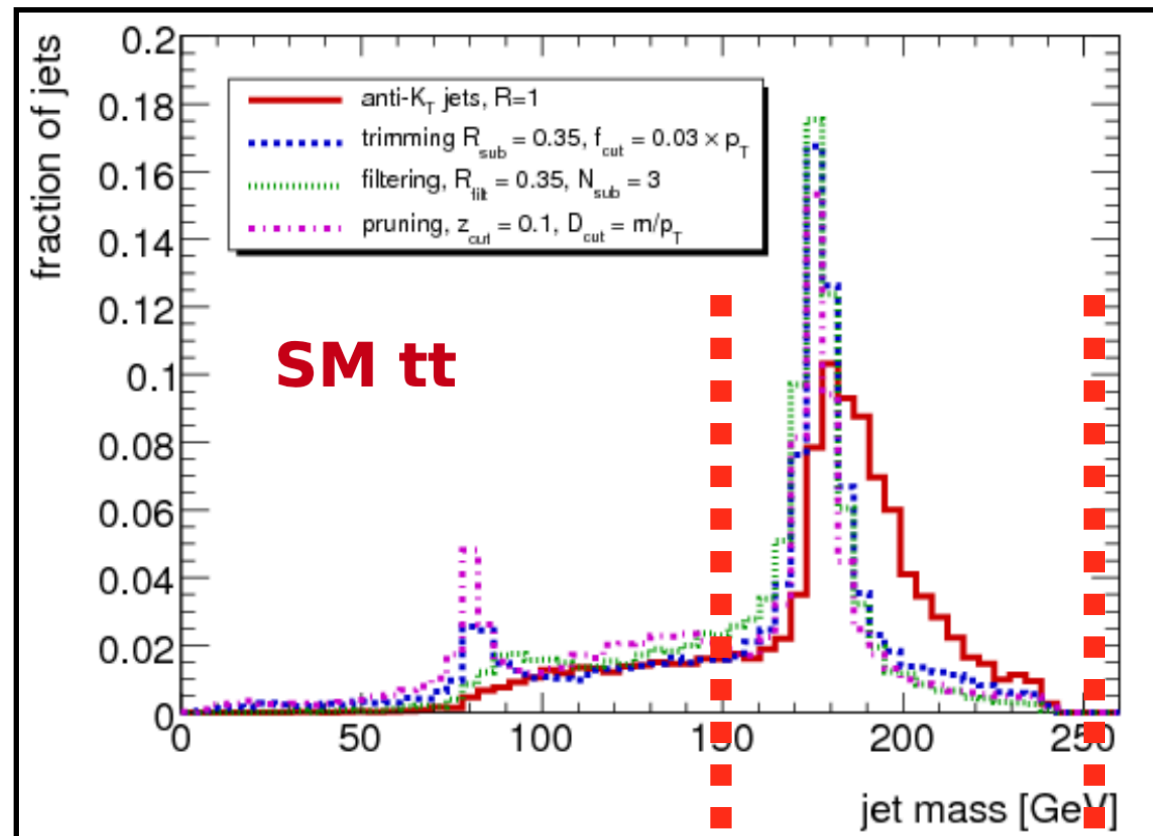


LHC is the first collider with $\sqrt{s} \gg$ EWSB scale!
 JSS techniques were explored earlier (Seymour '94, BDRS '08),
 now a **characteristic aspect of the LHC physics program!**

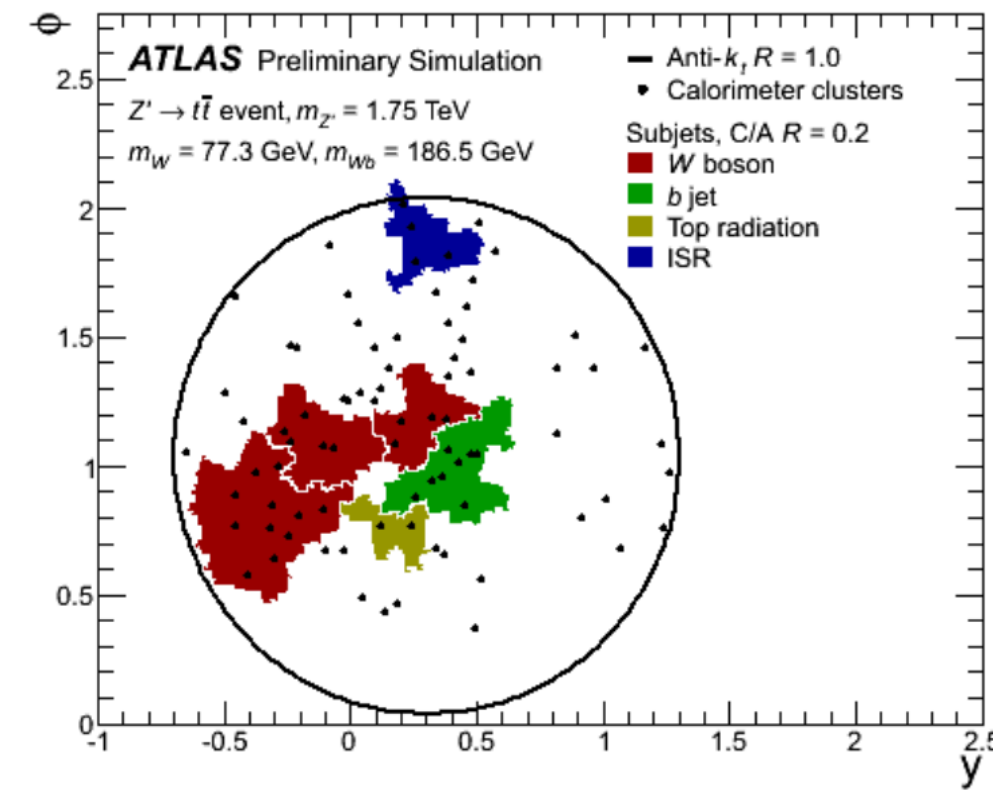
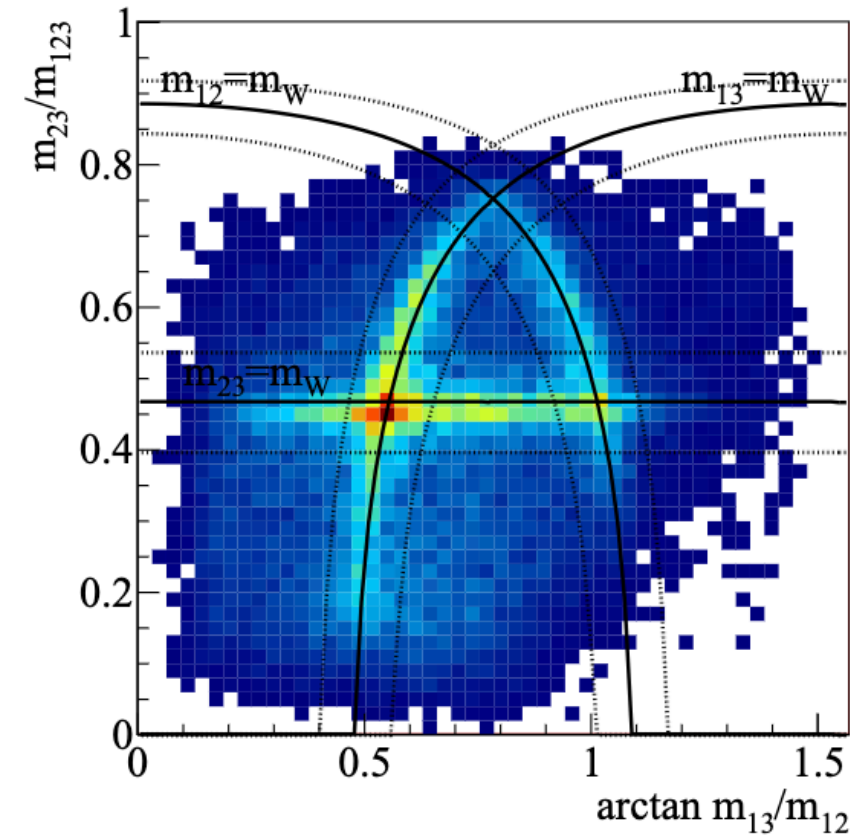


Then

M. Vos o.b.o. ATLAS+CMS, 2014



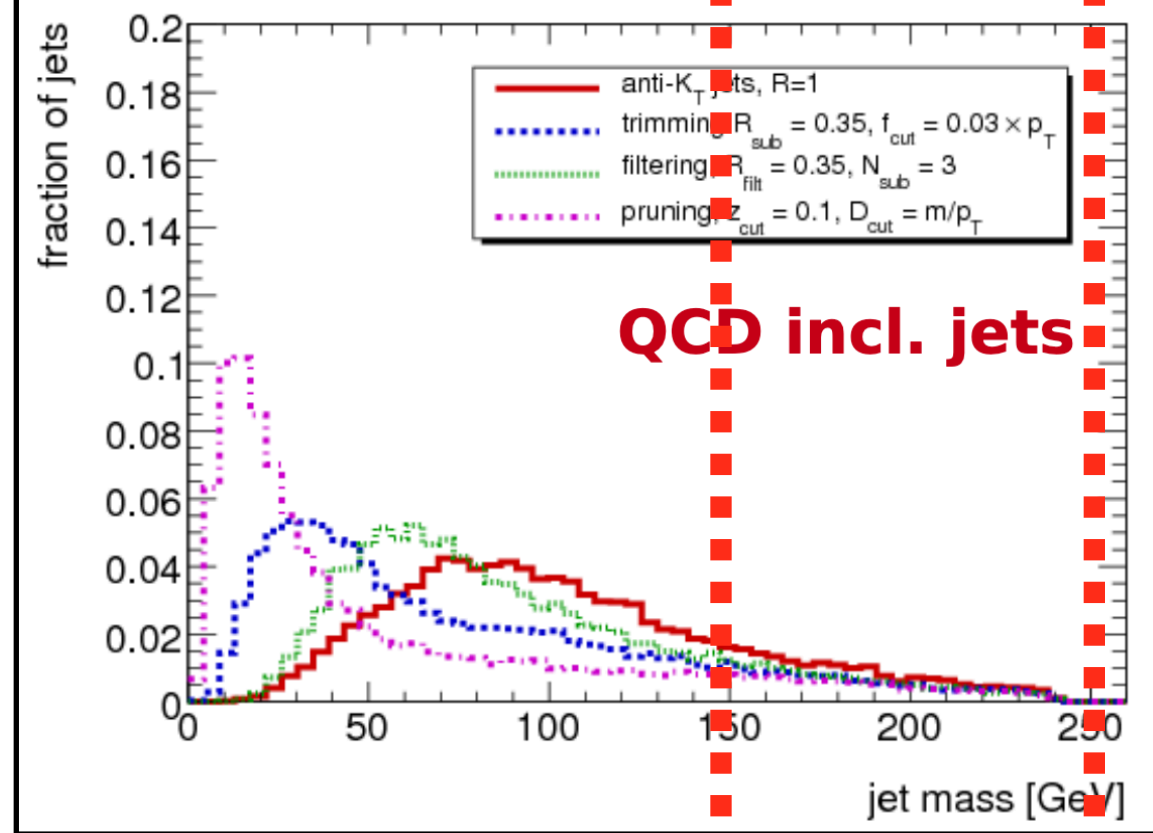
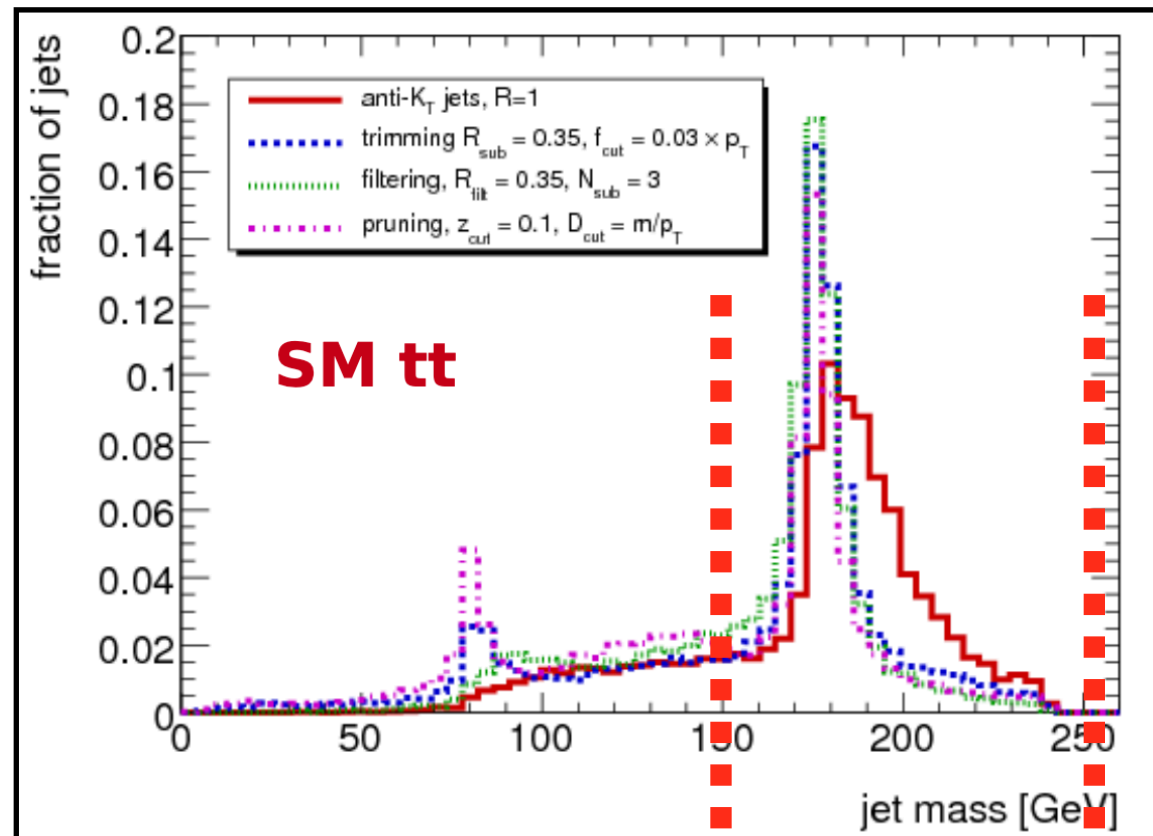
Cut on individual
JSS observables



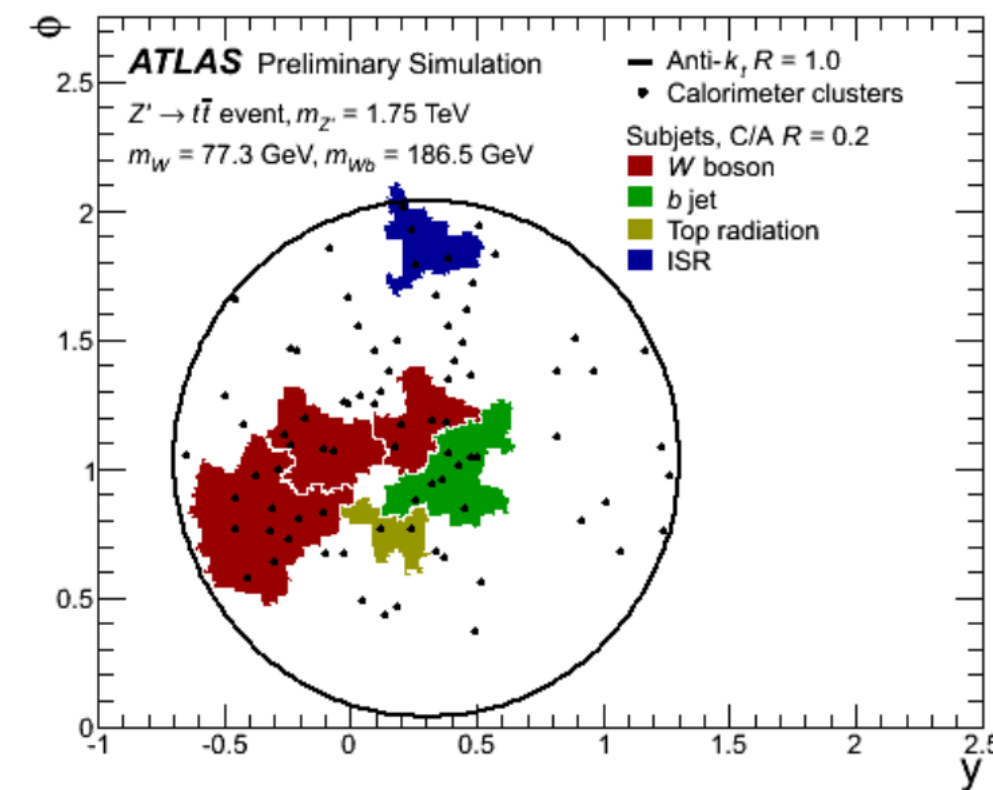
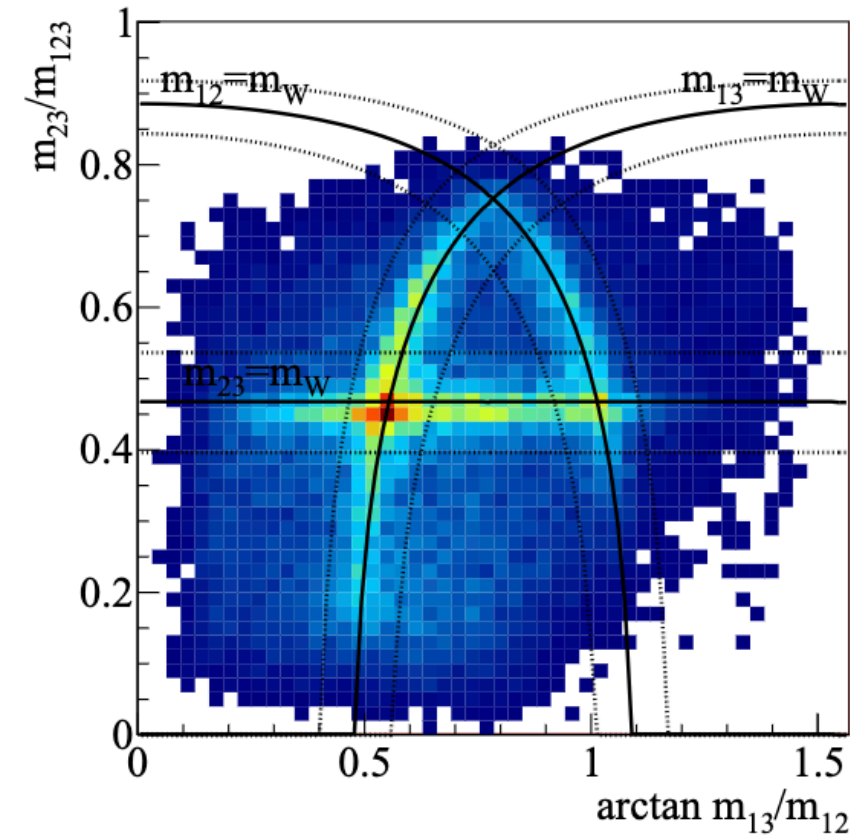
Rule-based
HepTopTagger,
Shower Deconstruction

Then

M. Vos o.b.o. ATLAS+CMS, 2014



Cut on individual JSS observables



Rule-based
HepTopTagger,
Shower Deconstruction

Now

CERN Accelerating science Sign in Directory

ATLAS EXPERIMENT Collaboration Site | Physics Results

Updates > Briefing >
Machine learning is revolutionising our understanding of particle "jets"

Physics Briefing

Tags: BOOST 2023, machine learning

Machine learning is revolutionising our understanding of particle "jets"

3 August 2023 | By ATLAS Collaboration

What happens when – instead of recording a single particle track or energy deposit in your detector – you see a complex collection of many particles, with many tracks, that leaves a large amount of energy in your calorimeters? Then congratulations: you've recorded a "jet"!

Jets are the complicated experimental signatures left behind by showers of strongly-interacting quarks and gluons. By studying the internal energy flow of a jet – also known as the "jet substructure" – physicists can

Feedback

(a) Particle Transformer

(b) Particle Attention Block

(c) Class Attention Block

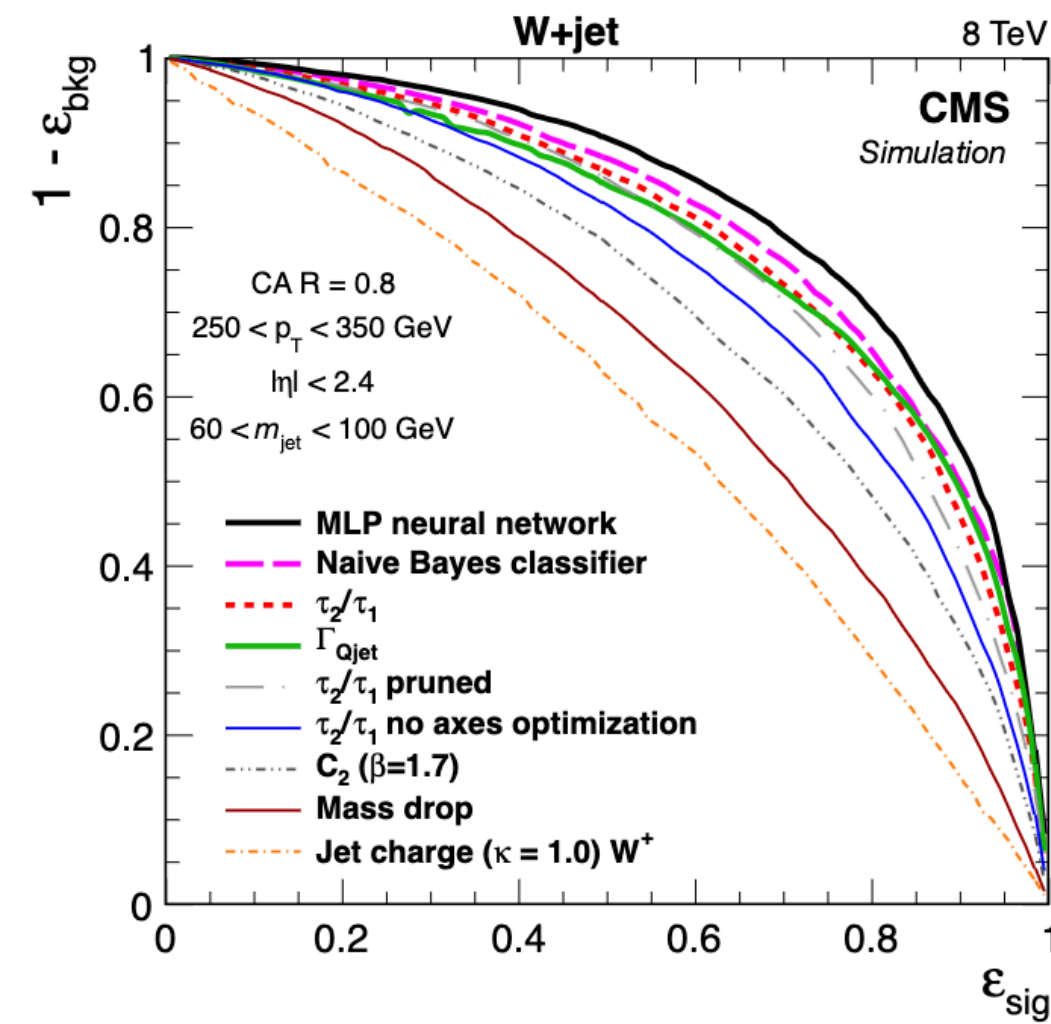
Figure 3. The architecture of (a) Particle Transformer (b) Particle Attention Block (c) Class Attention Block.

Boosted object tagging (nostalgic retrospective)

- Many different approaches are constantly being utilized in physics analysis: continuous development in this area for 16 BOOSTs (and counting!)

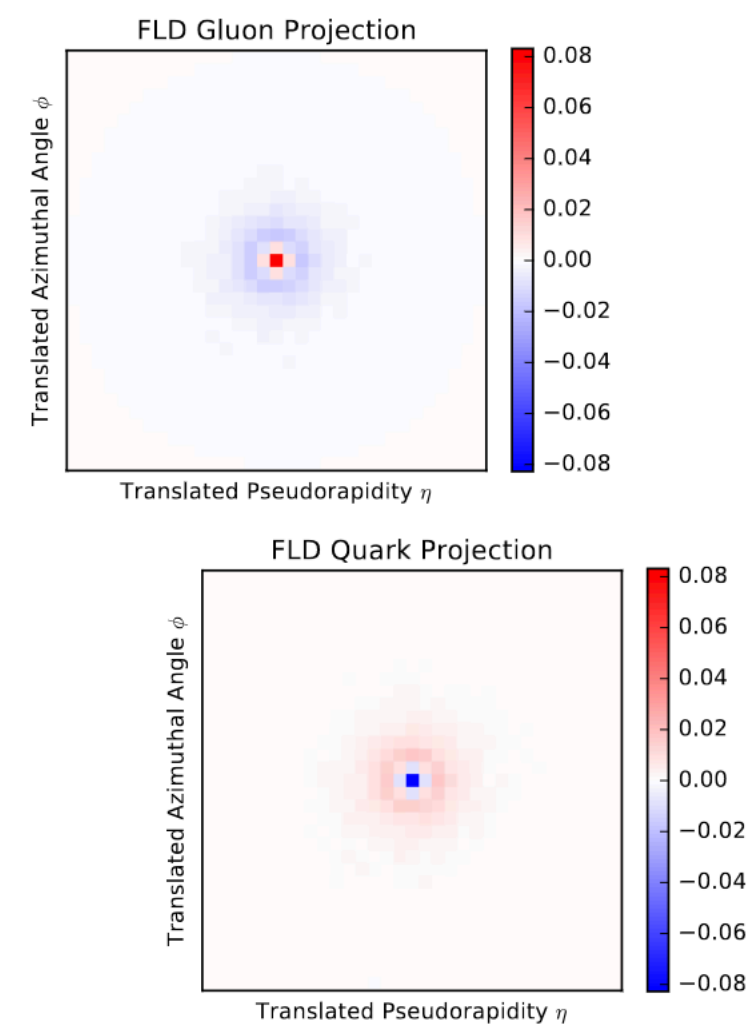
Expert features

e.g. CMS, 1410.4227



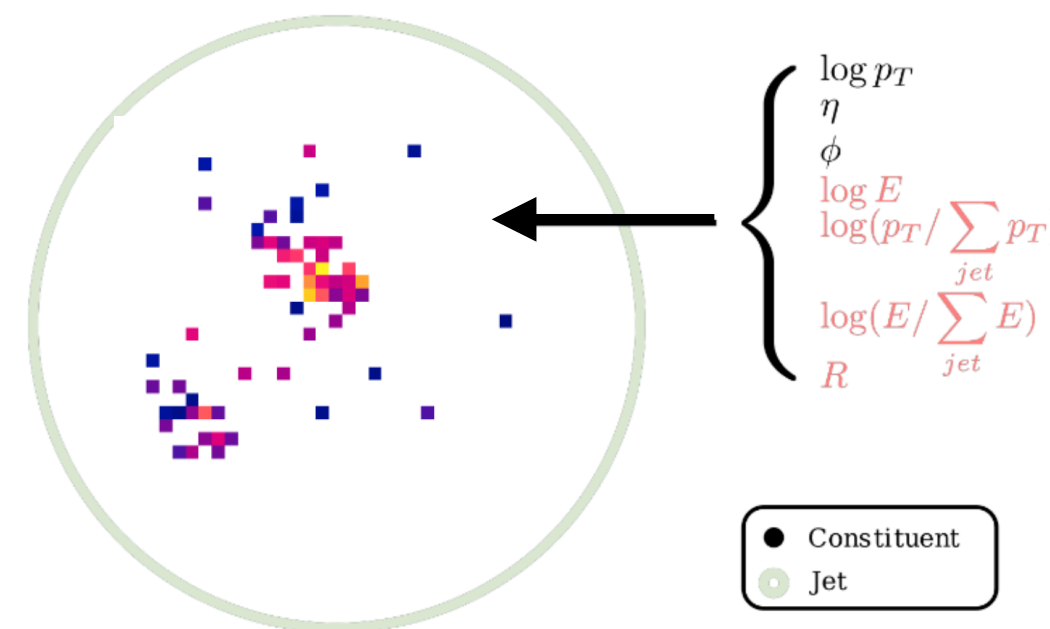
Images

e.g. Komiske et al. 1612.01551



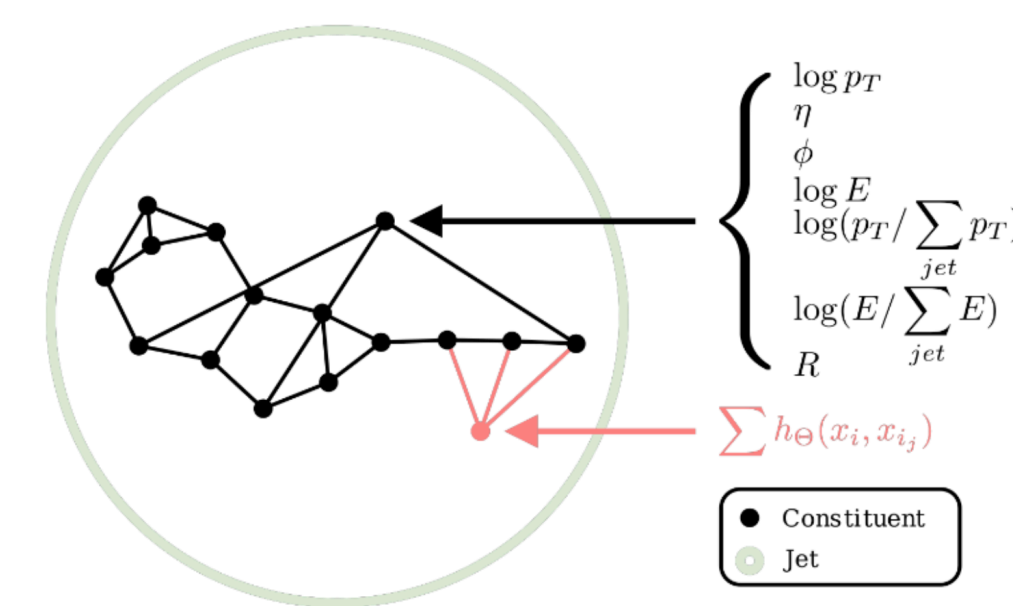
Point Cloud (PFN/EFN)

e.g. Komiske et al. 1810.05165



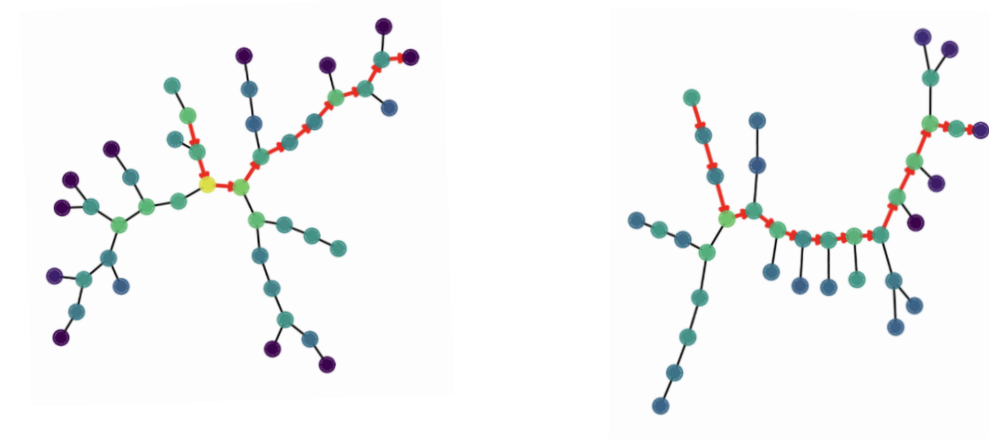
Graph (ParticleNet)

e.g. Gouskos & Qu 1902.08570



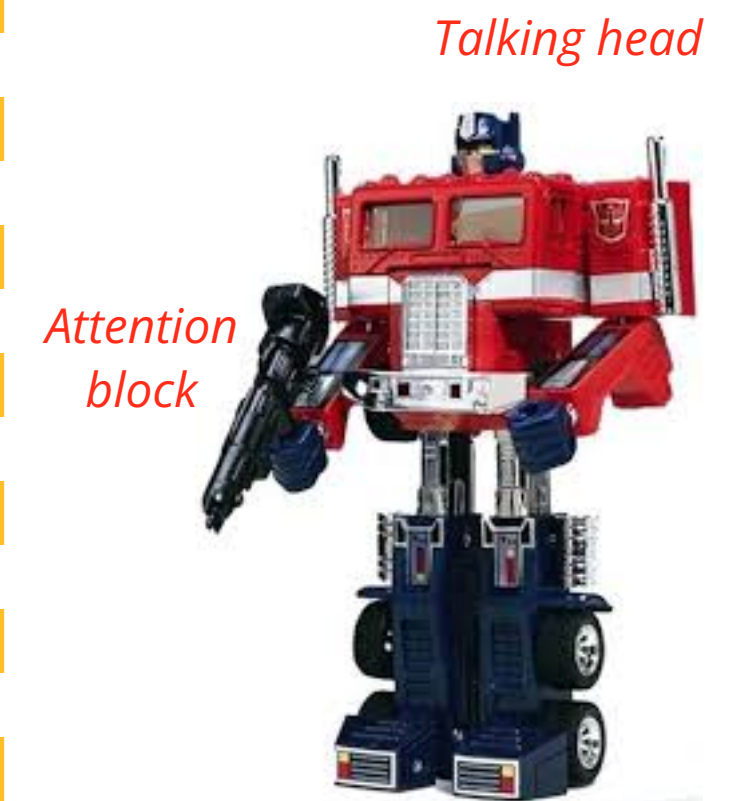
Graph definitions not unique — c.f. ATLAS LundNet tagger (C/A Graphics)

ATL-PHYS-PUB-2023-017



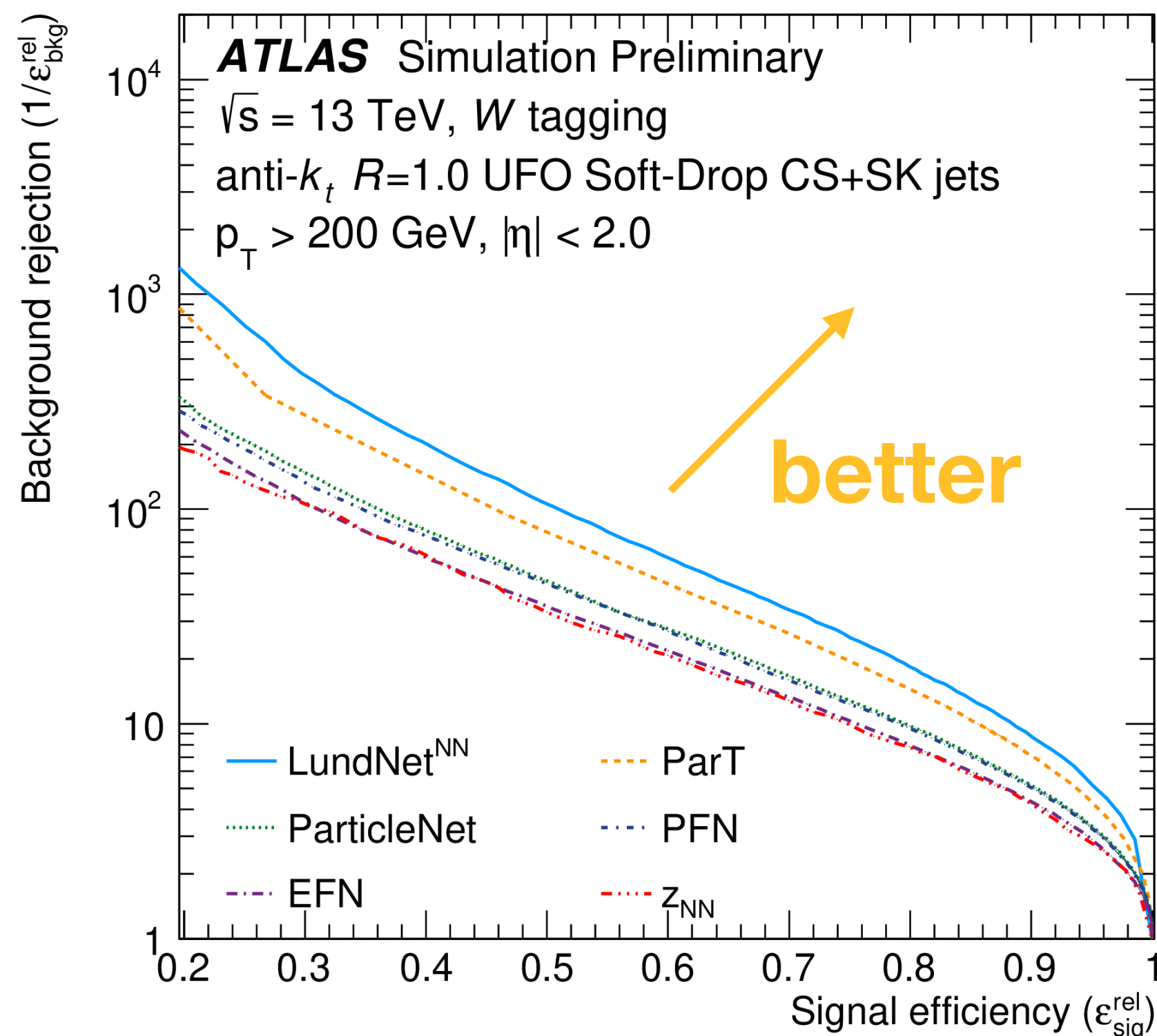
Transformers

e.g. Qu, Li & Quian 2202.03772

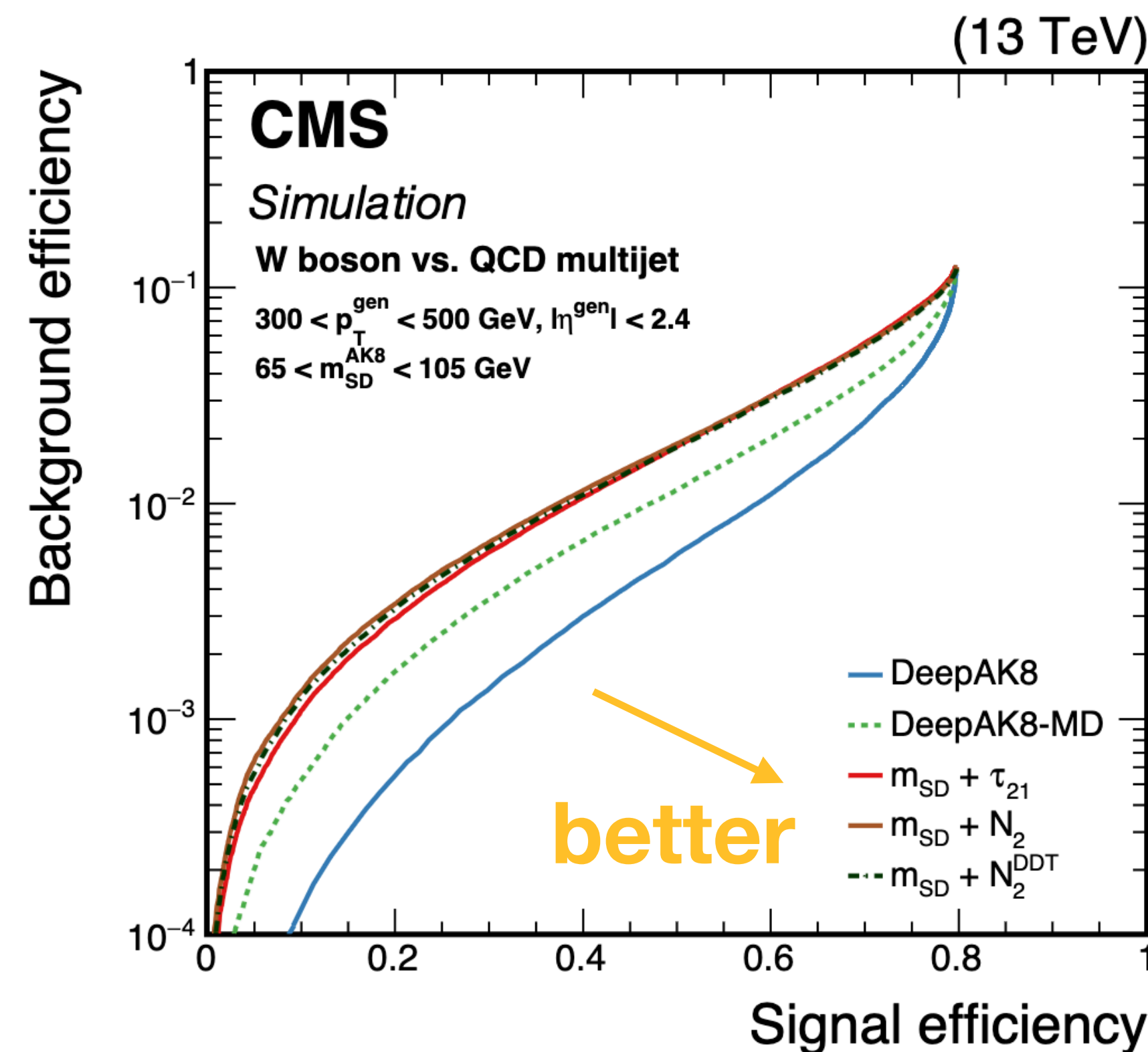


How to read a ROC curve

- Different people draw these with different ordinate quantities (efficiency vs. rejection)

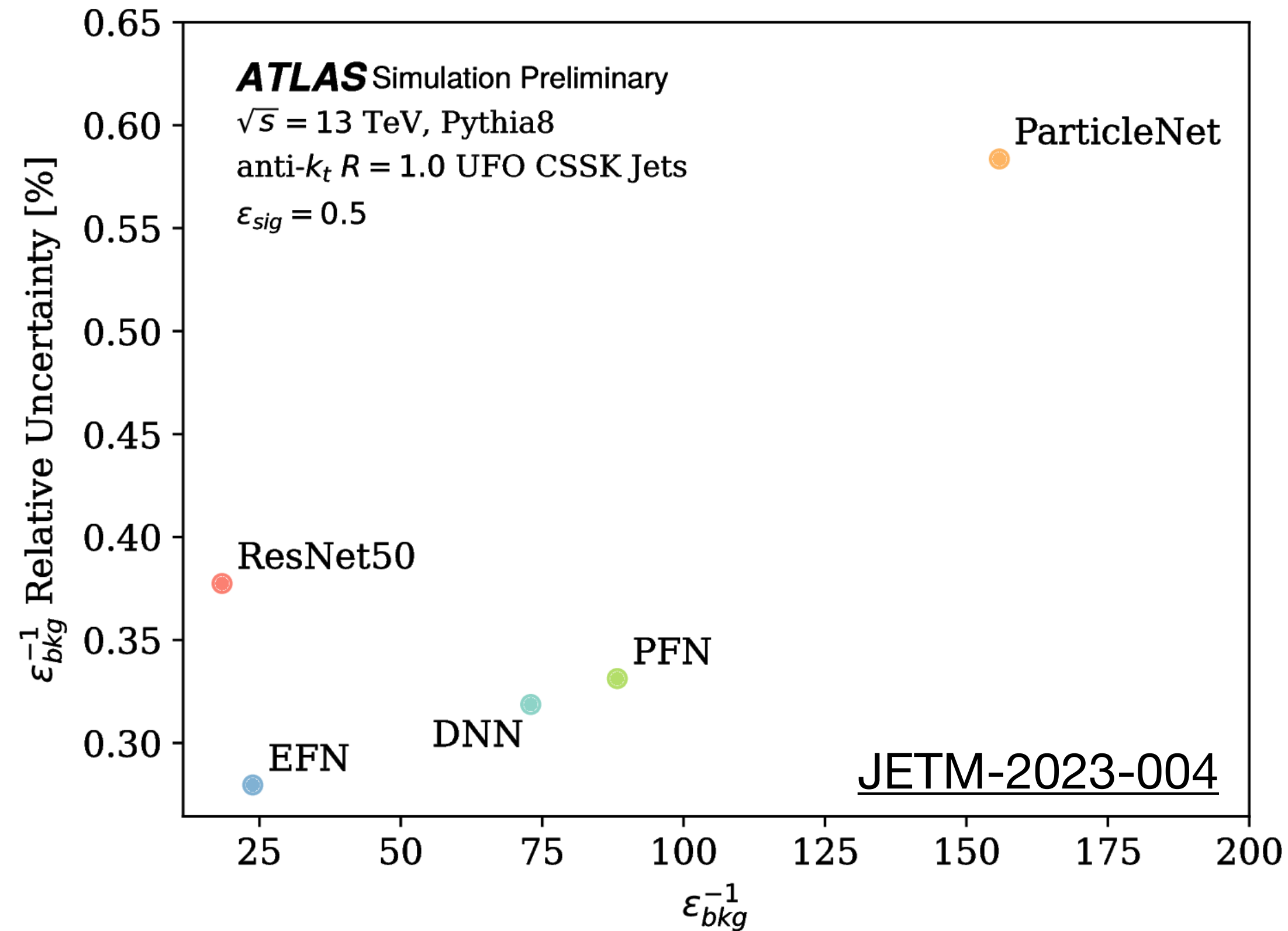


ATLAS, [JETM-2023-003](#)



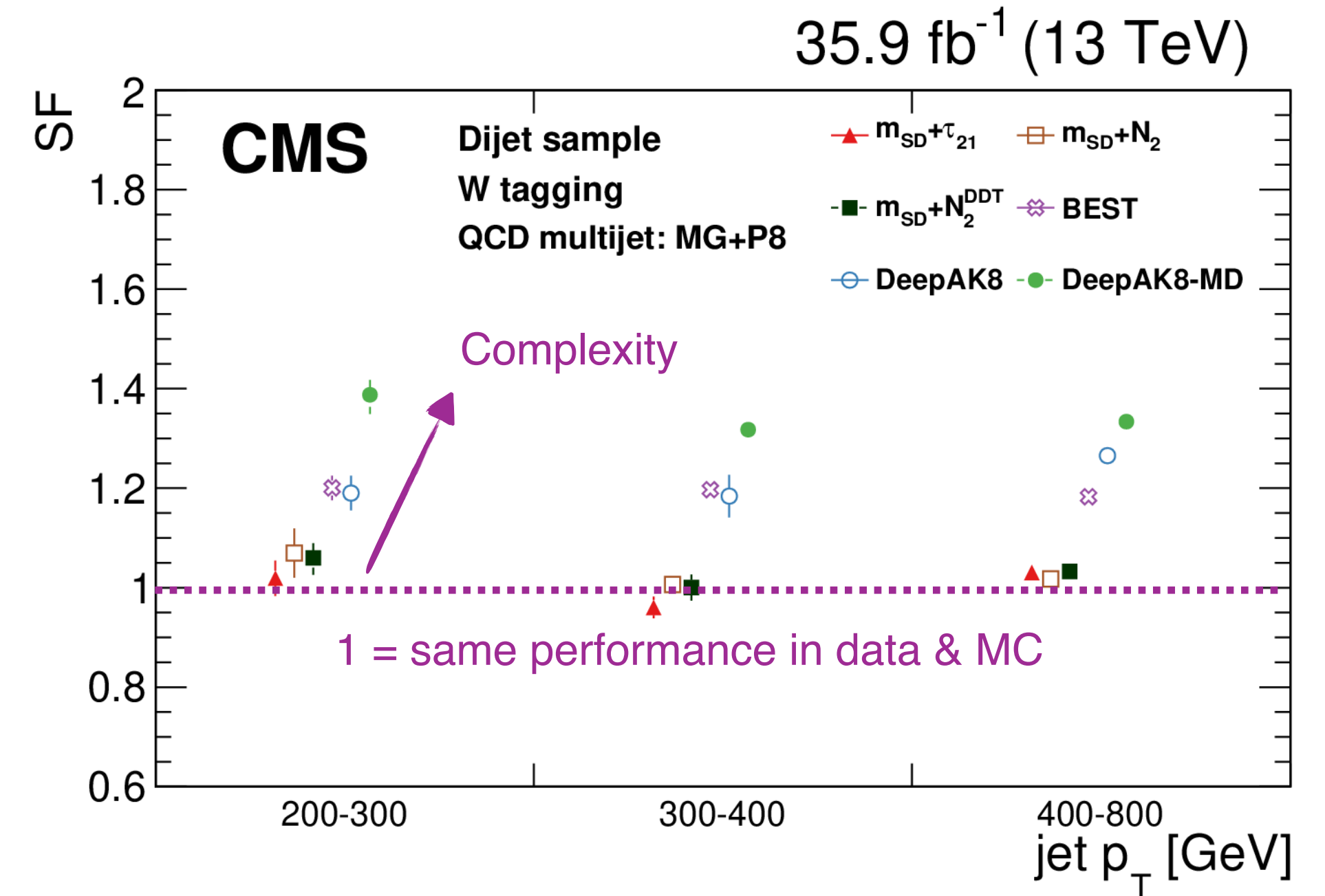
CMS, [JINST 15 \(2020\) P06005](#)

- Be mindful that AUC is not the only important metric!
- Trade-offs between power & precision: *Is performance being realized in data?*



ATLAS Top Tagging Study

Sensitivity to sources of uncertainty increases with tagger complexity.

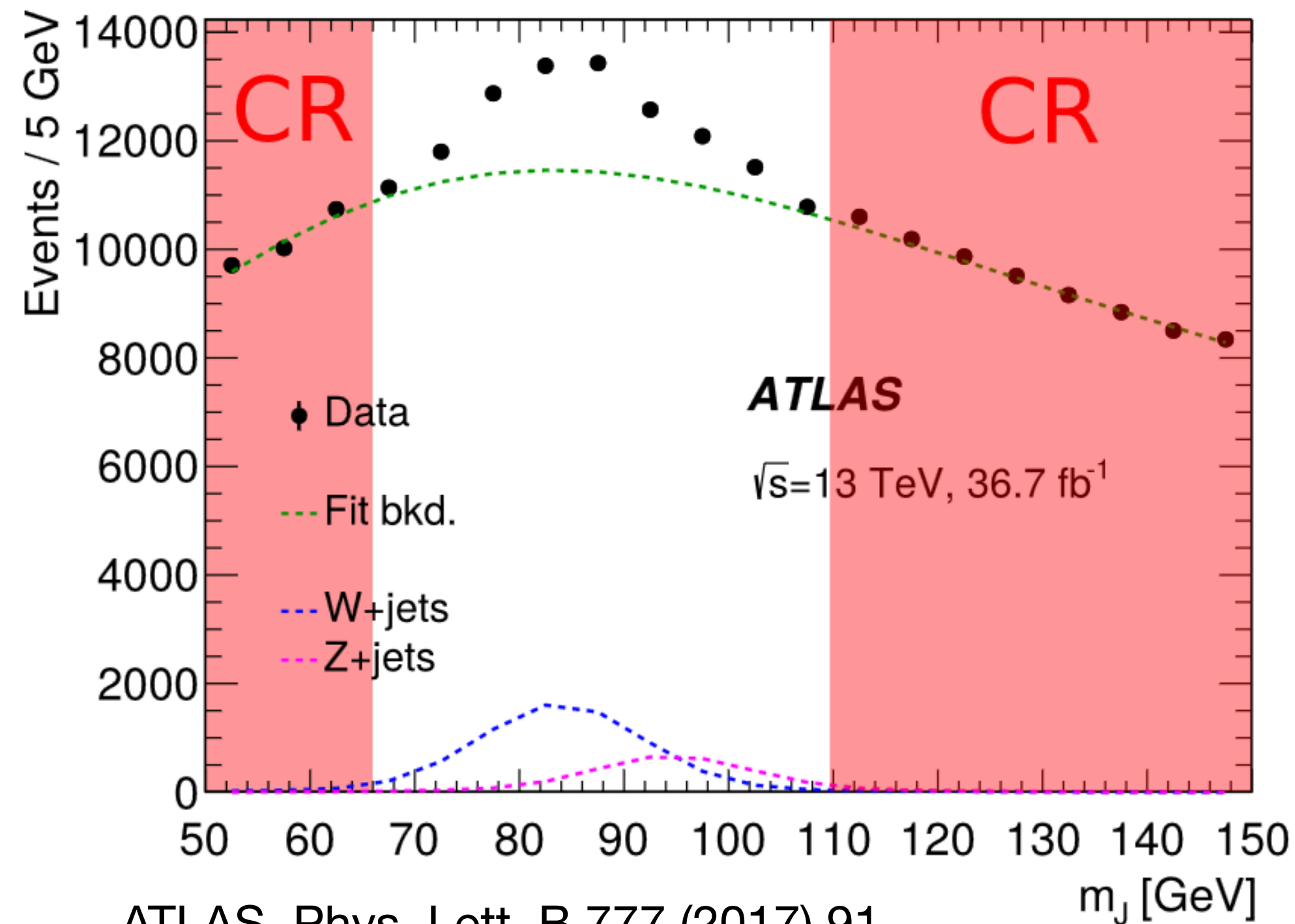


Data/MC Scale Factors for CMS W-Tagger BG Efficiency

~30-40% Data/MC differences for most complex neural network in comparison!
 (DeepAK8, DeepAK8-MD)

Mass decorrelation

- Certain background estimation strategies require taggers that do not sculpt the distribution of interest.
 - e.g. invariant mass spectrum \rightarrow fit sideband regions to estimate contribution in signal region



ATLAS, Phys. Lett. B 777 (2017) 91

- Can de-correlate tagger performance from observable of interest.
 - Analytical approaches (Dolen et al. [JHEP 05 \(2016\) 156](#))
 - ML-based approaches (e.g. ANNs, [ATL-PHYS-PUB-2021-029](#))

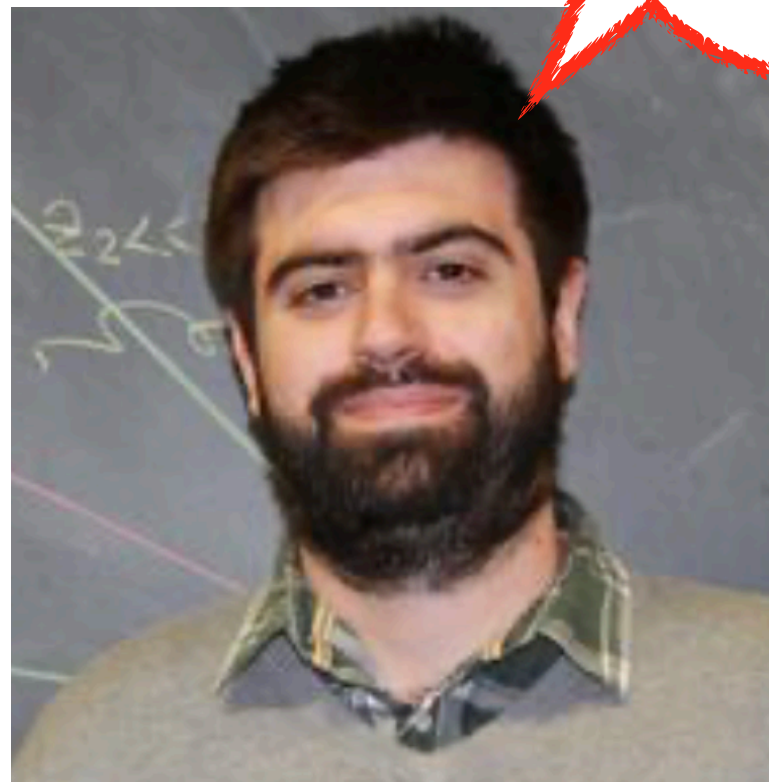
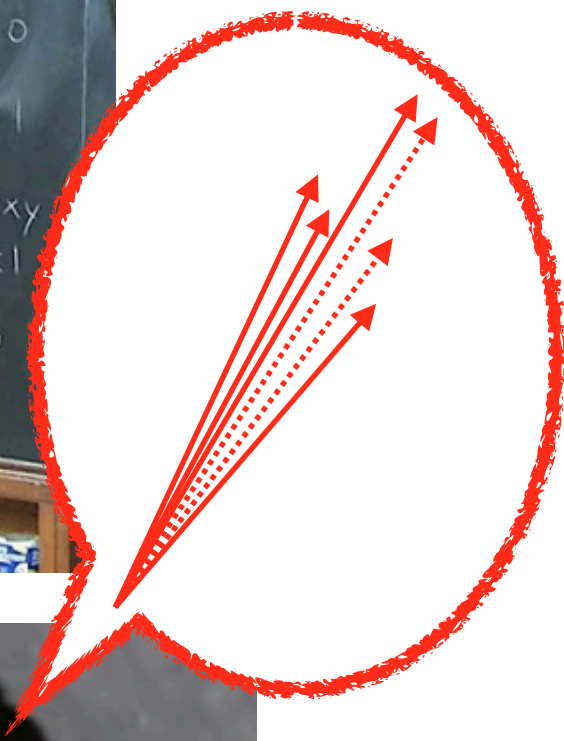
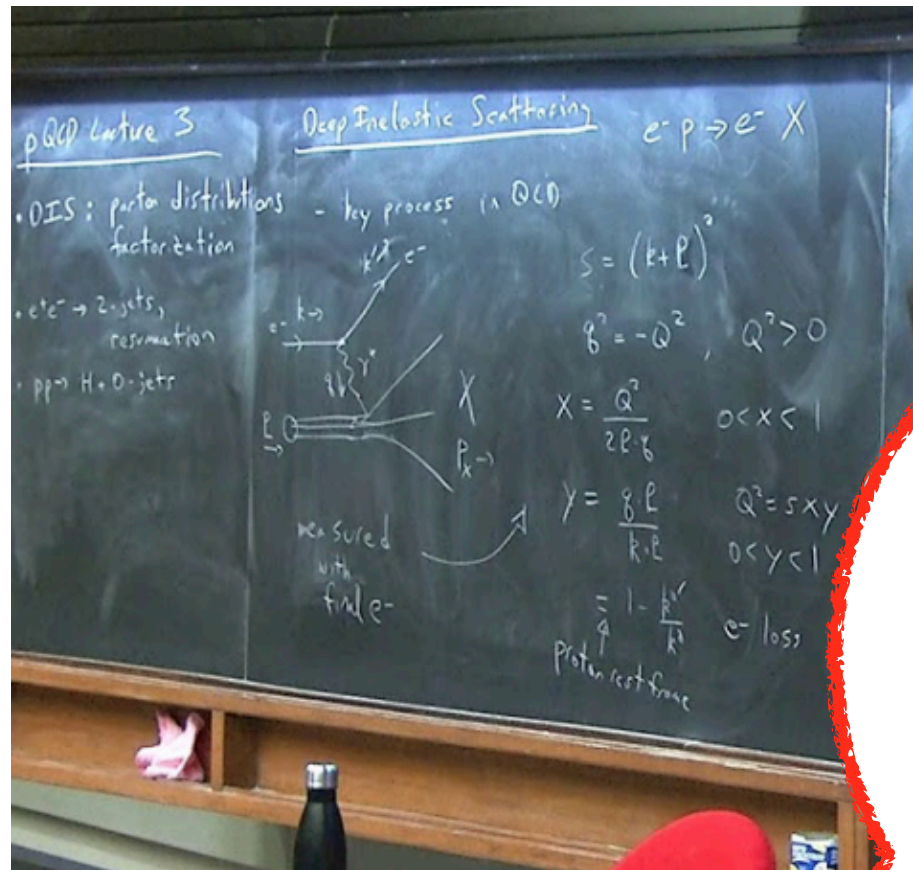
Unfolding



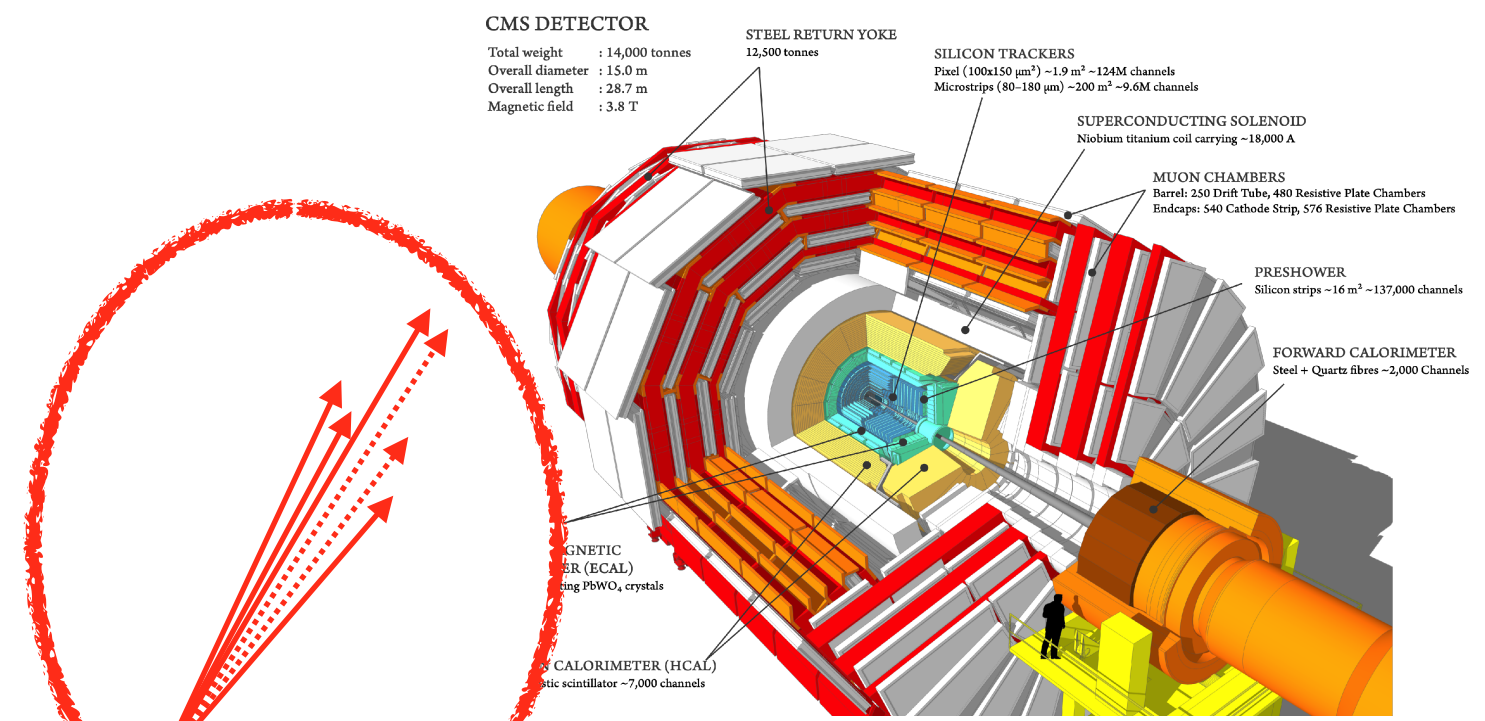
Theorist



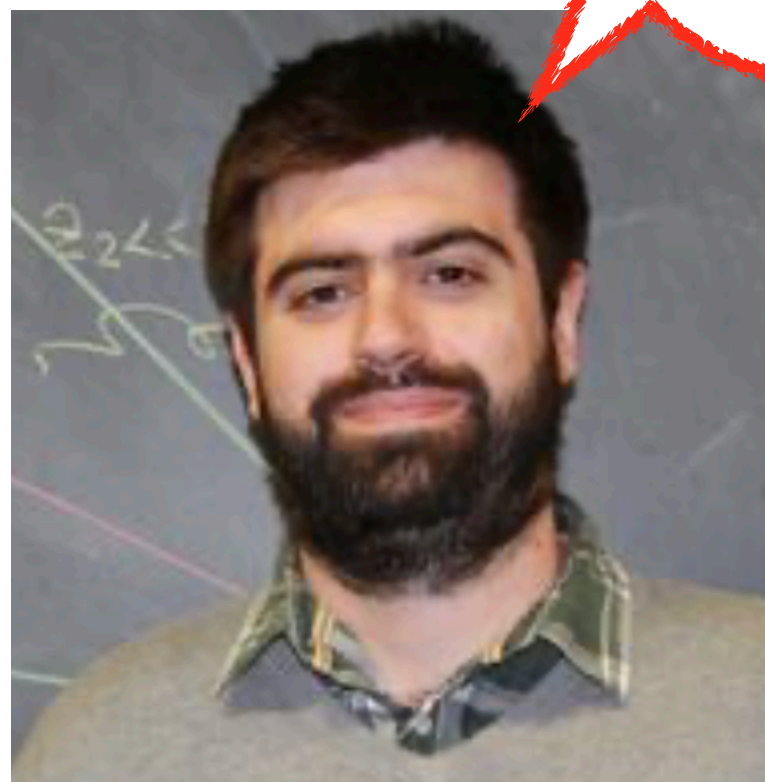
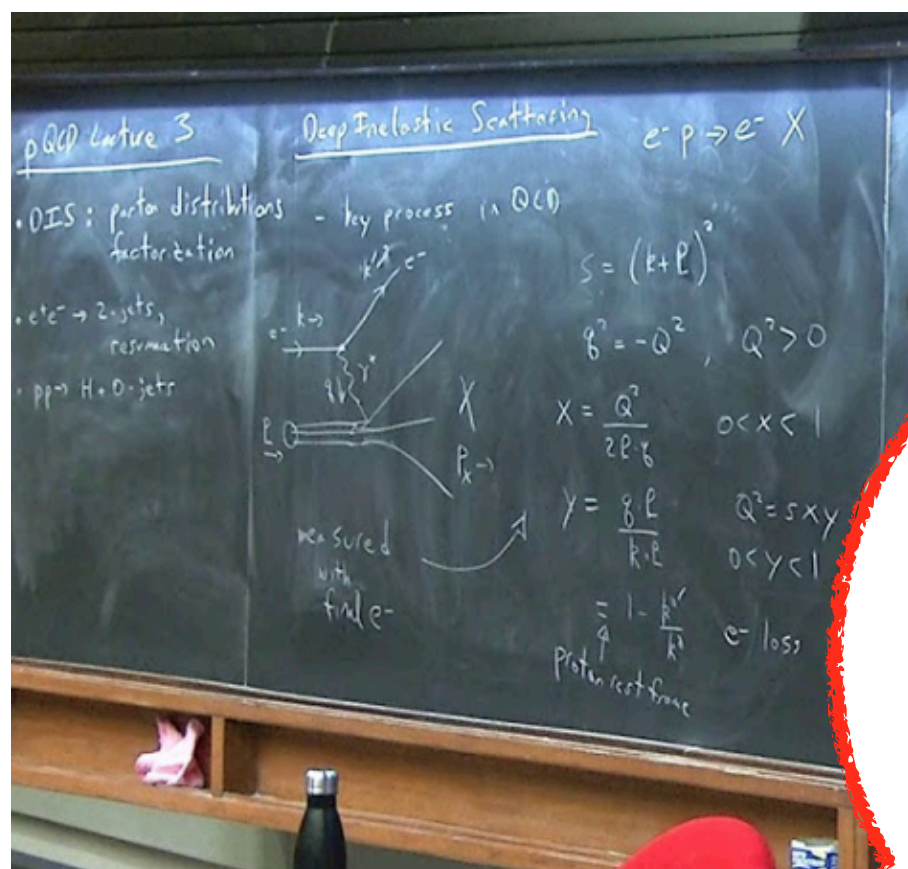
Experimentalist



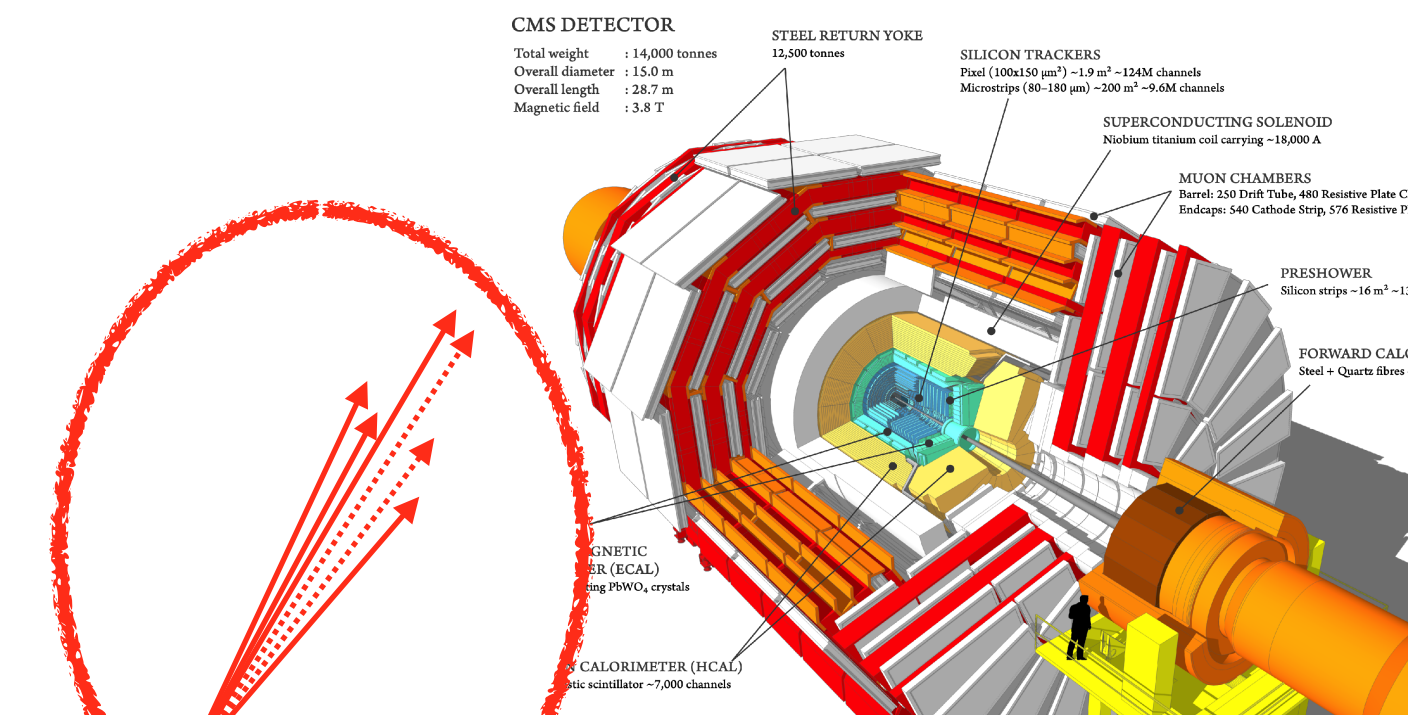
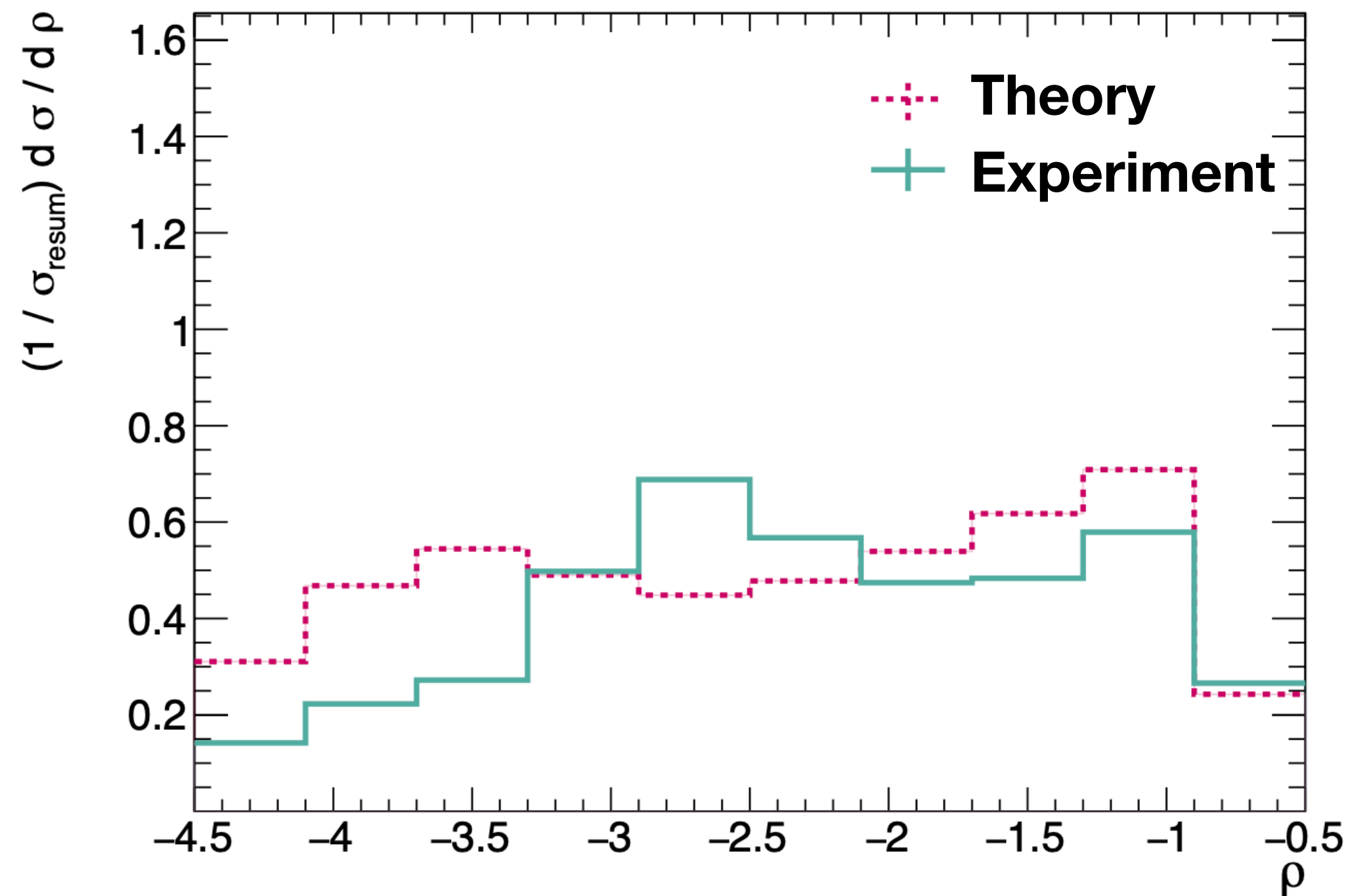
Theorist



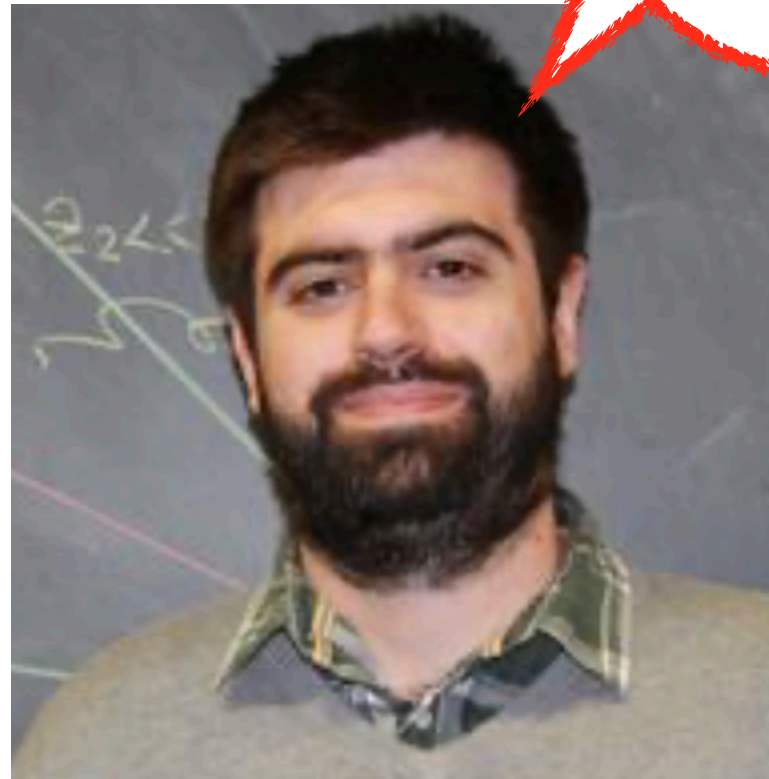
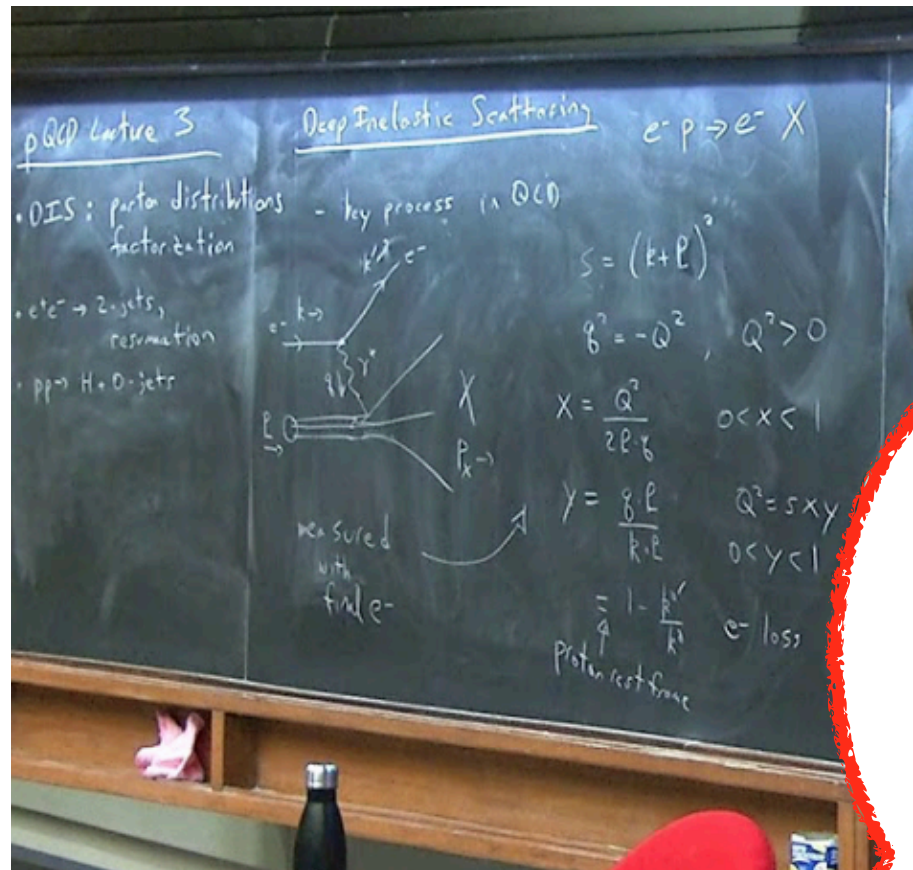
Experimentalist



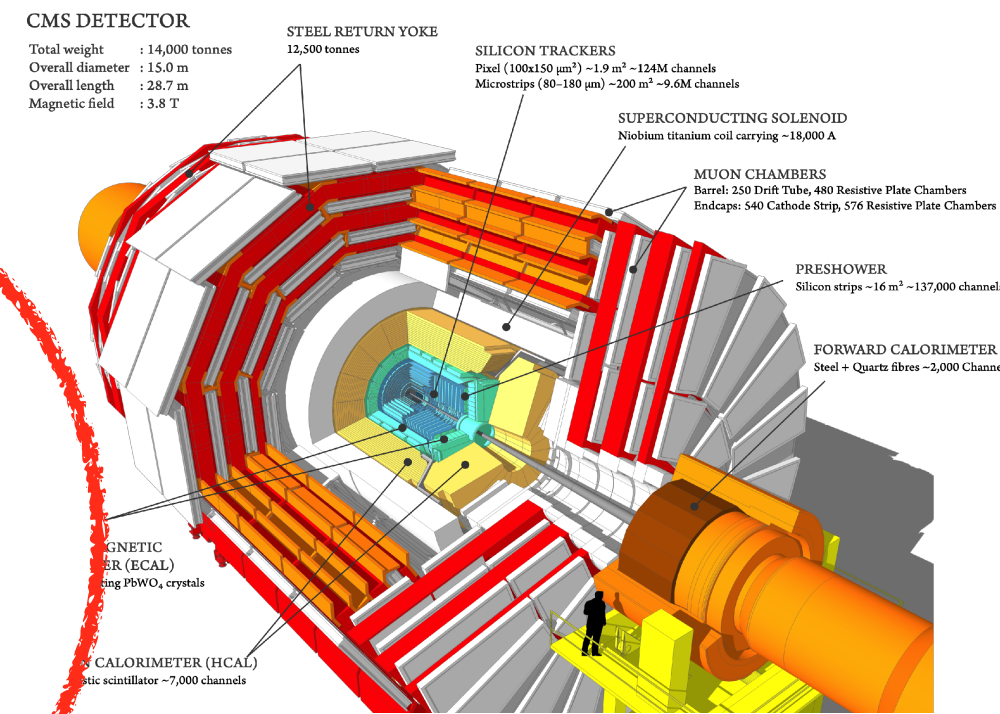
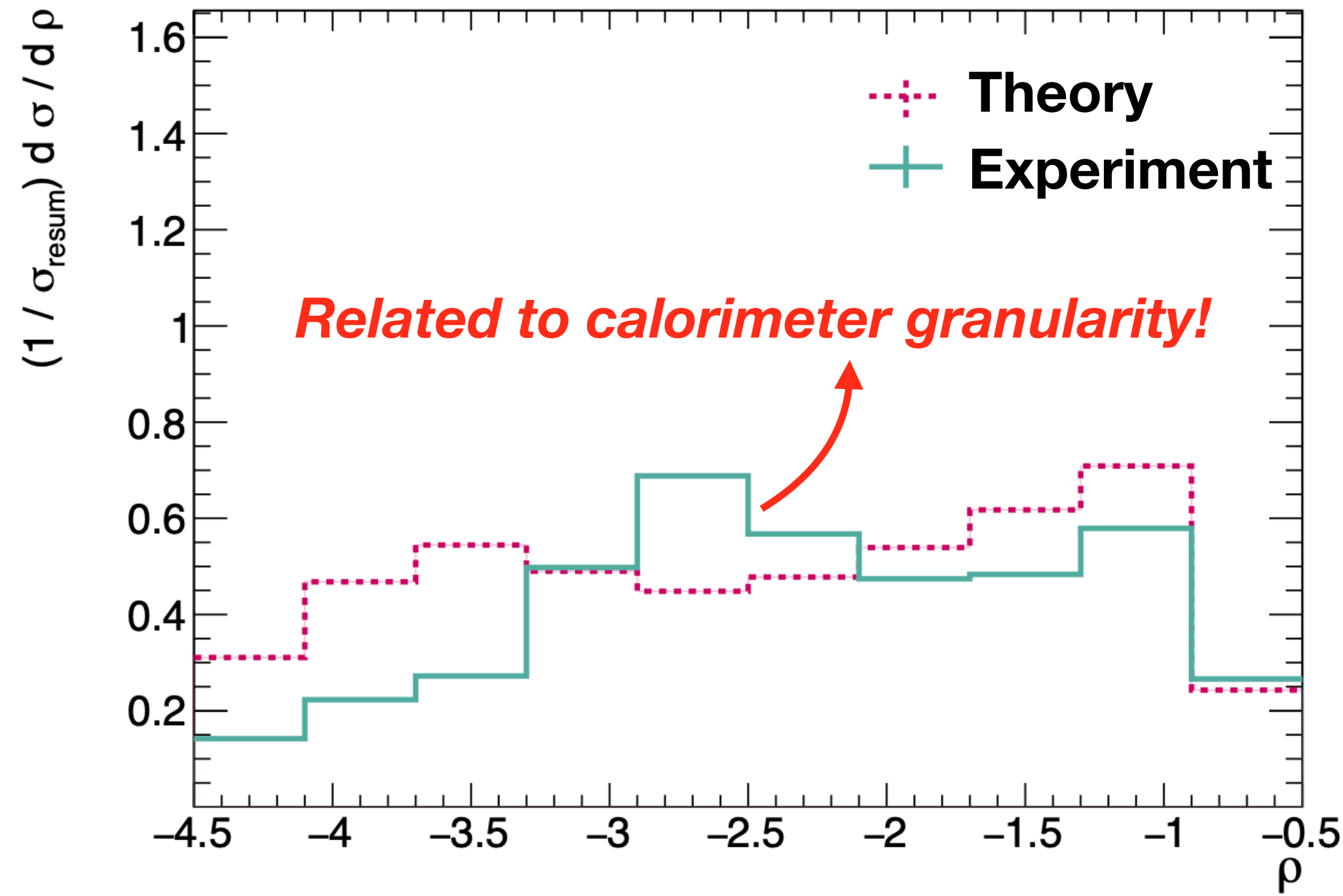
Theorist



Experimentalist

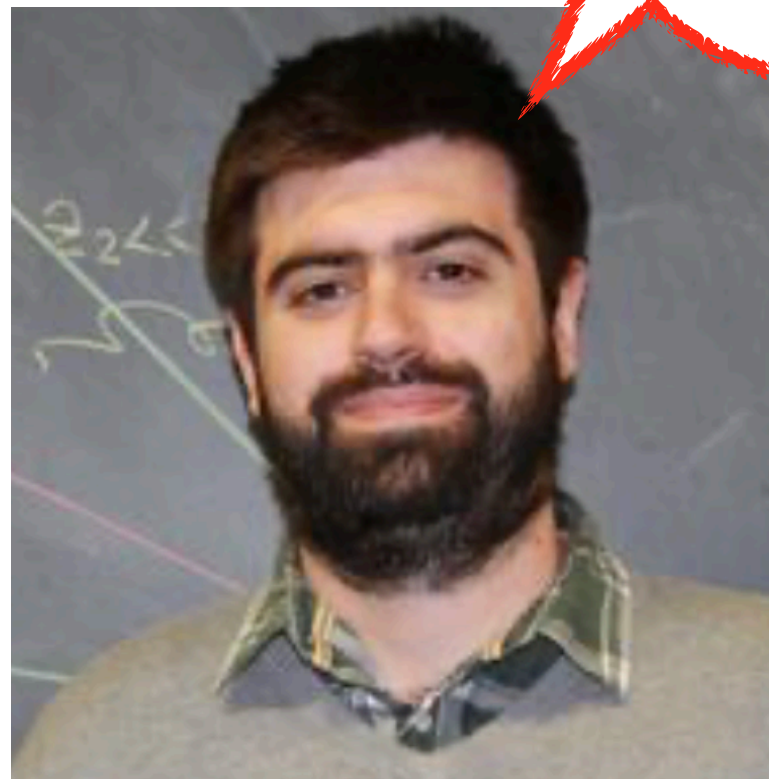
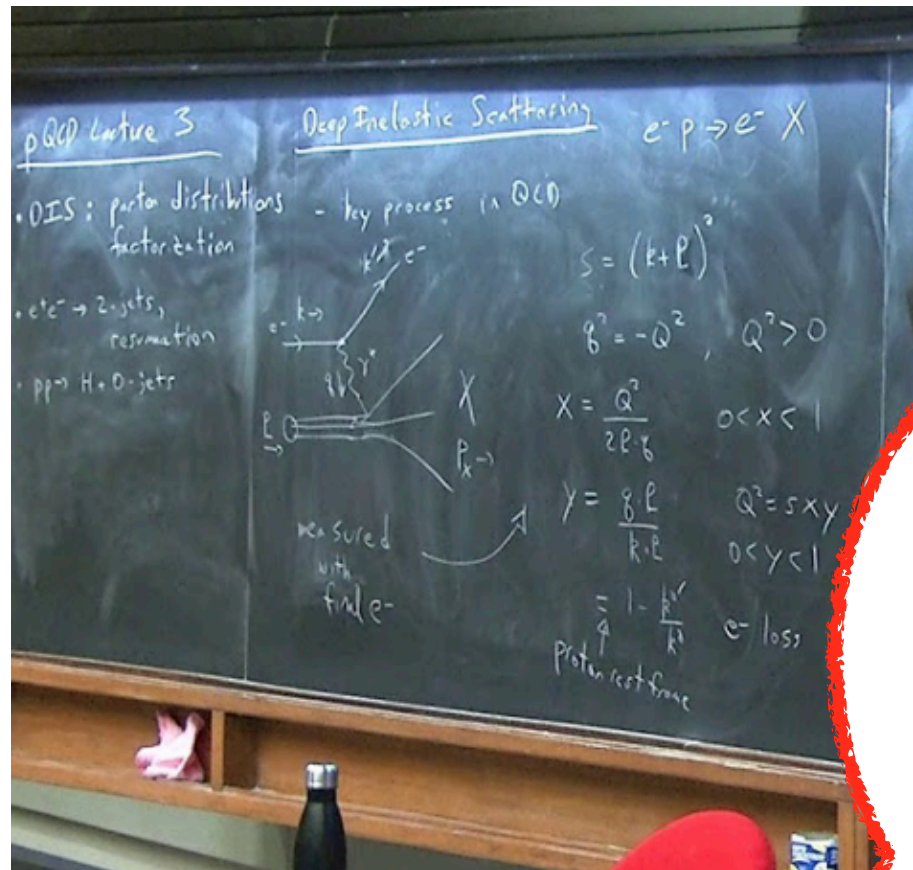


Theorist

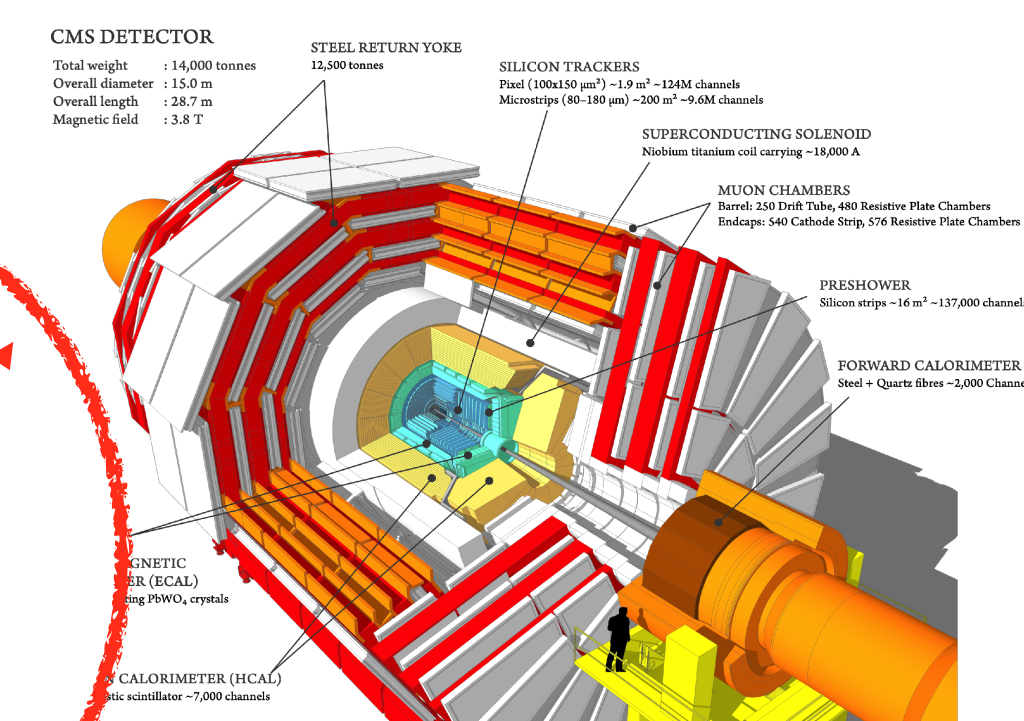
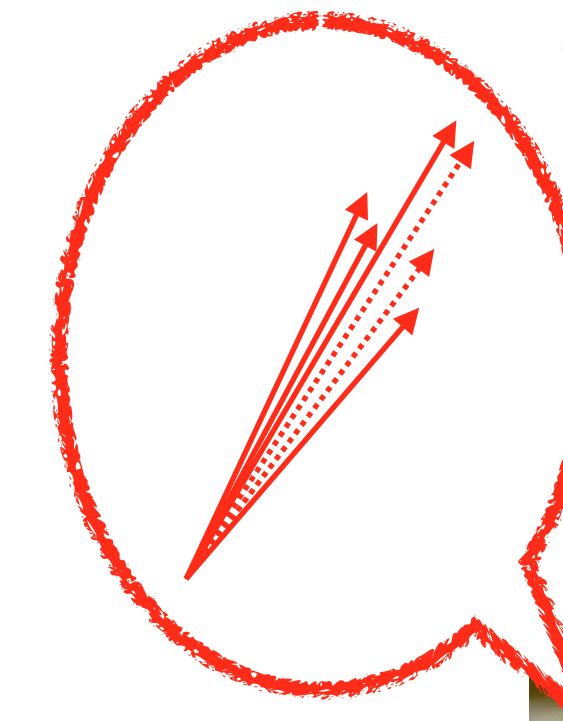
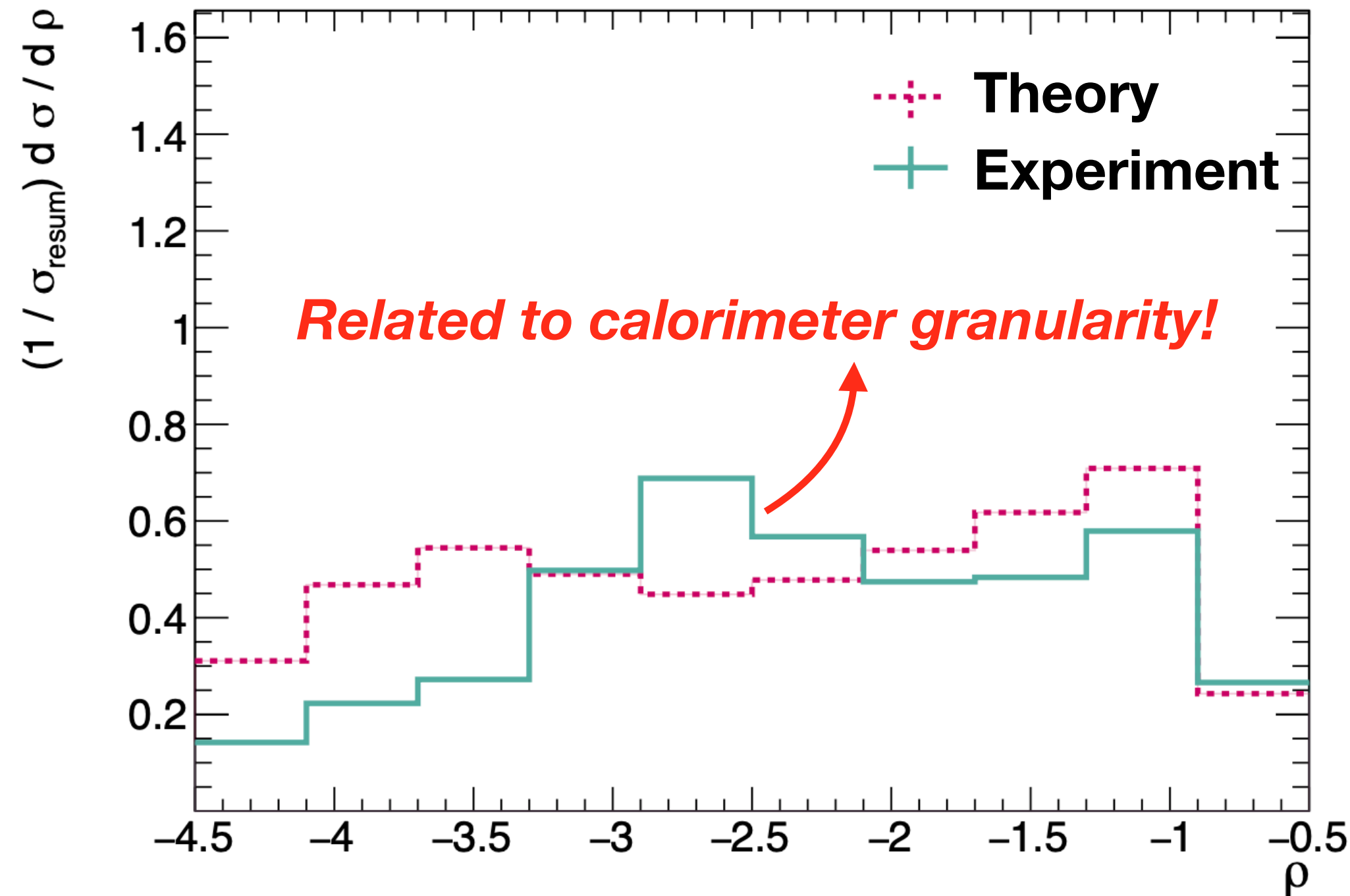


Experimentalist

Unfolding corrects for detector-related resolution & acceptance effects.



Theorist



Experimentalist

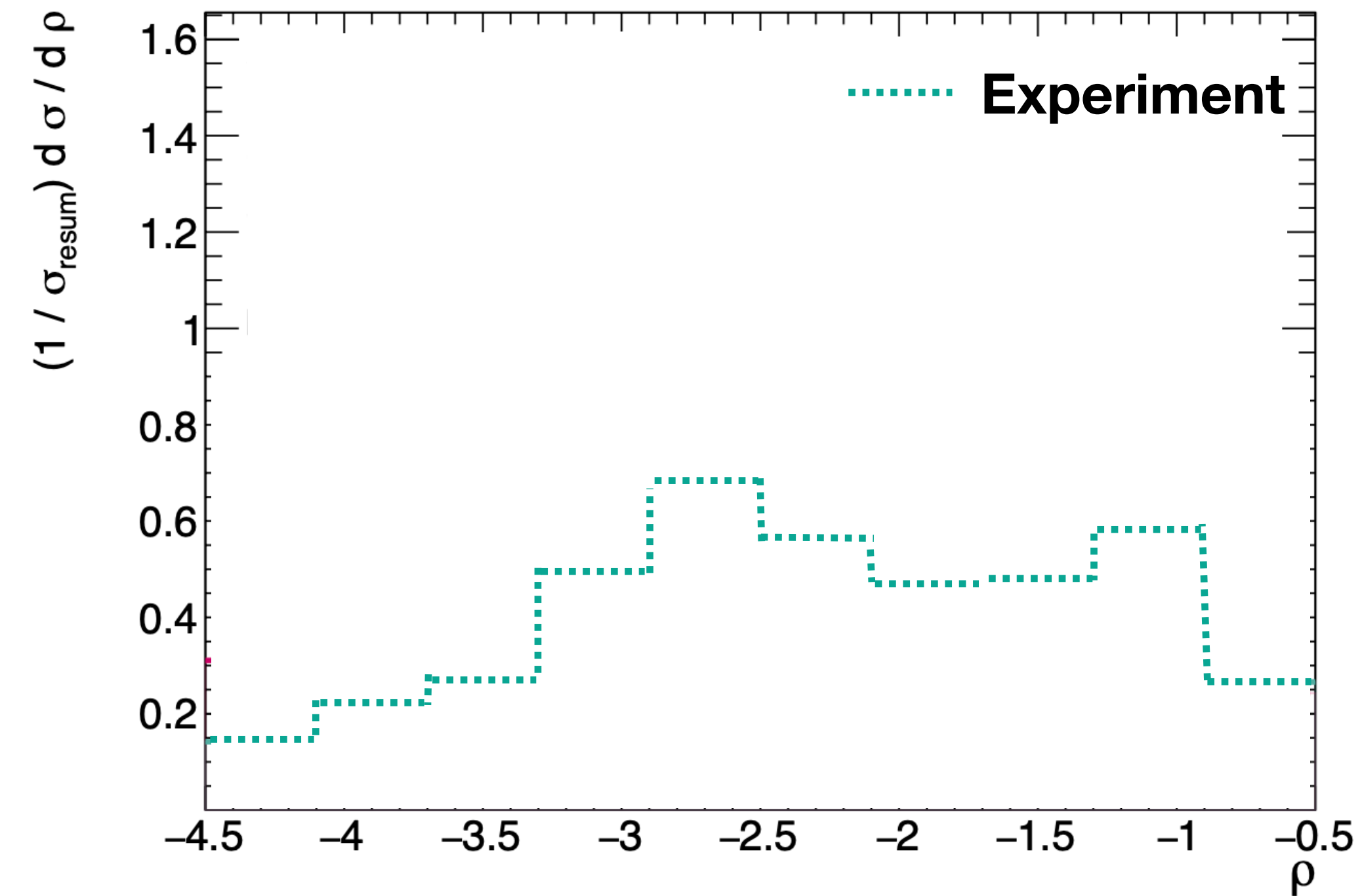
n.b. Could also Fold the theory
But other advantages to having unloaded distributions

(Availability for MC tuning, PDF fits, CONTUR, etc.)

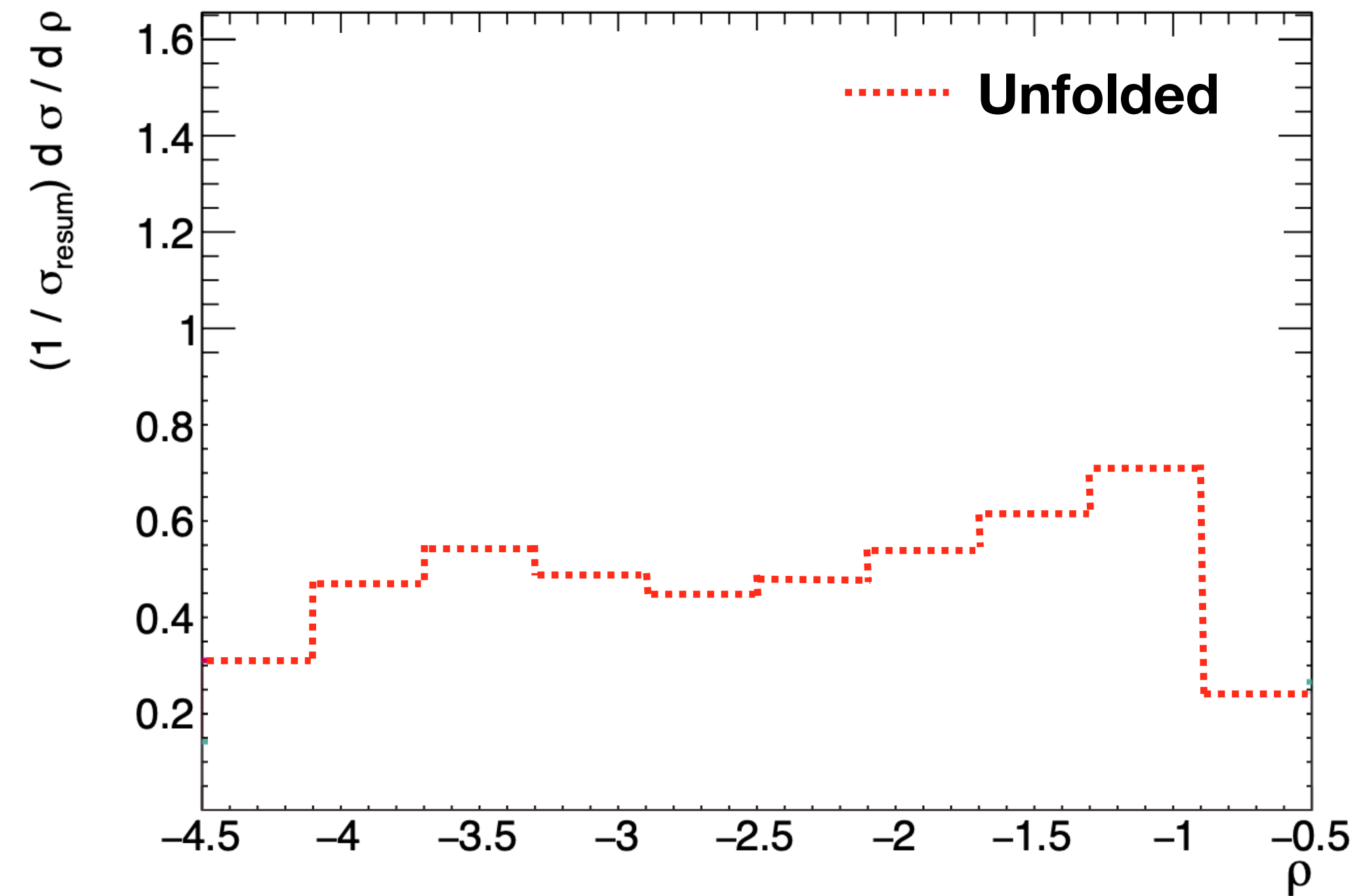
$$\mathbf{g} = \mathbf{K}\boldsymbol{\lambda}$$

- \mathbf{g} : observed detector-level distribution (“smeared”)
- $\boldsymbol{\lambda}$: desired particle-level distribution
- \mathbf{K} : kernel / response function (map between spaces)

$$\mathbf{g} = \mathbf{K}\boldsymbol{\lambda}$$

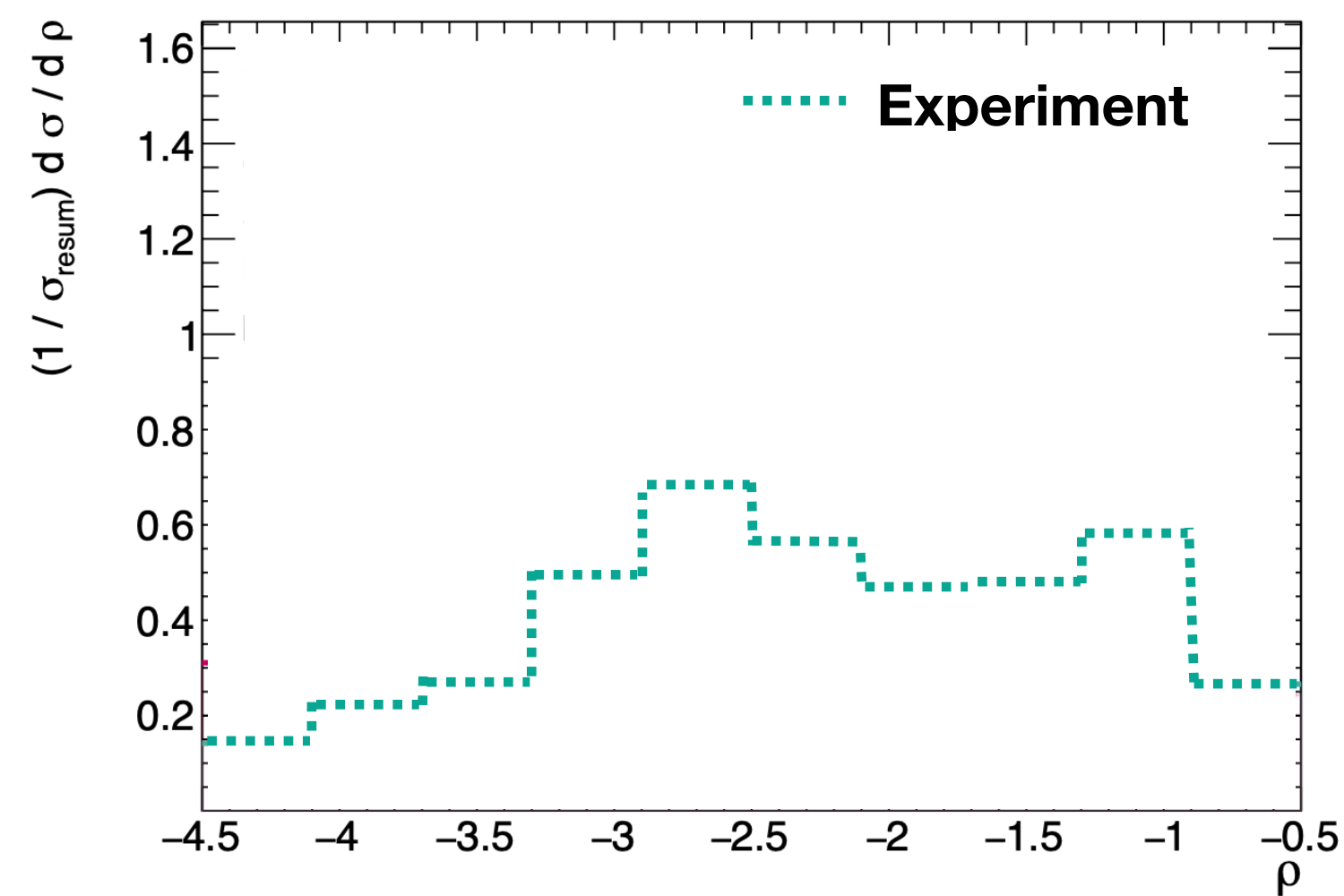


$$= \mathbf{K}$$

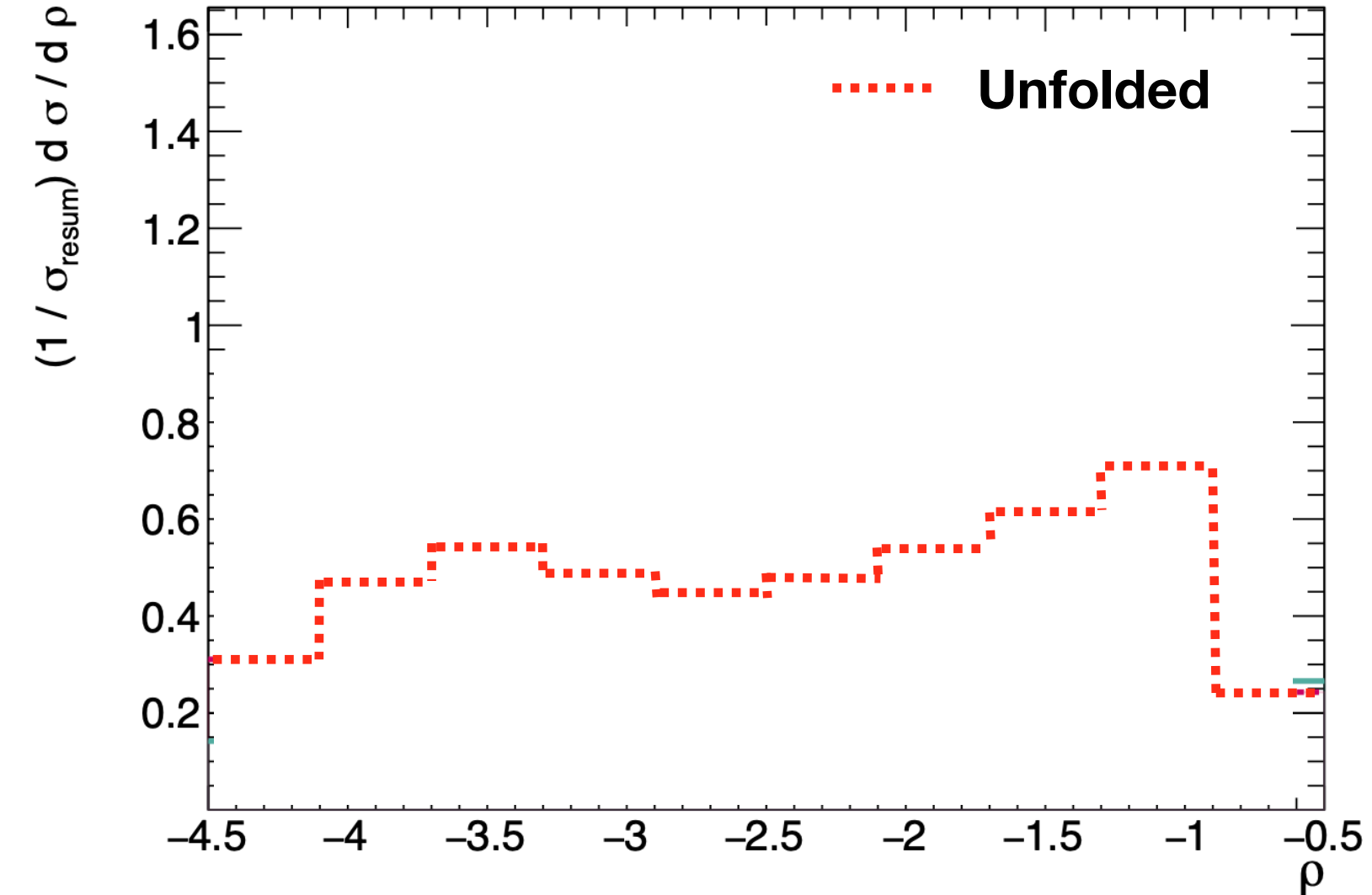
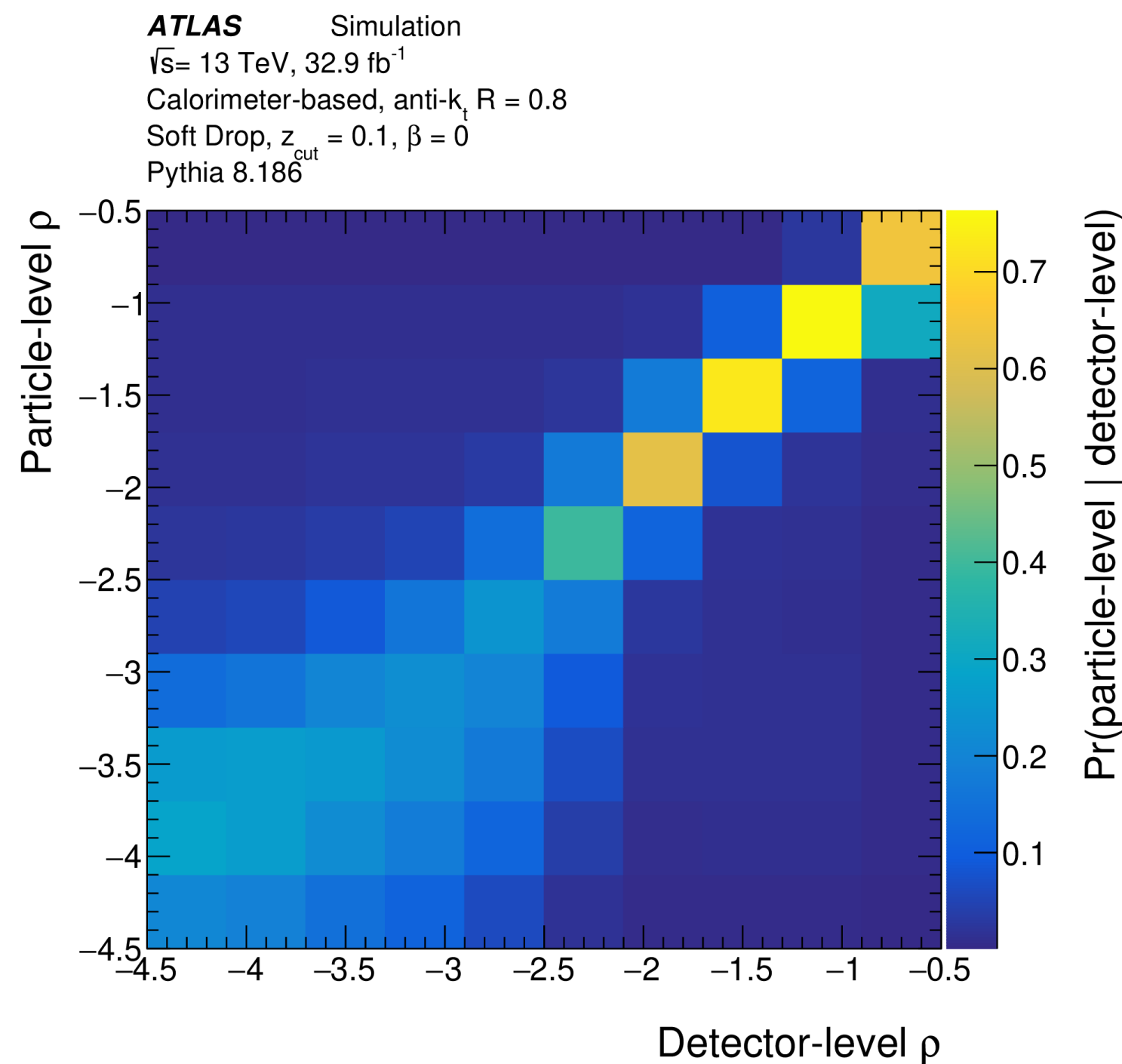


- \mathbf{g} : observed detector-level distribution ('smeared')
- $\boldsymbol{\lambda}$: desired particle-level distribution
- \mathbf{K} : kernel / response function (map between spaces)

$$\mathbf{g} = \mathbf{K} \boldsymbol{\lambda}$$



=



- \mathbf{g} : observed detector-level distribution ('smeared')
- $\boldsymbol{\lambda}$: desired particle-level distribution
- \mathbf{K} : kernel / response function (map between spaces)

MC used to obtain \mathbf{K} !

More migrations \rightarrow more reliant on MC.
(Uncertainties from modeling, non closure etc.)

Unfolding methods & metrics (not exhaustive)

- Unfolding is an “ill-posed” inverse problem; many different approaches to solving it:
 - Binned (matrix-based):
 - Singular Value Decomposition (SVD)
 - Iterative Bayesian Unfolding (D’Agostini)
 - ML-based (see: 2211.01421 Sec. 5):
 - Classifier-reweighing: e.g. Omnifold
 - Generative: e.g. cINNs, flow matching, diffusion [1,2,3,4]
 - Regularization ou non?
 - **Choice of method surprisingly contentious: not a good use of time to argue about them now.**
 - 2024 France-Berkeley Unfolding Workshop: <https://indico.cern.ch/event/1357972/>
- Often asking **similar questions** to assess the performance of these algorithms:
 - **How diagonal is the kernel?**
(What is the experimental resolution?
→ determines bin granularity)
 - **Does changing the prior change the result? Is the method converging?**
(→ Non-closure uncertainty)
 - **Is there sensitivity to the MC model used to construct the kernel? Hidden variables?**
(→ MC modeling uncertainty, *etc.*)
 - Questions tend to matter even more as the kernel becomes further off-diagonal.
 - If there is a large bias, estimates of uncertainties can be untrustworthy.

Other aspects: efficiency & purity (fakes) corrections

Nice examples — ATLAS Diphotons @ 13 TeV, JHEP 11 (2021) 169

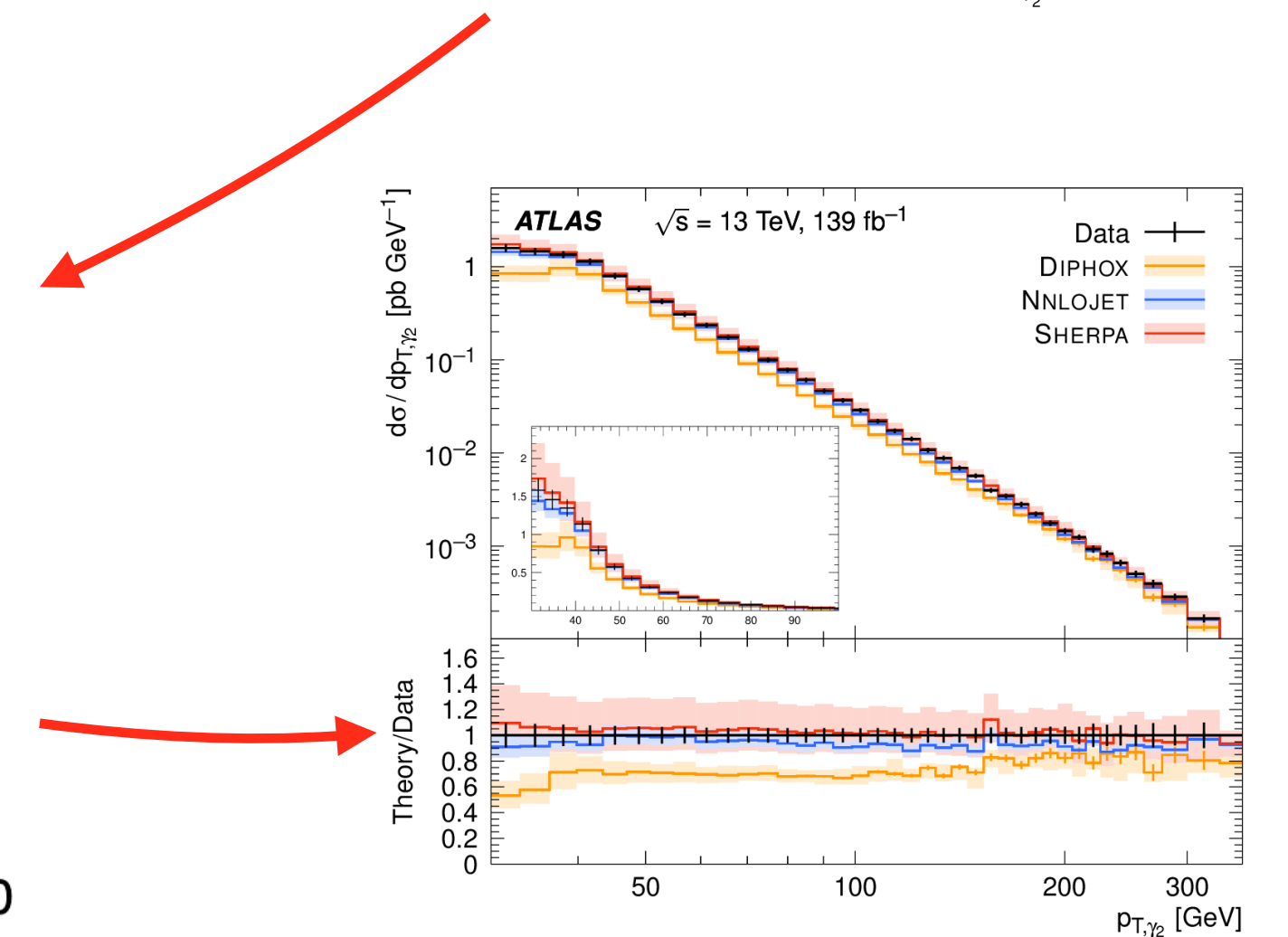
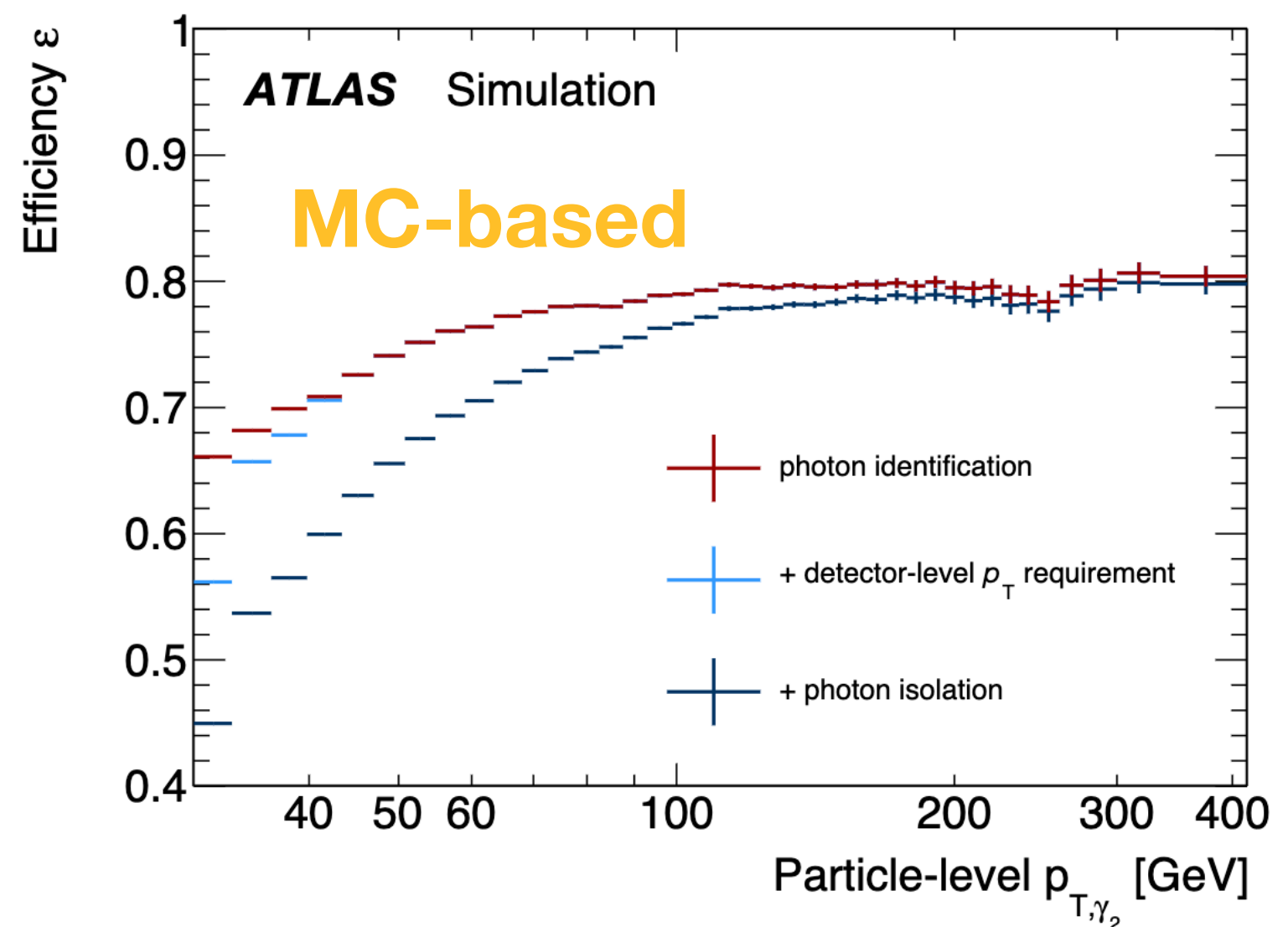
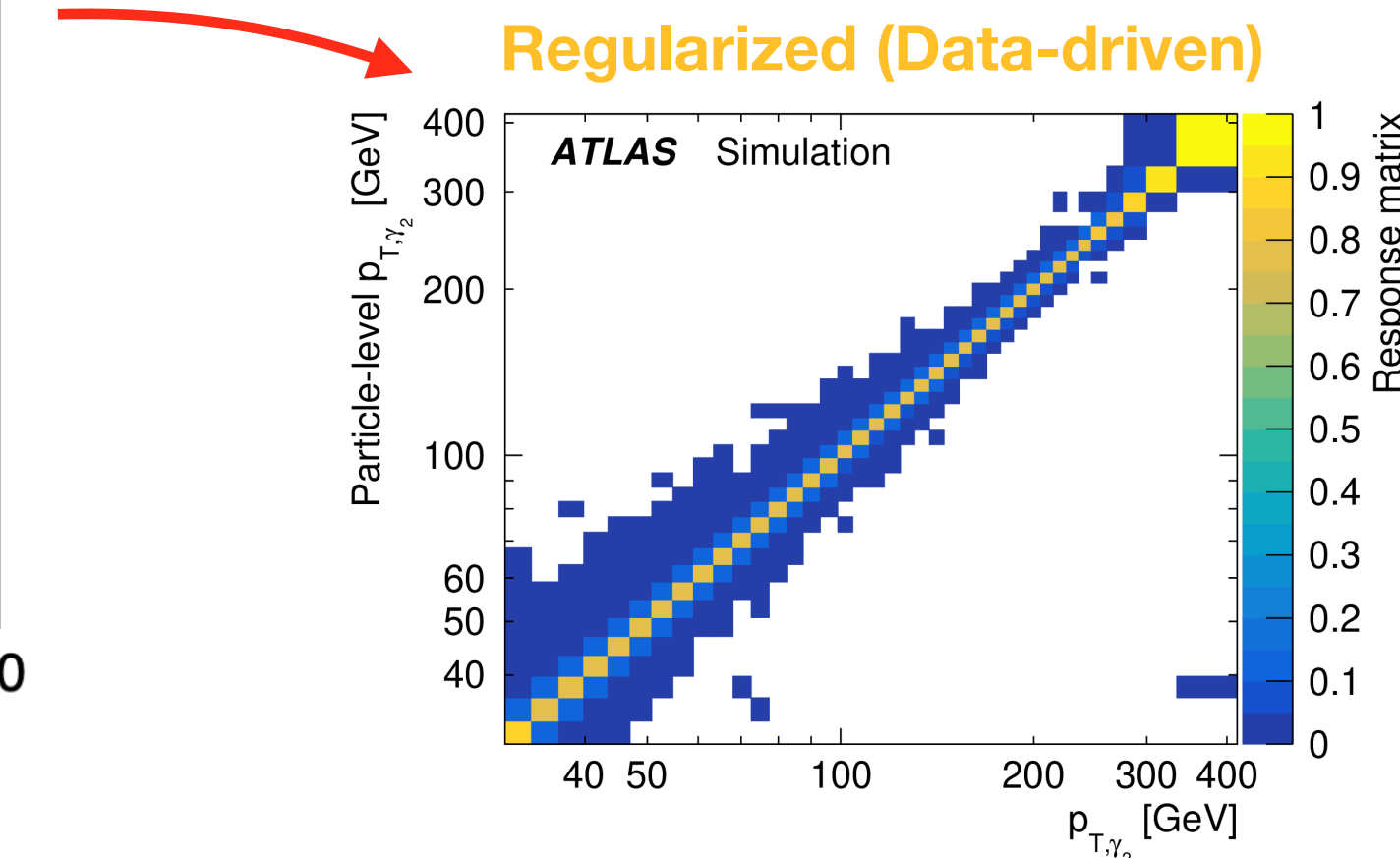
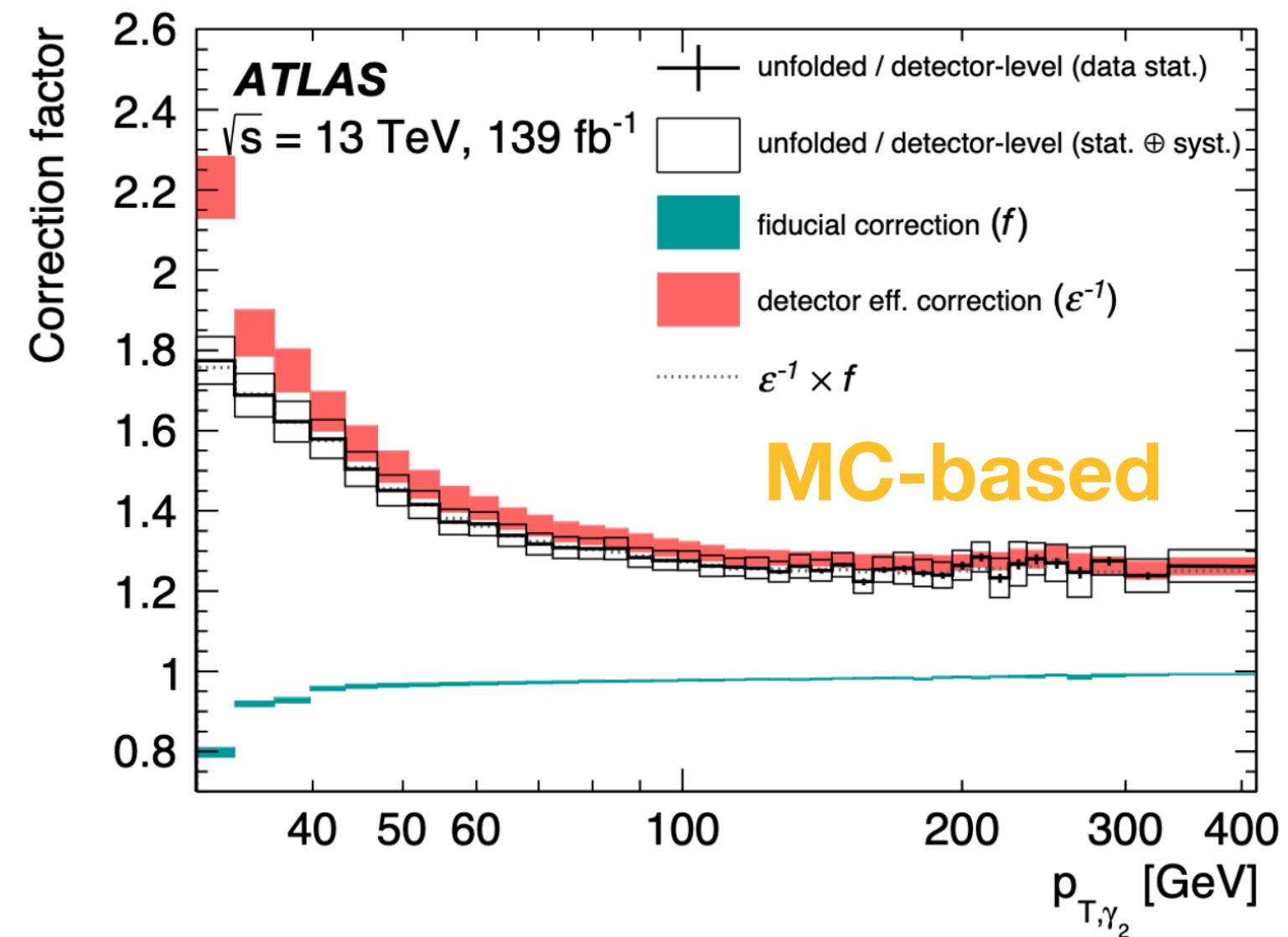
- Also need to account for acceptance-related effects (under/overflow):

- Fiducial ('fake-factor') corrections :** pass detector-level selection, not truth-level selection

- (In)Efficiency correction :** pass truth-level selection, not detector-level selection

- Often estimated directly from MC.**

- Can add 'sideband' under/overflow regions to kernel to decrease extrapolations, reliance on MC.



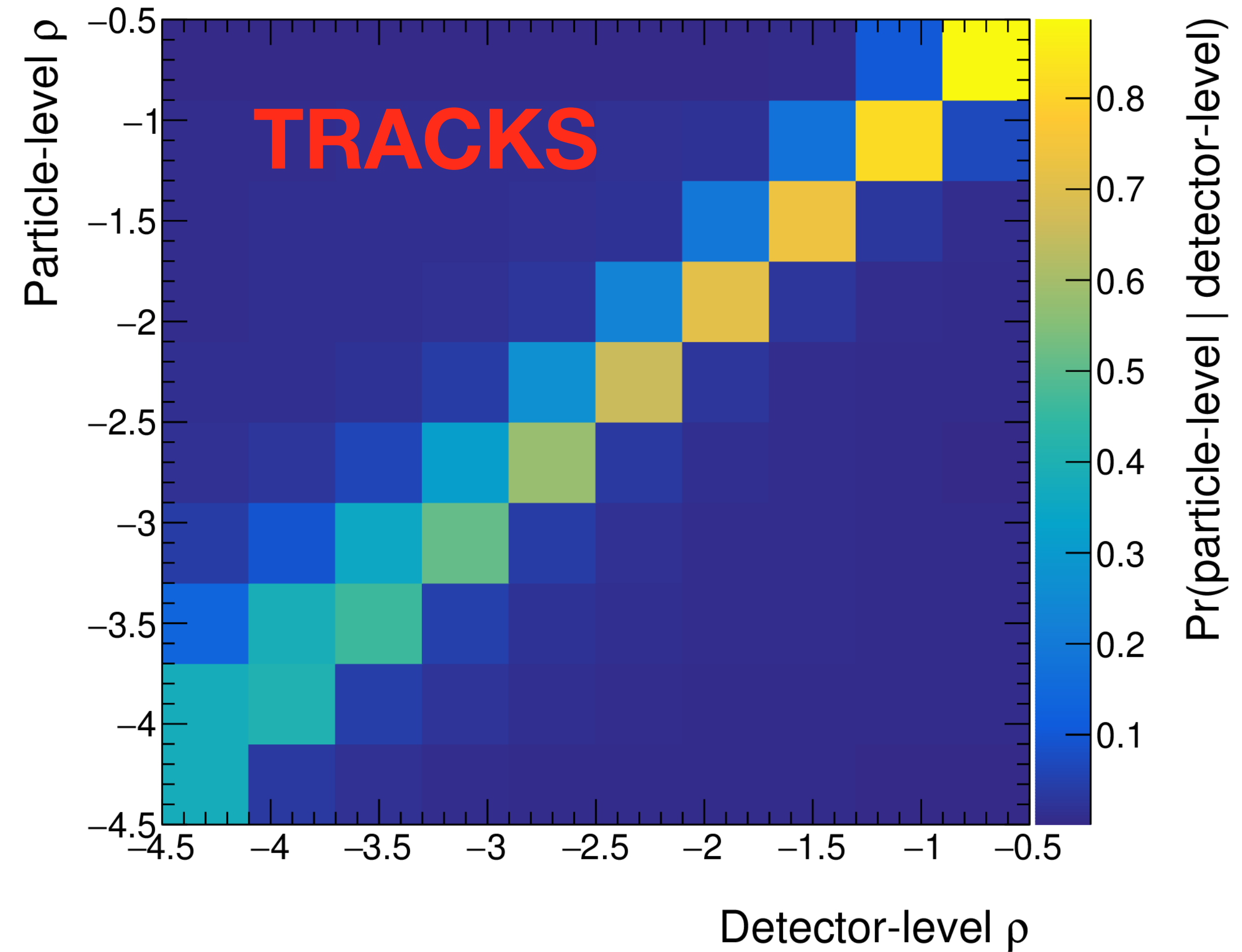
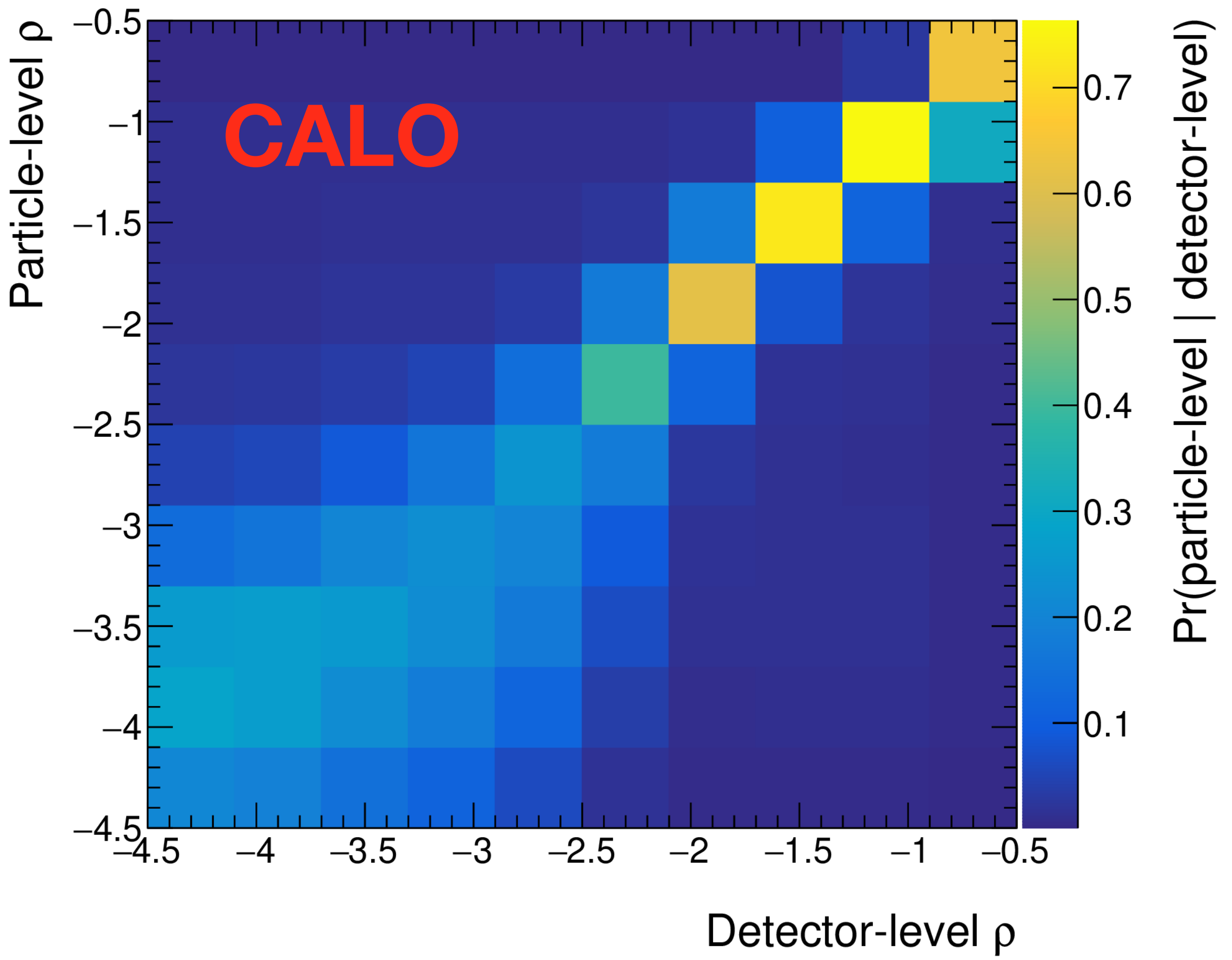
n.b. results end at 300 GeV, but matrix goes further!

Soft-drop observables: calo vs. track response matrices

ATLAS, *Phys. Rev. D* 101, 052007 (2020), *Phys. Rev. Lett.* 121, 092001 (2018)

ATLAS Simulation
 $\sqrt{s} = 13 \text{ TeV}, 32.9 \text{ fb}^{-1}$
Calorimeter-based, anti- k_t , $R = 0.8$
Soft Drop, $z_{\text{cut}} = 0.1, \beta = 0$
Pythia 8.186

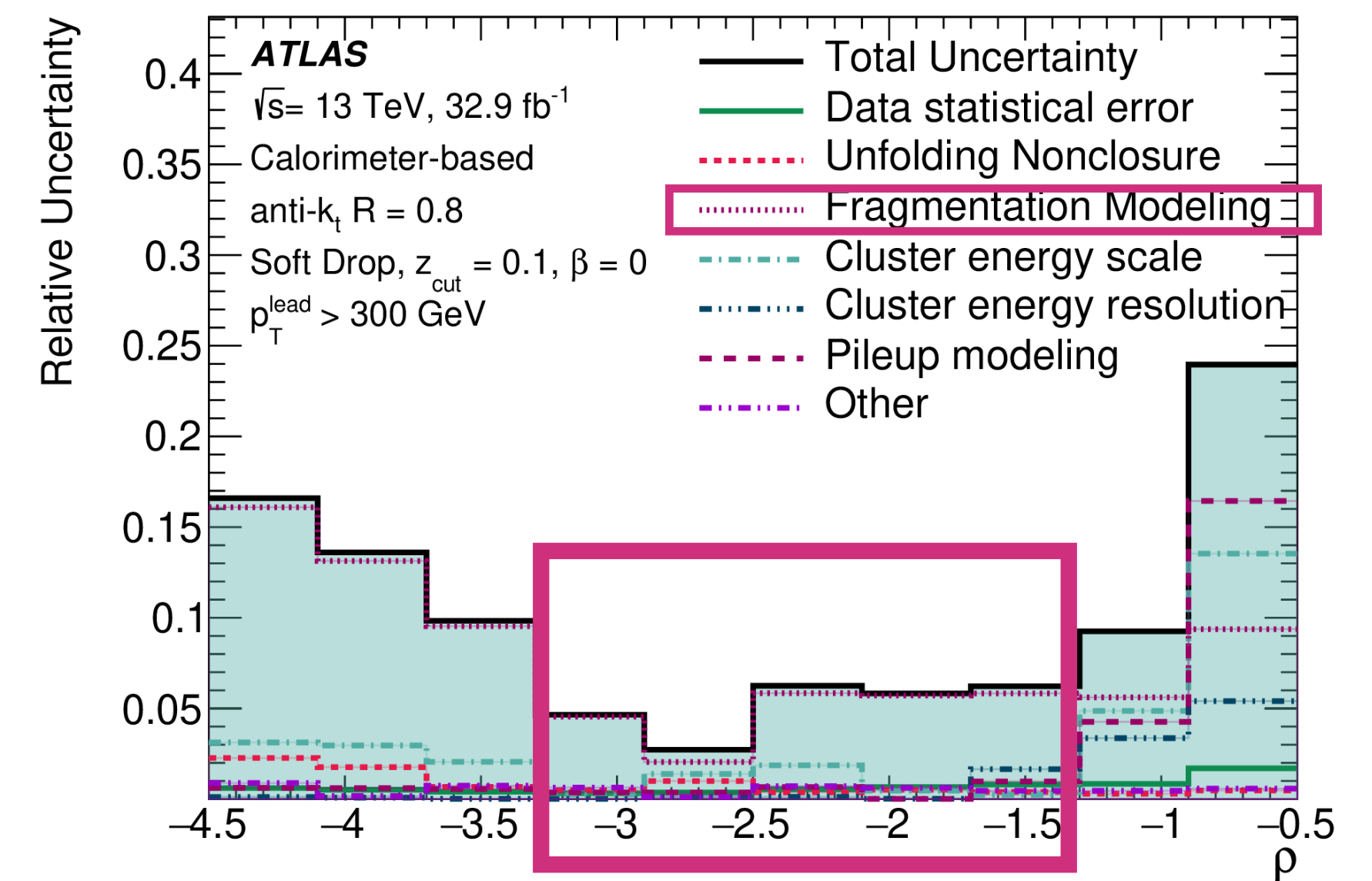
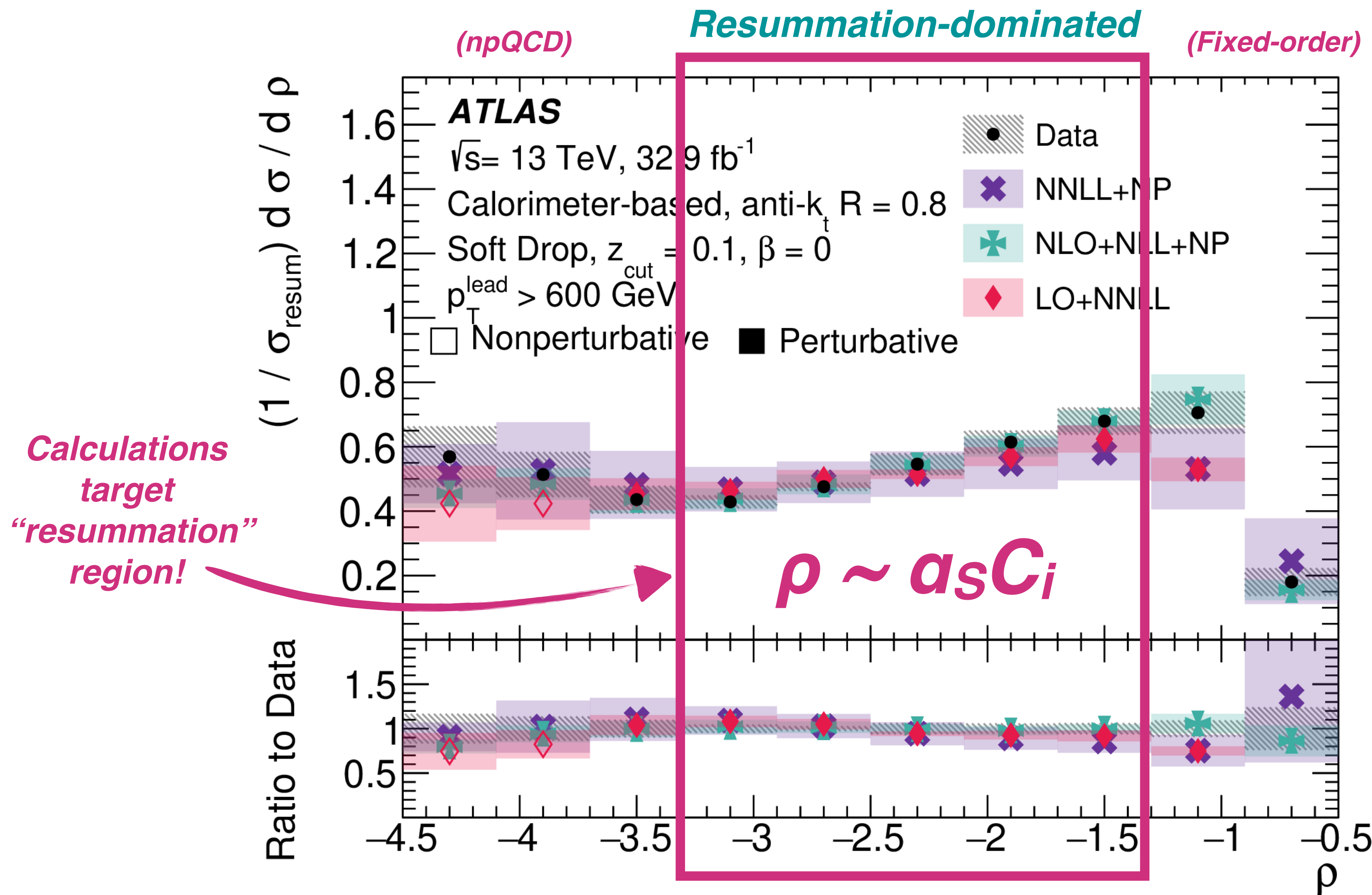
ATLAS Simulation
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Track-based, anti- k_t , $R = 0.8$
Soft Drop, $z_{\text{cut}} = 0.1, \beta = 0$
Pythia 8.186



Soft-Drop Jet Mass

ATLAS, *Phys. Rev. D* 101, 052007 (2020),
 & earlier, *Phys. Rev. Lett.* 121, 092001 (2018)

“Relative Mass”
 $\rho = \log(m_{SD}^2/p_T^2)$



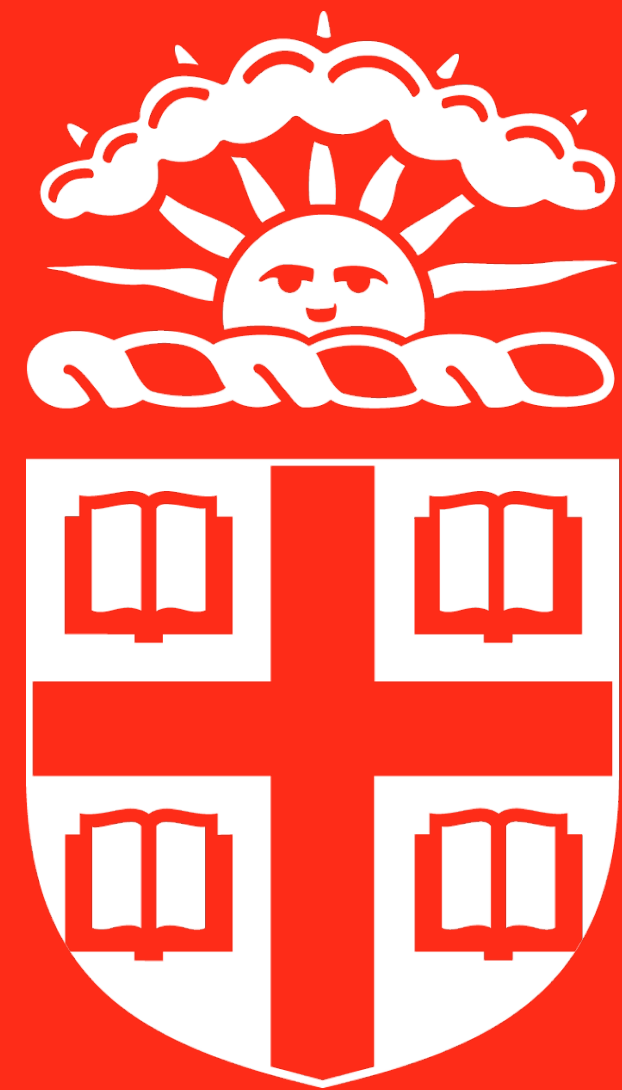
Concluding remarks

Recap:

- **Detectors:** increased segmentation allows for better signal calibration and less readout noise. Please remember this when you are thinking about future collider detectors!
- **Particle Flow:** optimal combination of calo & tracker signals important for JSS performance.
- **Pile-up:** best to mitigate a little in each step (cluster/track reco, constituent-, jet-level, *etc.*).
- **Heavy ions:** centrality quantifies the fraction of participating nucleons in a collision.
- **Jet Tagging:** many approaches being used simultaneously for different purposes: important to also understand older approaches, context in which they are used.
- **Unfolding:** different approaches are used, but we typically ask similar questions to assess performance.

Let's BOOST!

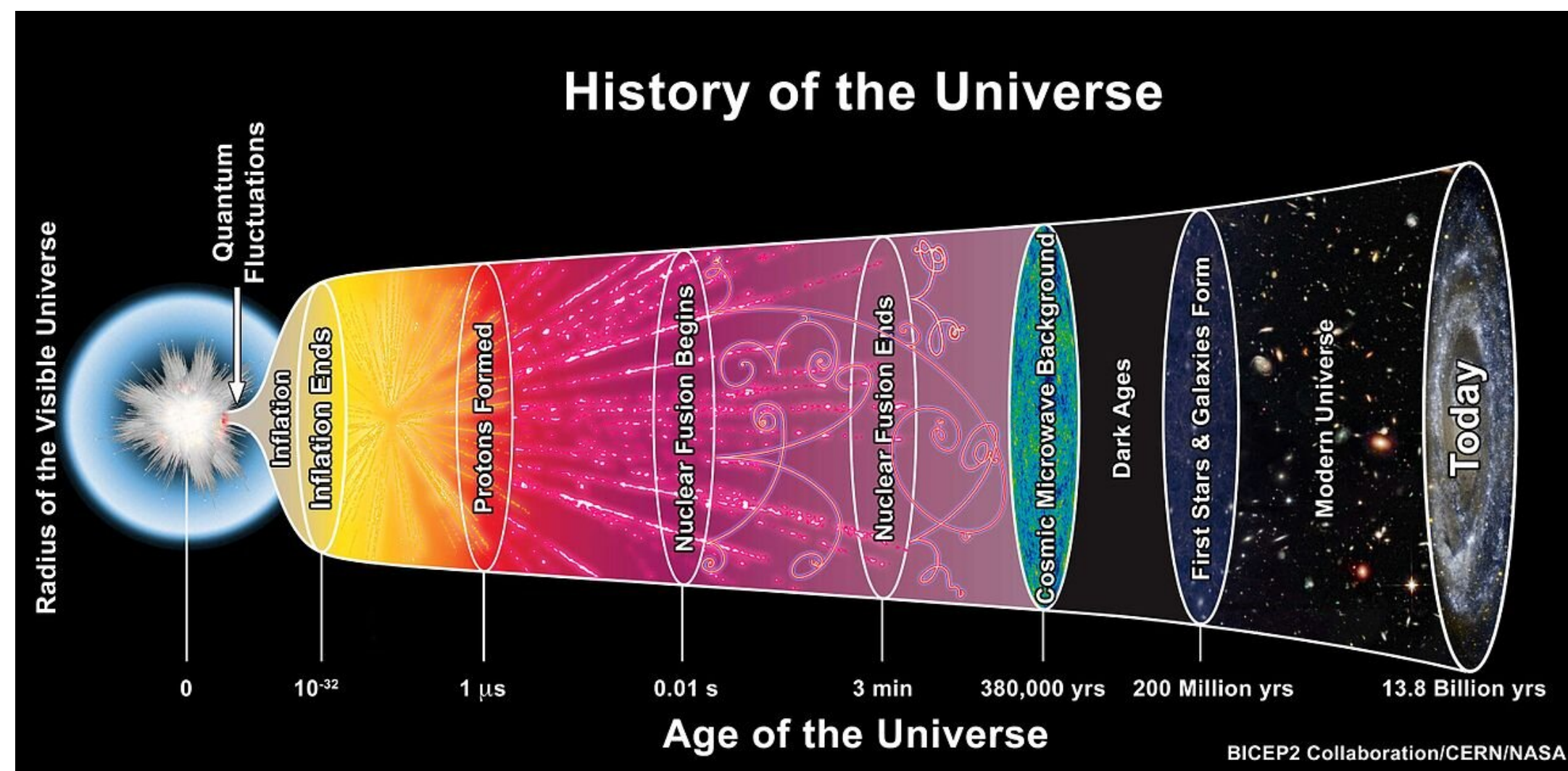
Thanks for listening!



BROWN

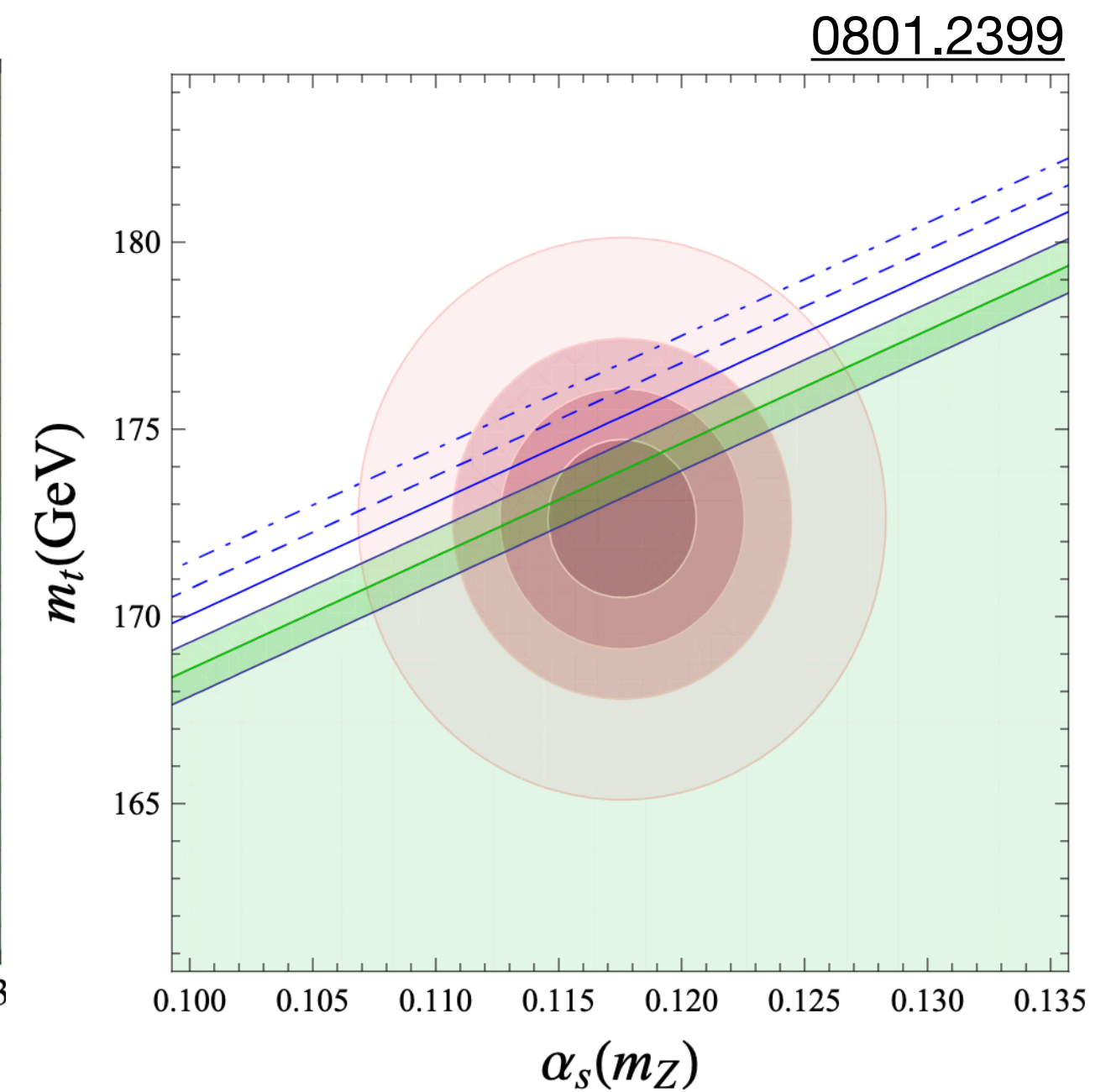
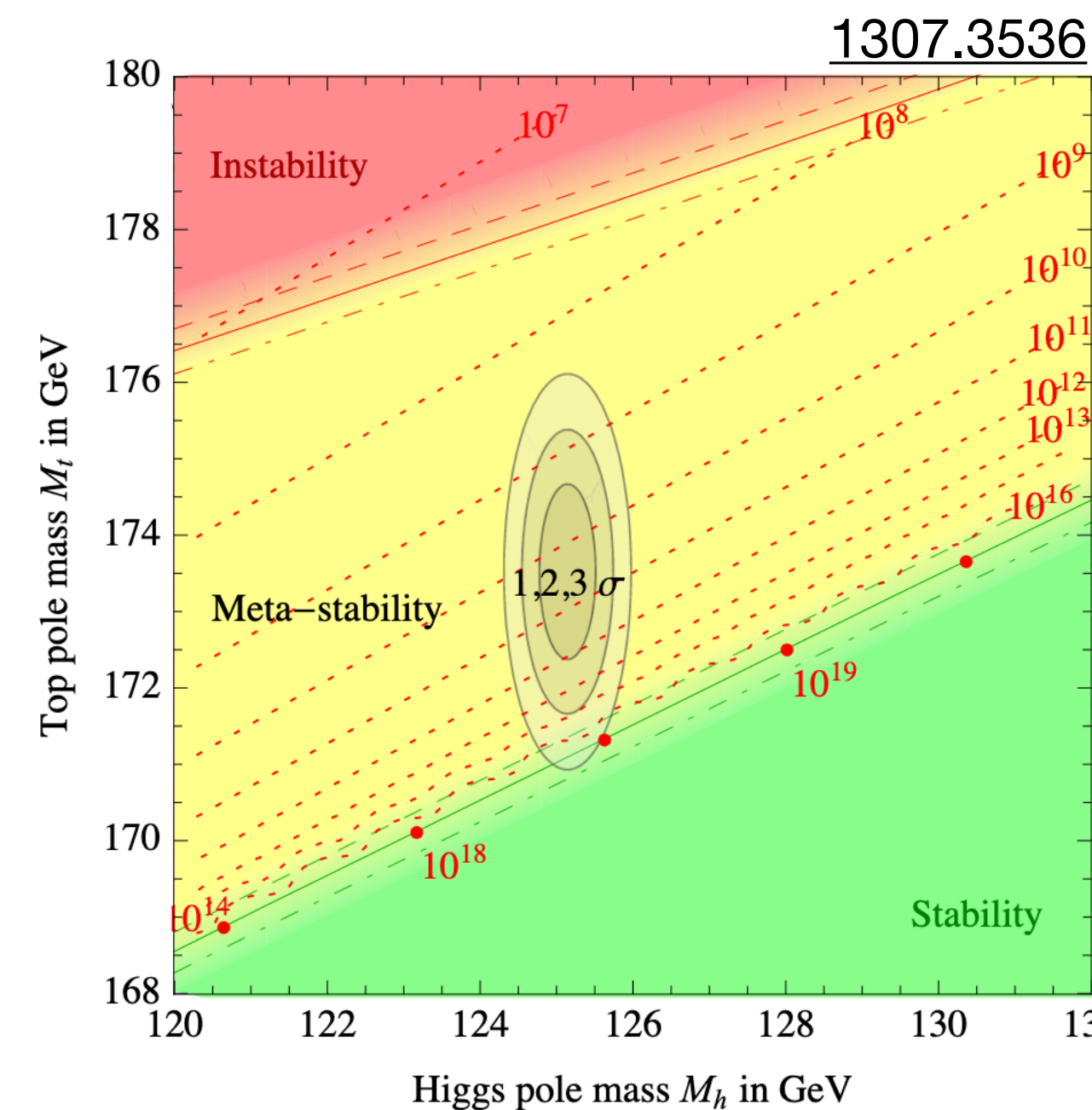
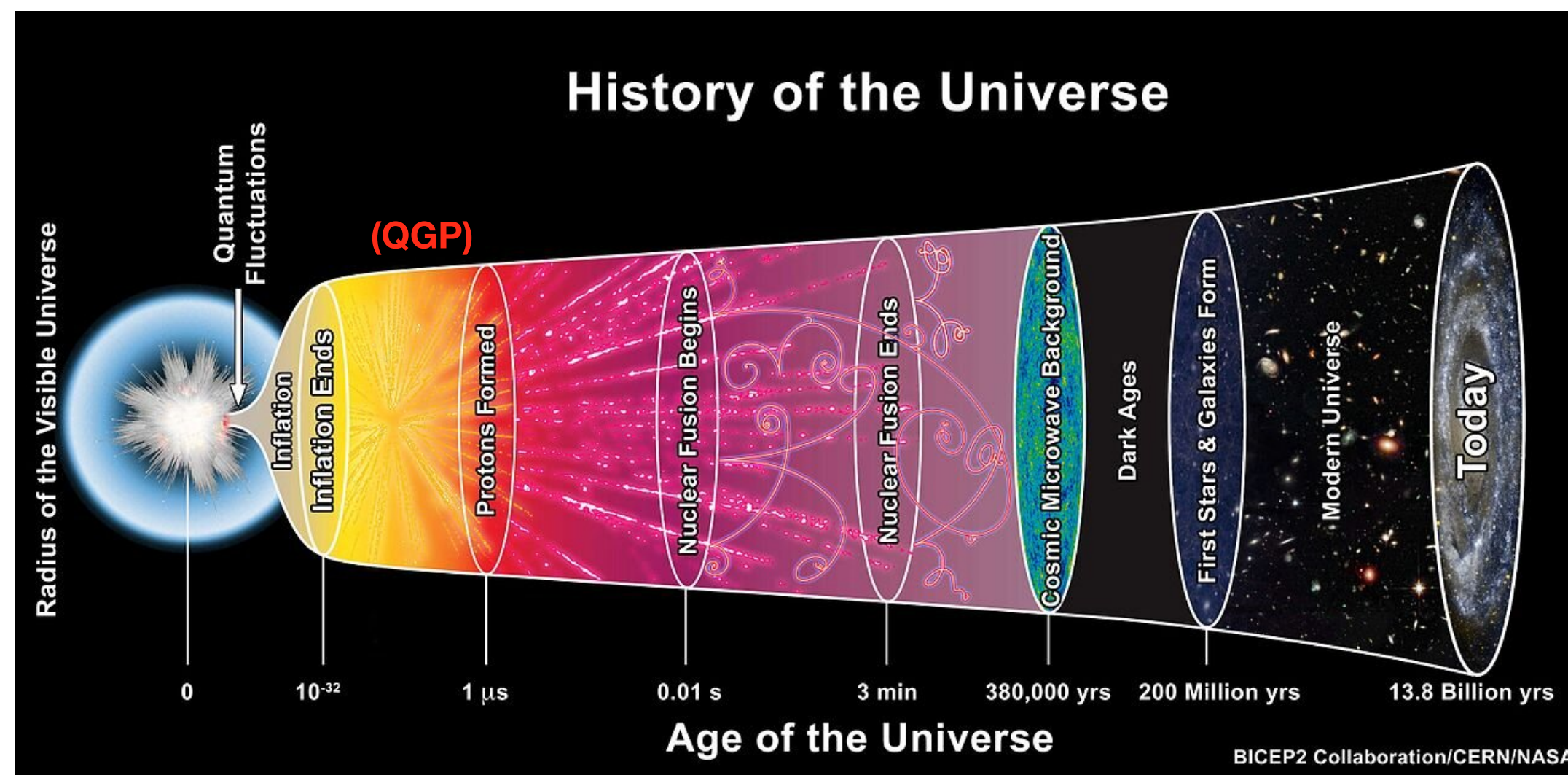
Auxiliary material

Particle Physicists learn about our universe by studying the properties of particles.



What was the nature of our early universe?

Particle Physicists learn about our universe by studying the properties of particles.

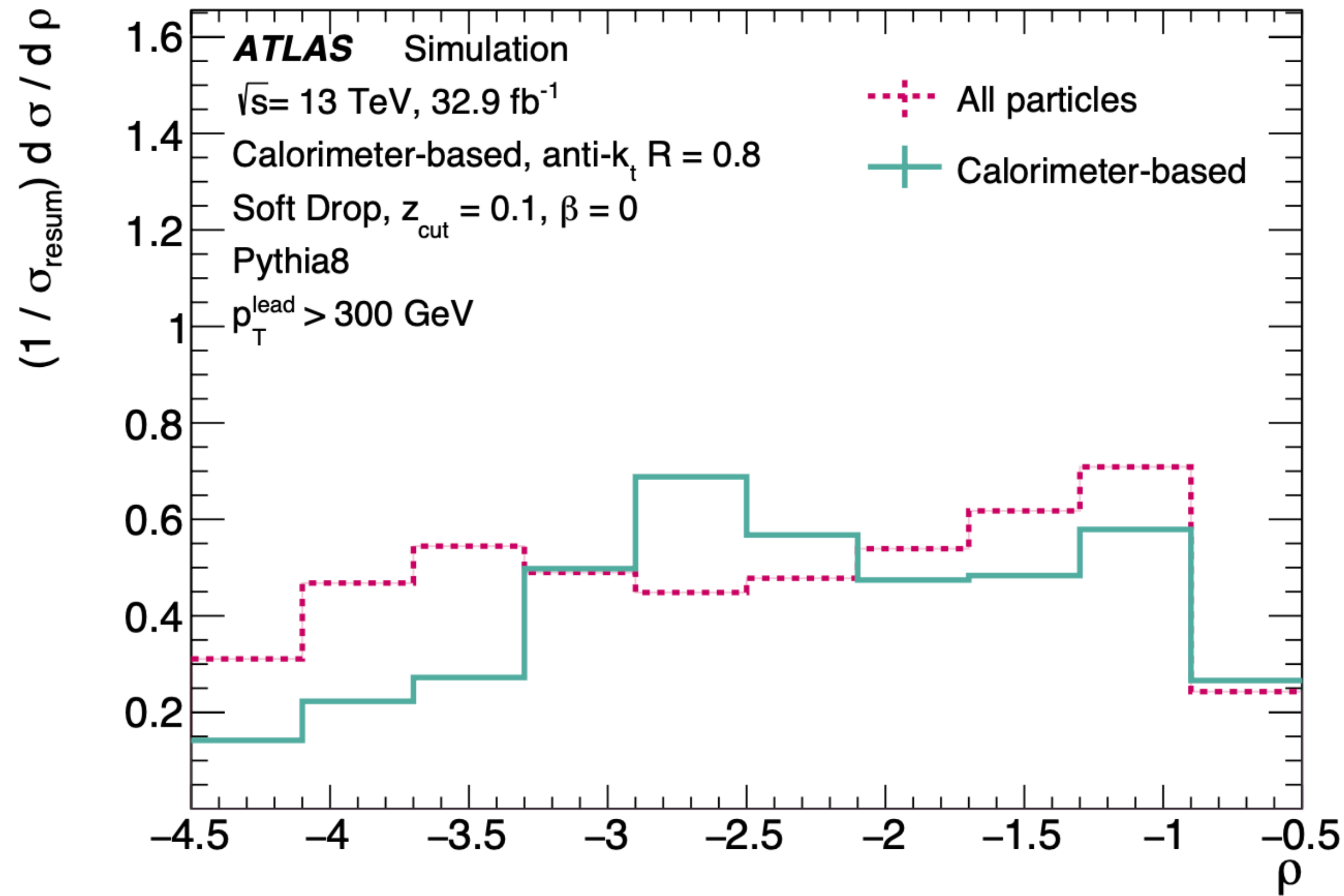


What was the nature of our early universe?

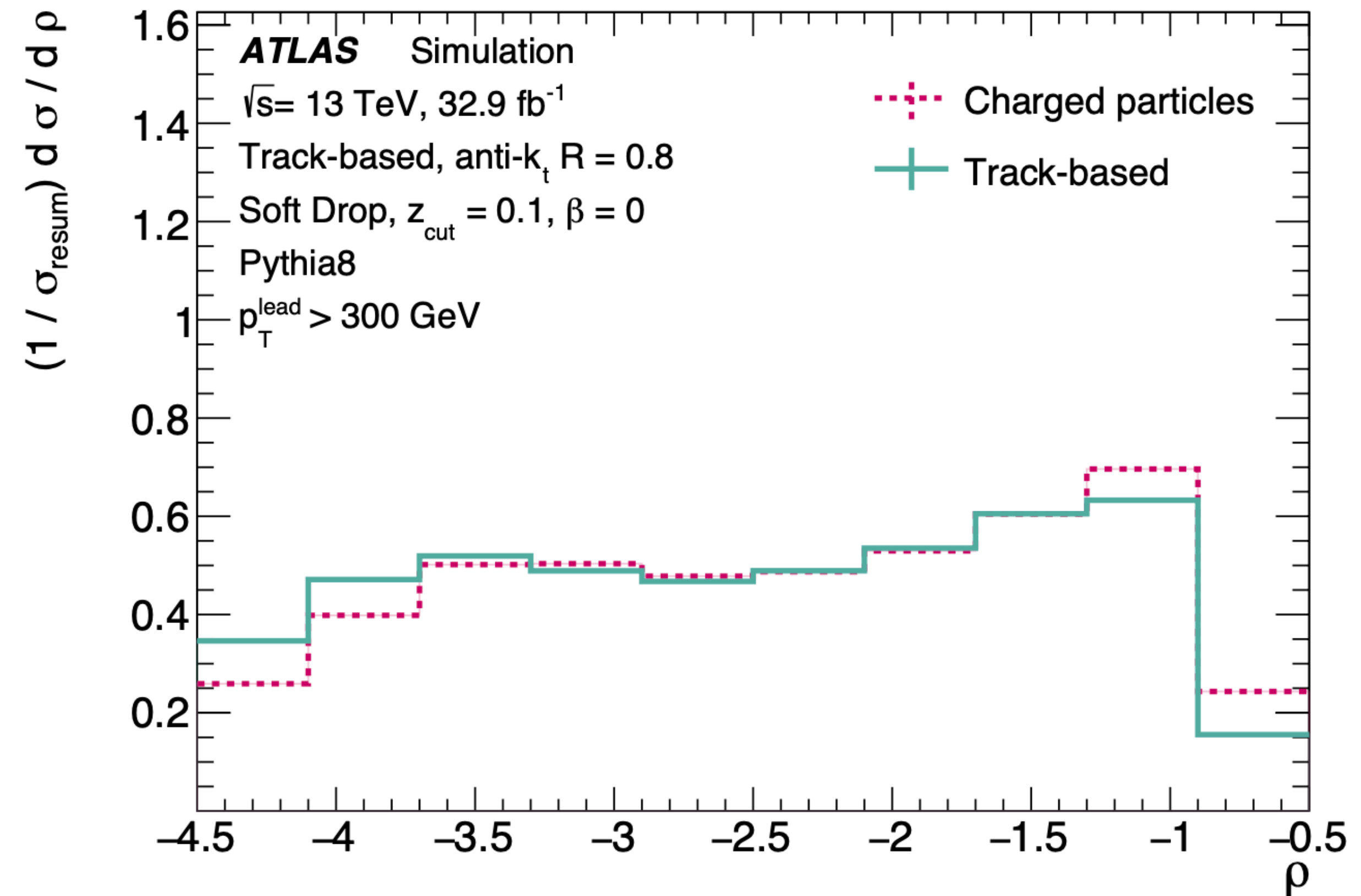
... and what is its ultimate fate?

Soft-drop observables: calo vs. track before unfolding

ATLAS, *Phys. Rev. D* 101, 052007 (2020), *Phys. Rev. Lett.* 121, 092001 (2018)



Calo-based



Track-based

Aside: CMS JSS Angularities Measurement

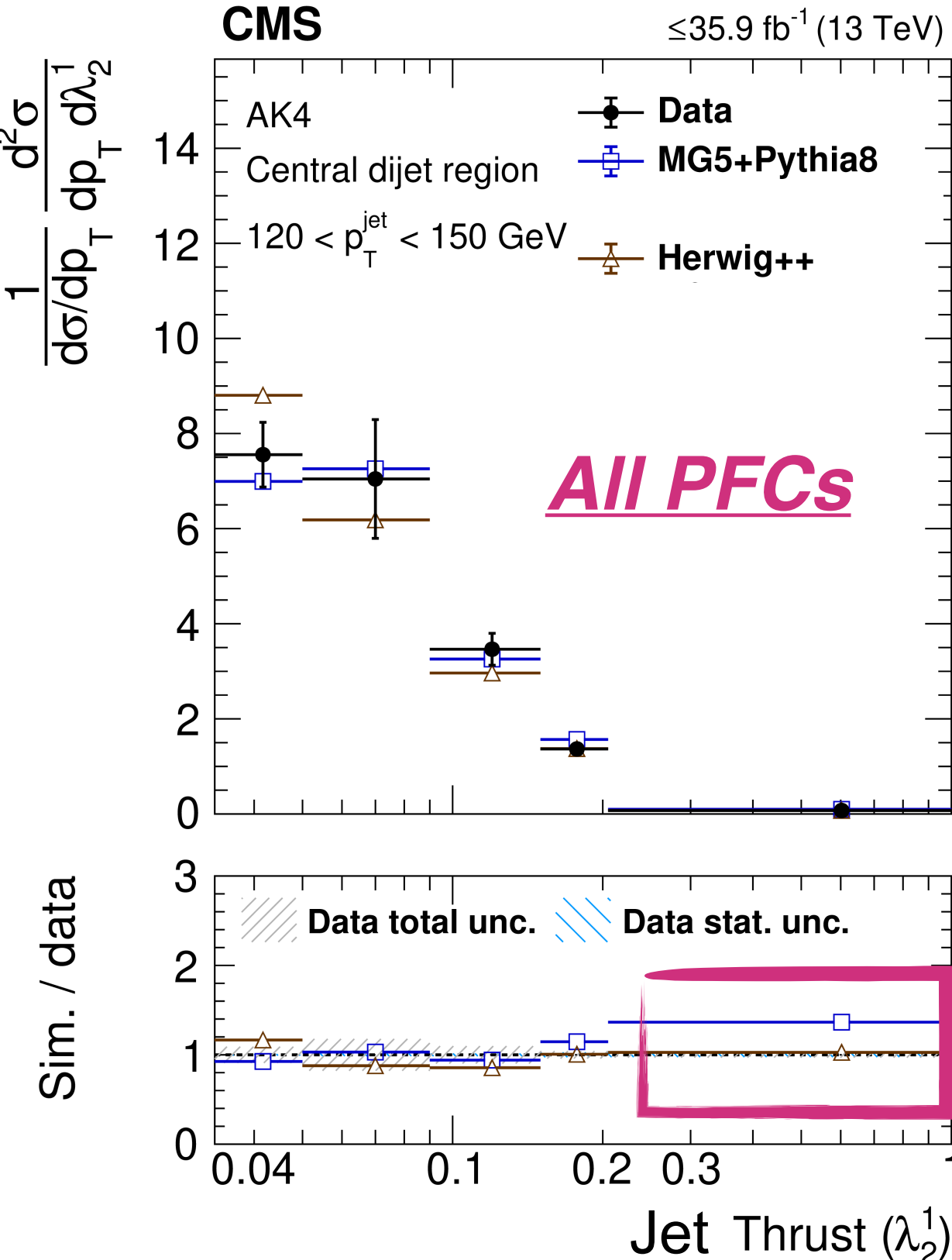
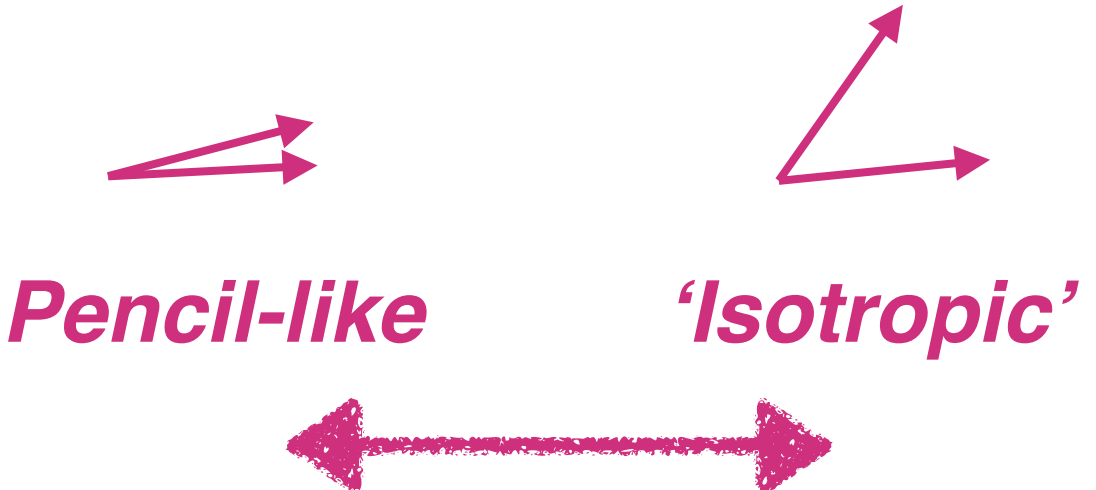
CMS, JHEP 01 (2022) 188

JSS Angularities

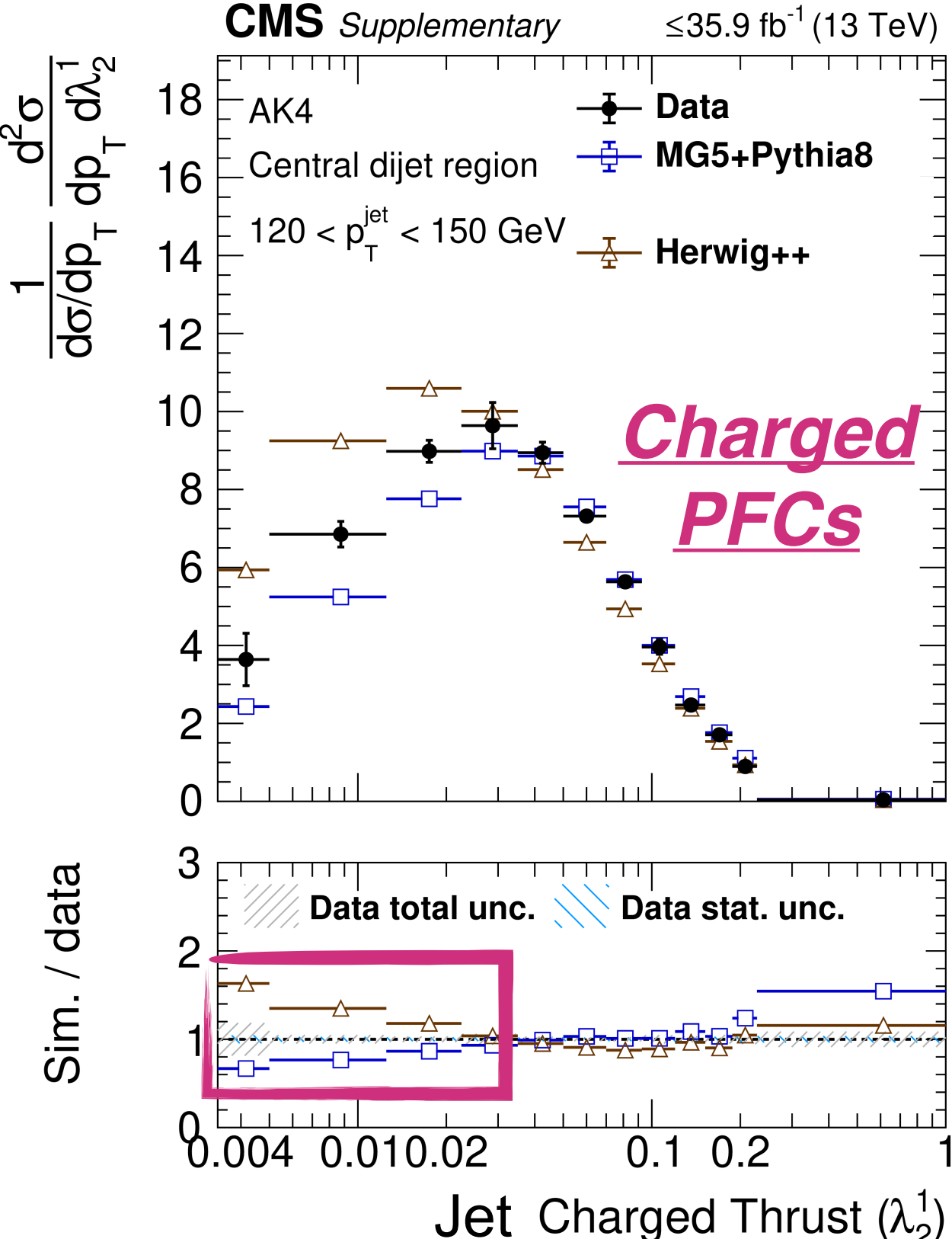
Products of relative constituent energies and angles, varying weight of each component

$$\lambda_{\beta}^{\kappa} = \sum_{i \in \text{jet}} z_i^{\kappa} \left(\frac{\Delta R_i}{R} \right)^{\beta}$$

- Both ATLAS and CMS have made comparisons of **charged+neutral & charged-only pictures**.
- Similar observations can be made using data from both collaborations
- Perhaps surprising, given CMS's "particle-flow" reconstruction.

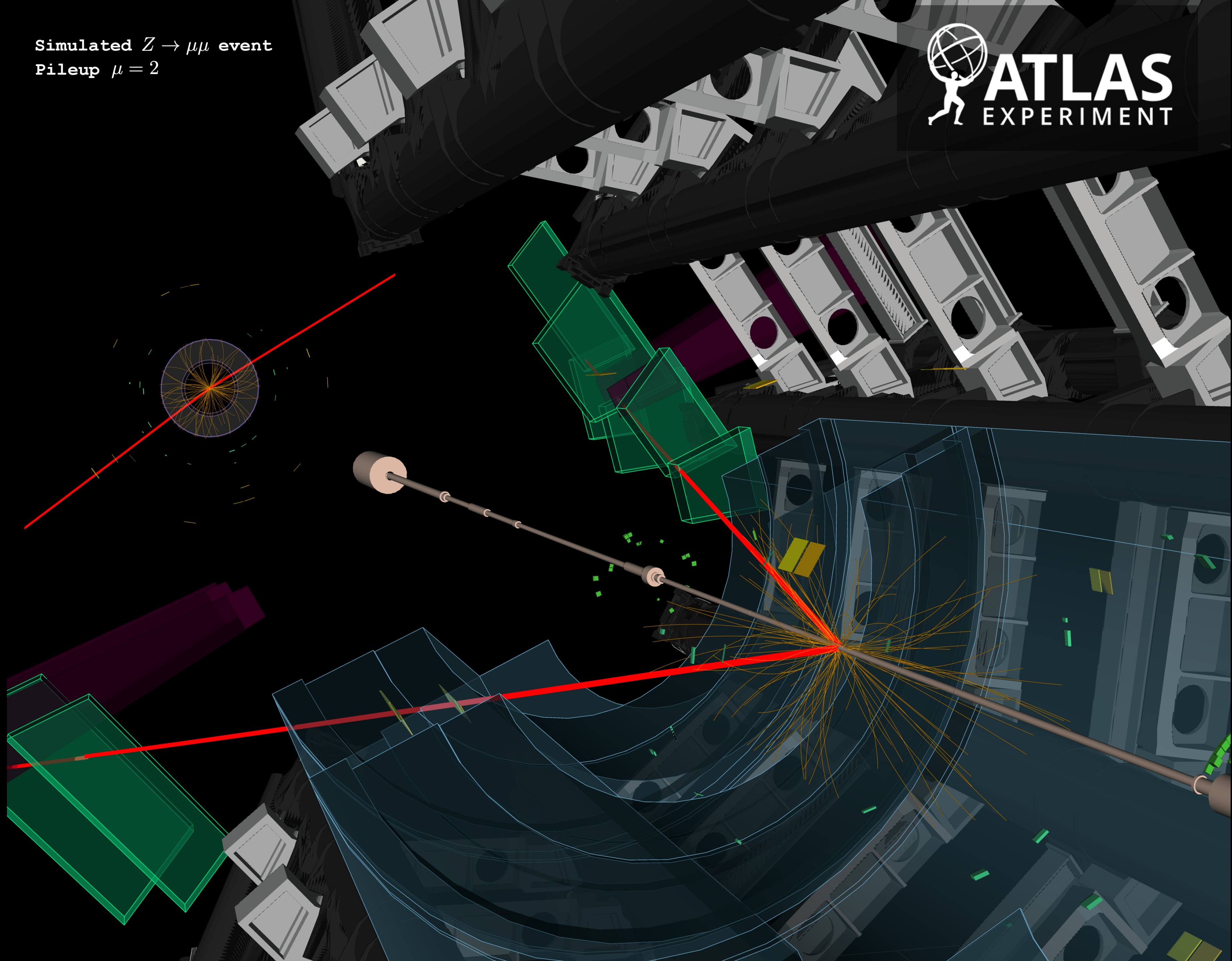


Similar levels of (dis)agreement!

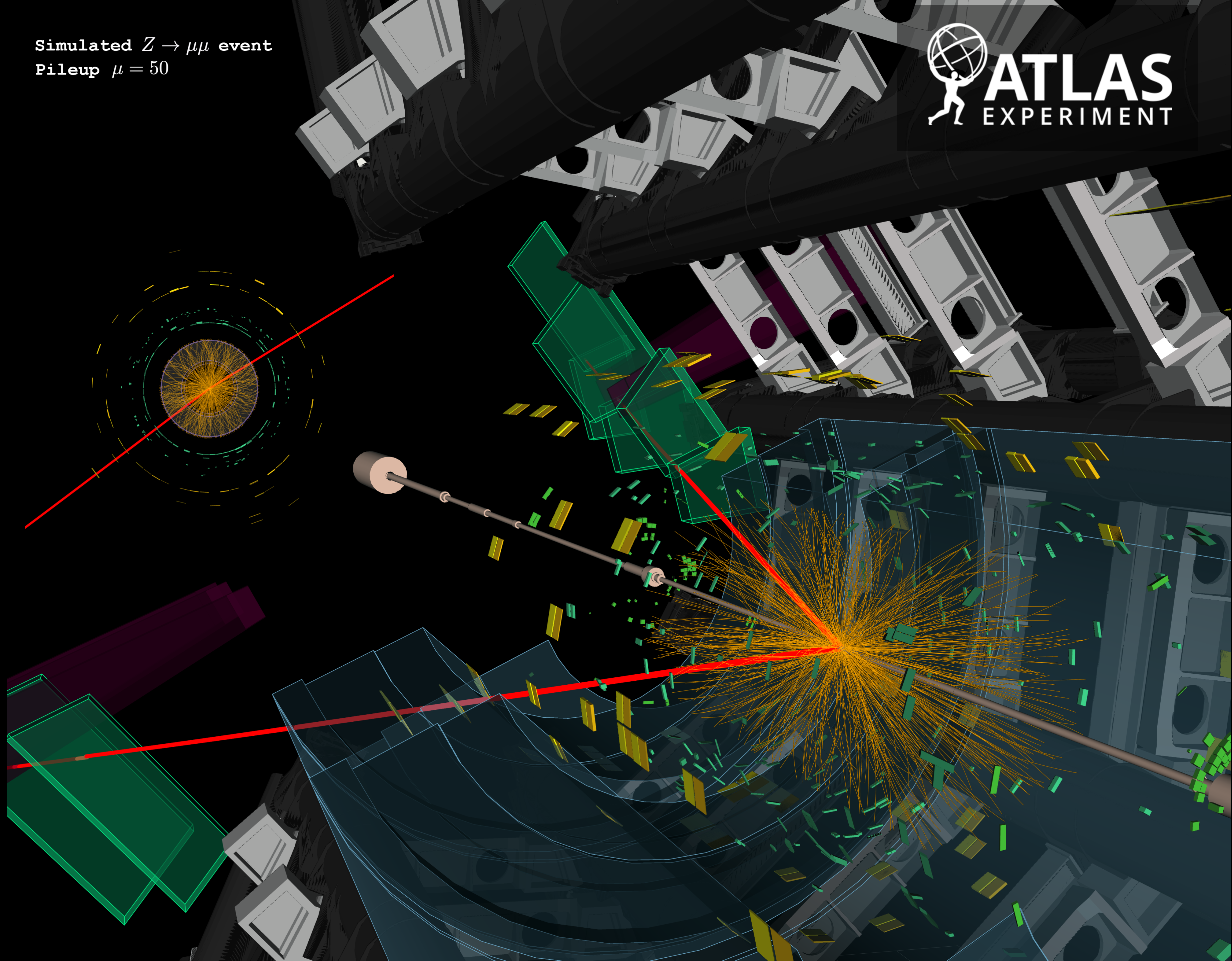


Charged picture has significantly more reach into collinear region!

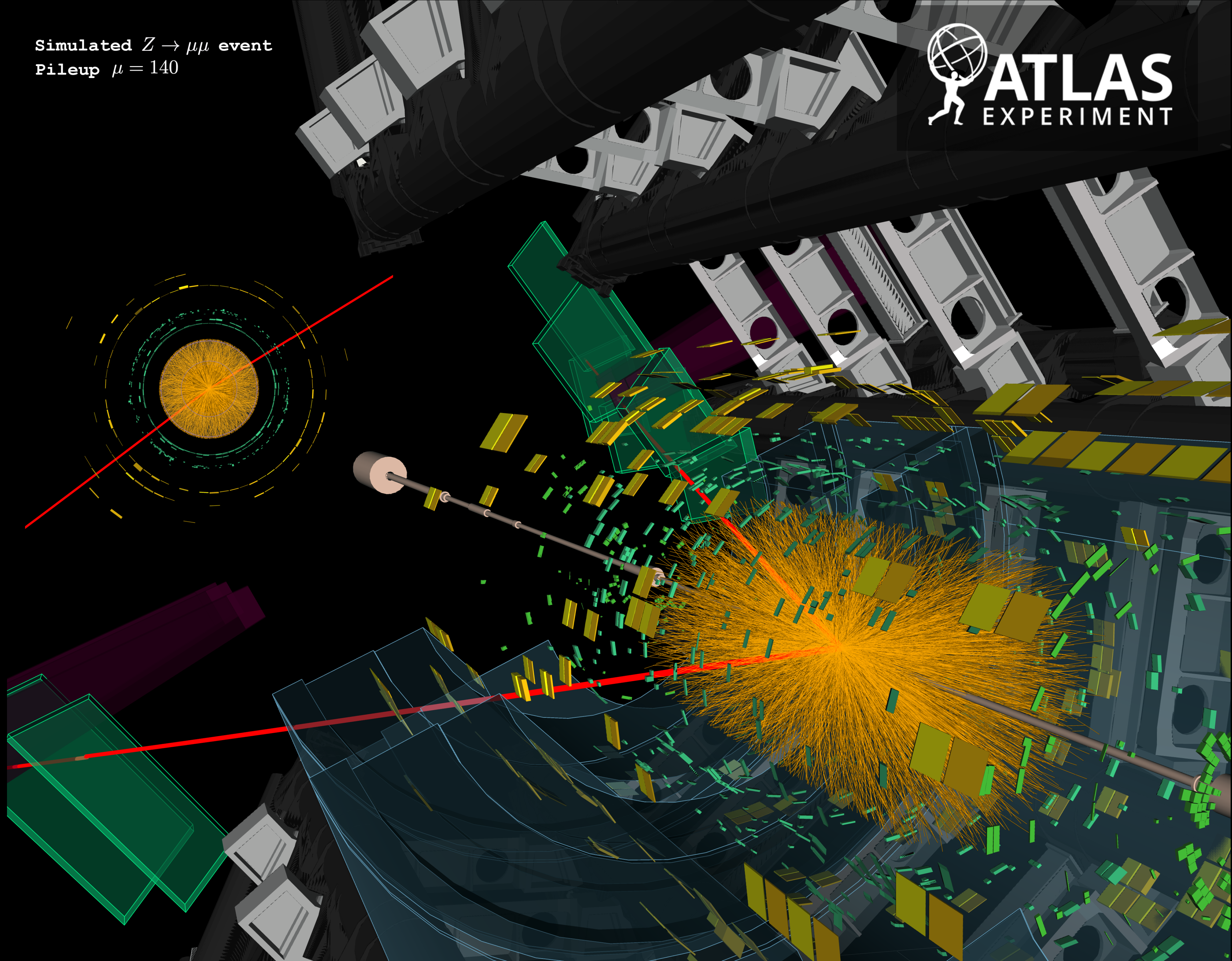
Simulated $Z \rightarrow \mu\mu$ event
Pileup $\mu = 2$



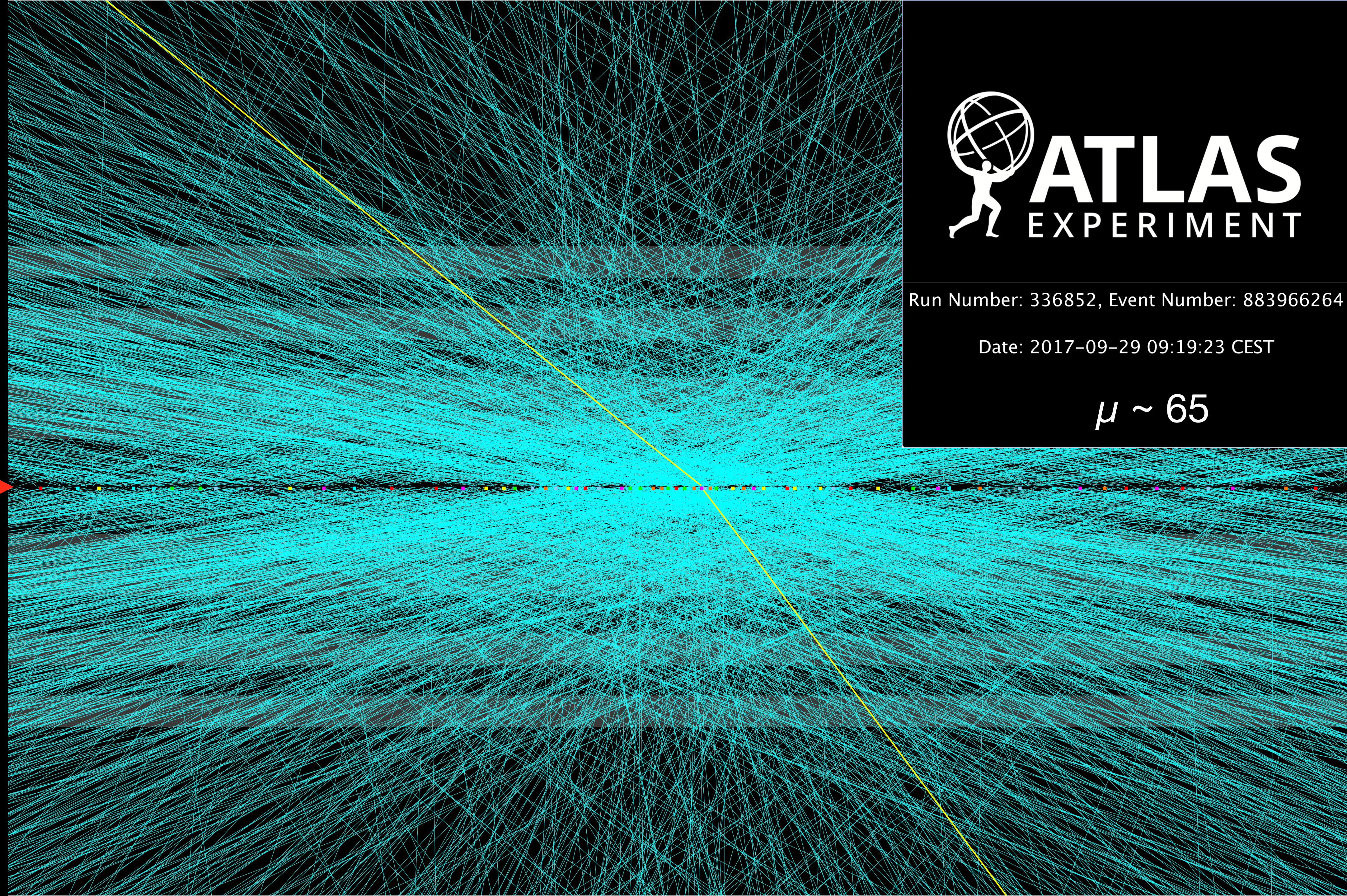
Simulated $Z \rightarrow \mu\mu$ event
Pileup $\mu = 50$



Simulated $Z \rightarrow \mu\mu$ event
Pileup $\mu = 140$



beam



ATLAS
EXPERIMENT

Run Number: 336852, Event Number: 883966264

Date: 2017-09-29 09:19:23 CEST

$\mu \sim 65$

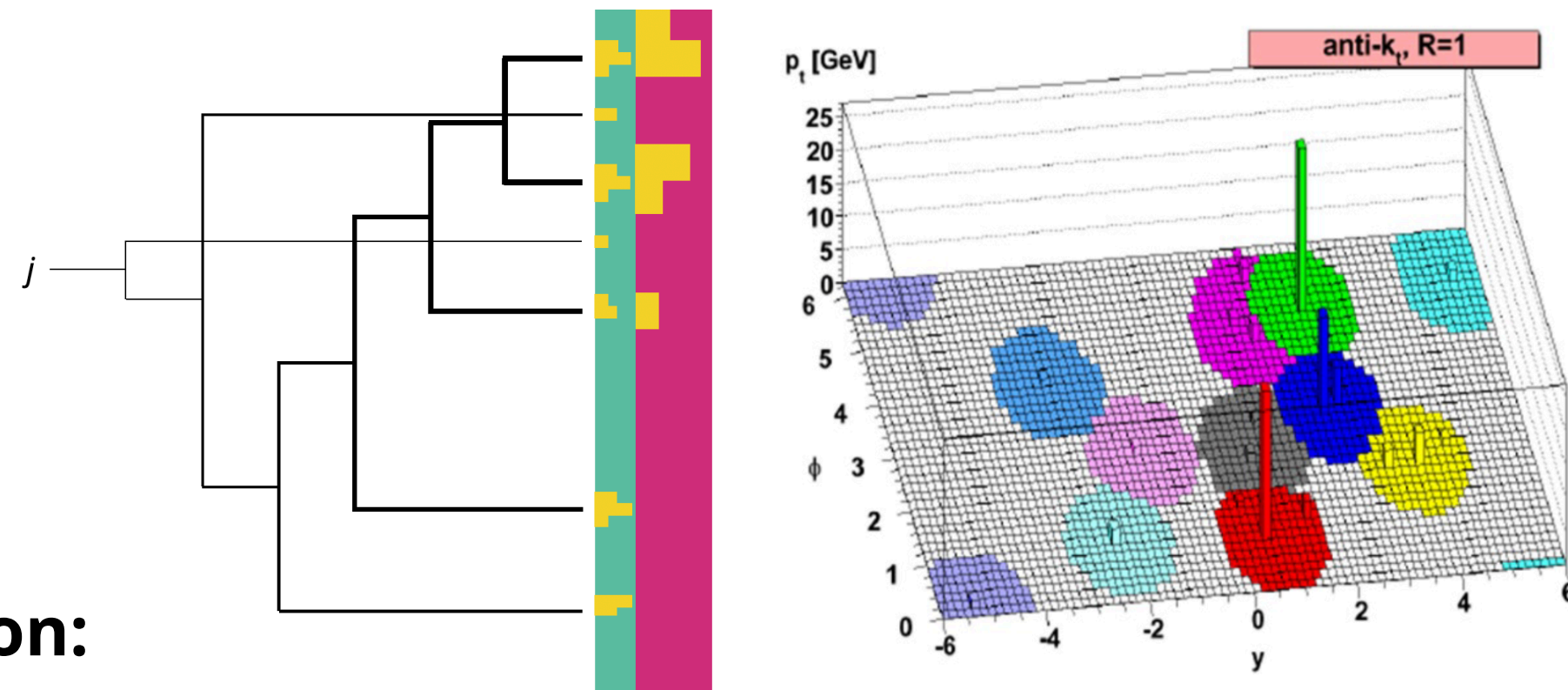
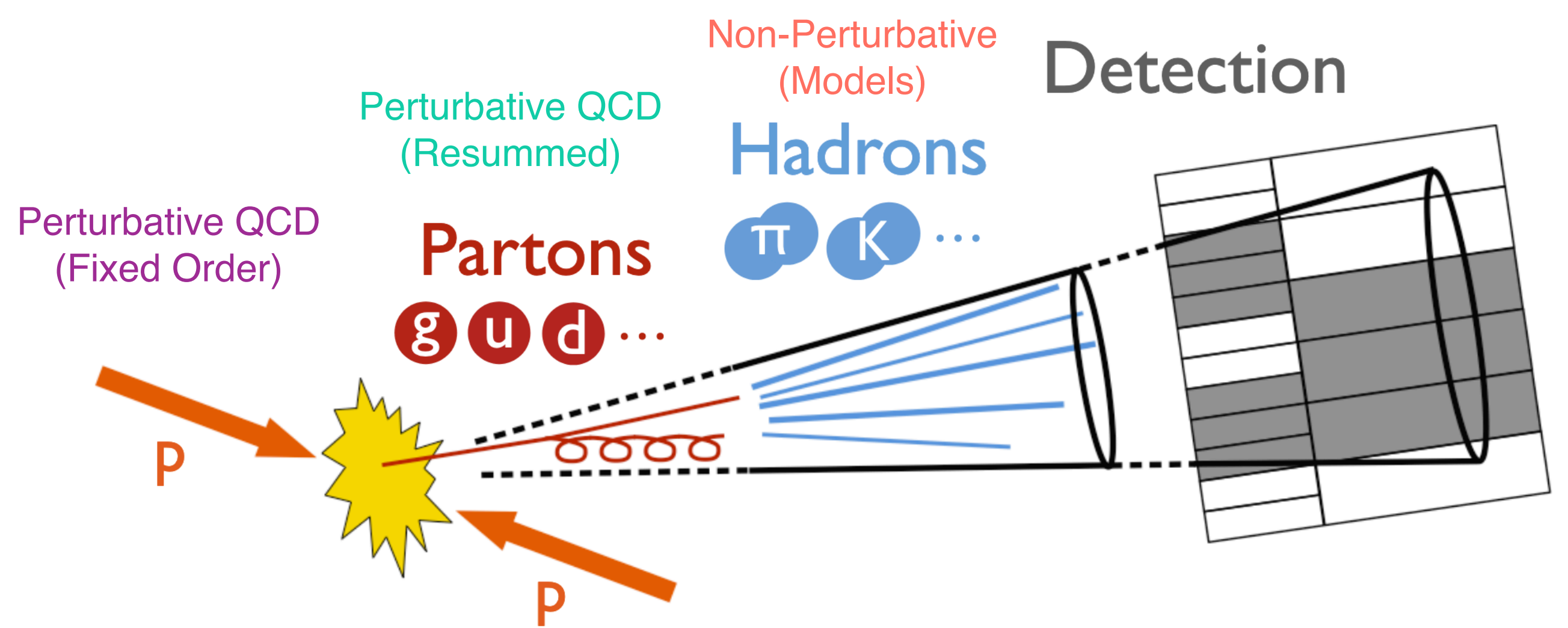


“What is a jet?”

Starting place for jet physics: Salam [0906.1833](#)

- **Jets are formed when high-energy quarks and gluons are produced in LHC collisions.**
 - Interpreted as a proxy for parton momentum.
- **Jets are complex:**
 - Multi-scale
 - Large area
 - Composite objects
 - Diverse signals:
 - Calorimetry
 - Charged-particle tracks
- Jets do not exist without their **definition:**
 - Recursive recombination algorithms (anti-kt, C/A, ...)
 - Radius parameter (R)
 - Other choices ...
 - Grooming? Pile-up mitigation?

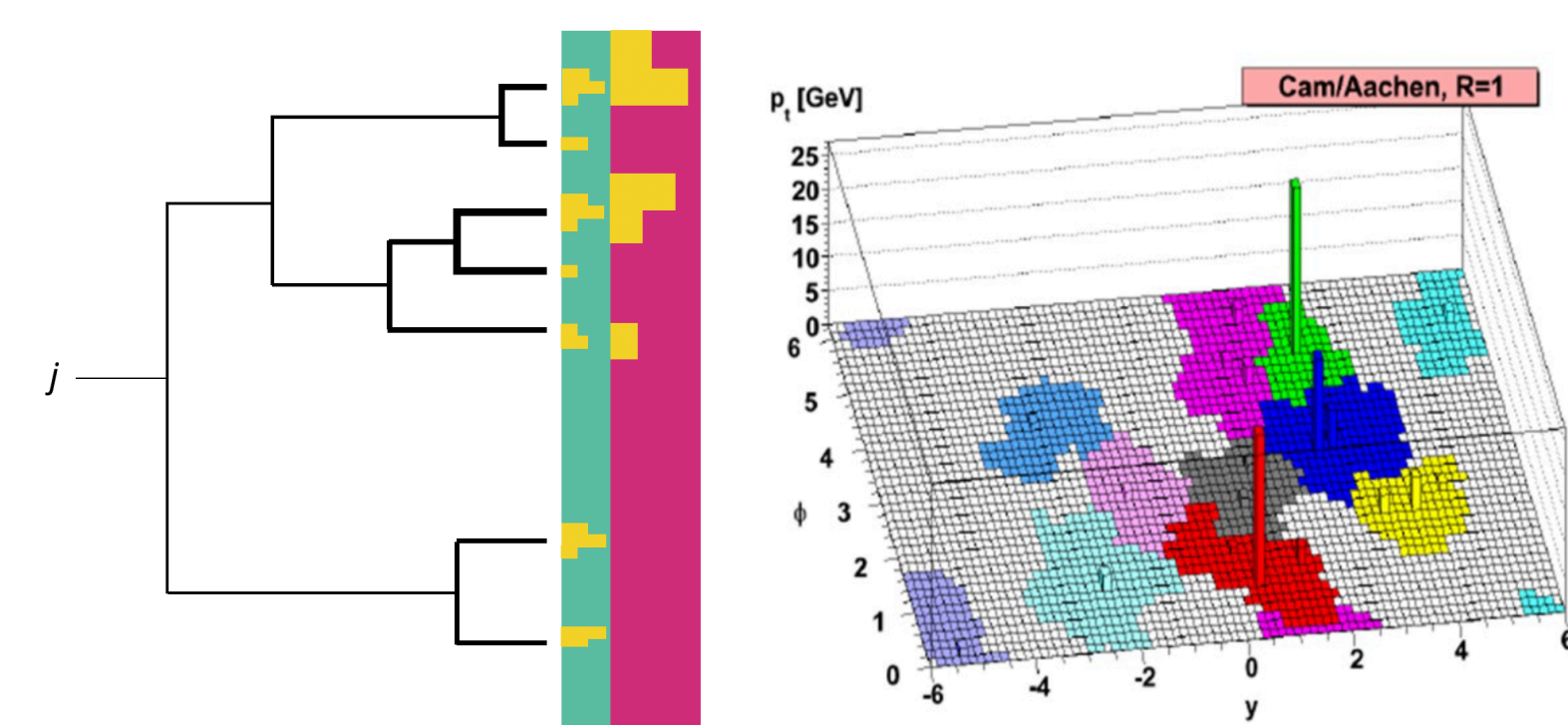
Common complaint: too many choices



Anti- k_t

“Soft things into hard things”

LHC daily driver



Cambridge / Aachen

“Angle-ordered”

Irregular shape: difficult to calibrate

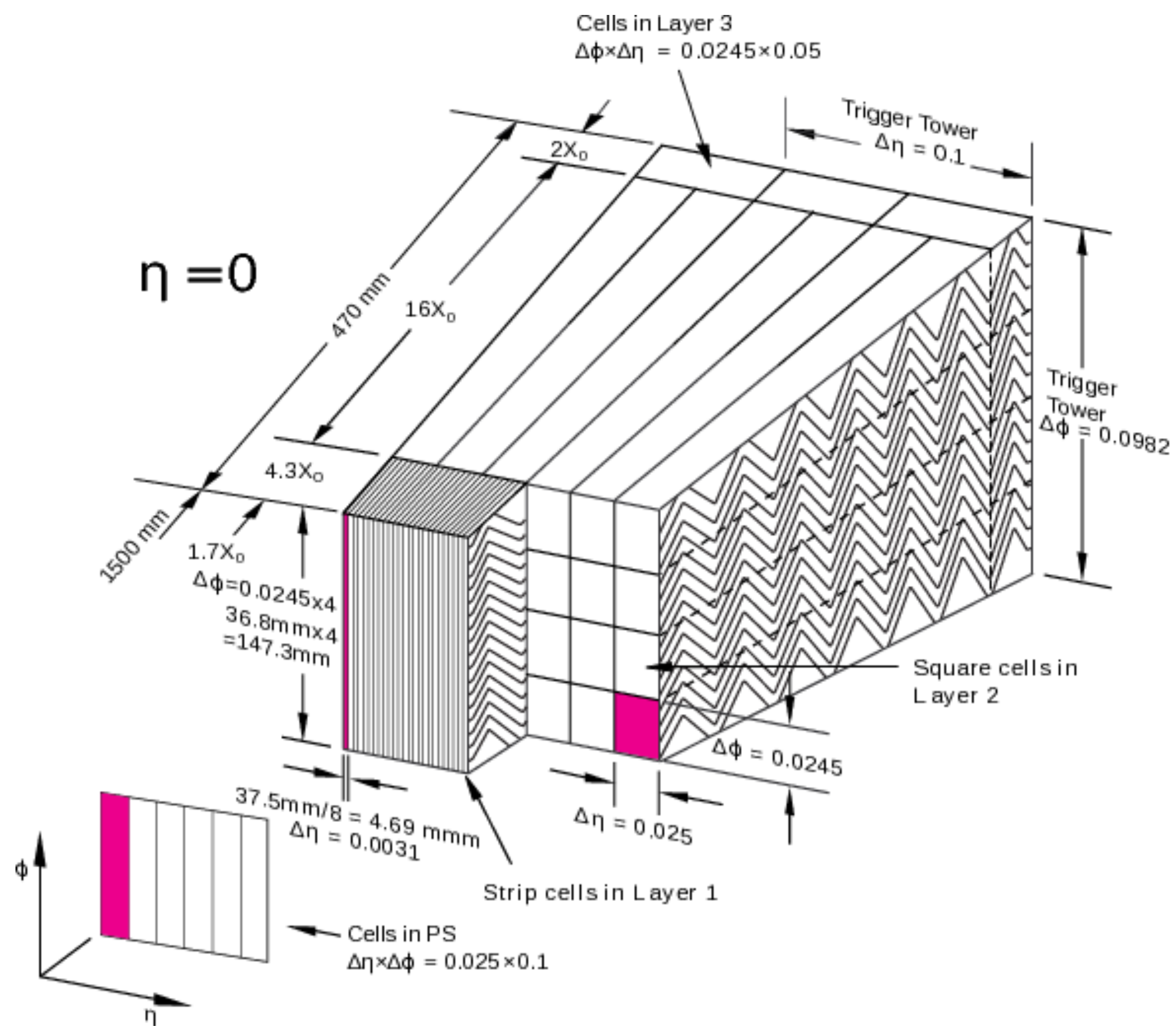
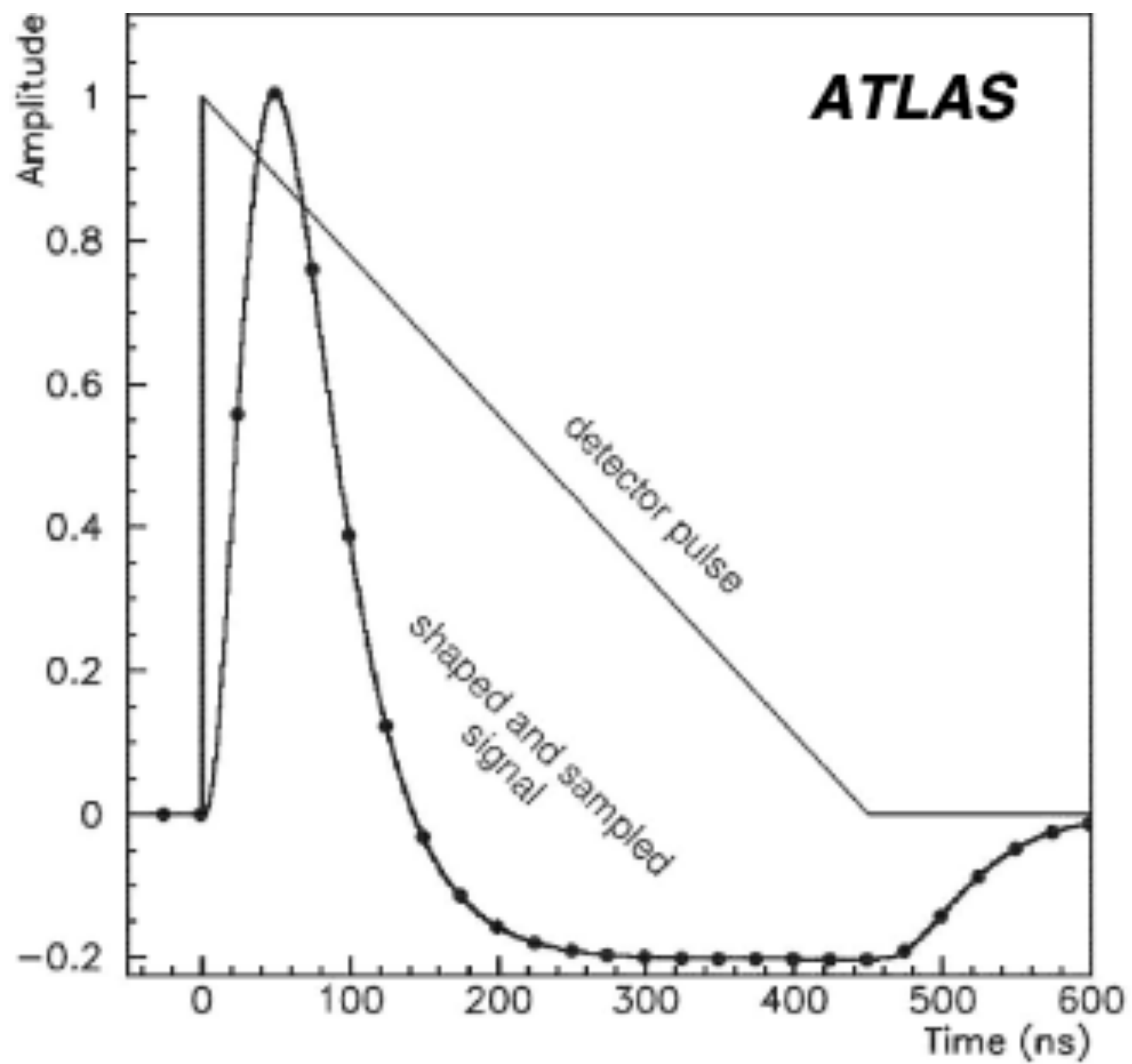
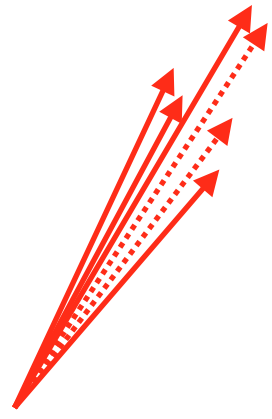
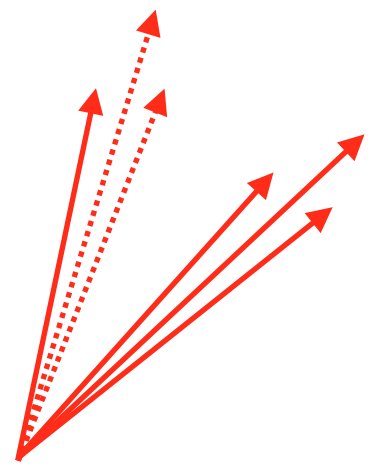


Table 1: The read-out granularity of the ATLAS calorimeter system [1], given in terms of $\Delta\eta \times \Delta\phi$ with the exception of the forward calorimeters, where it is given in linear measures $\Delta x \times \Delta y$, due to the non-pointing read-out geometry of the FCAL. For comparison, the FCAL granularity is approximately $\Delta\eta \times \Delta\phi = 0.15 \times 0.15 (0.3 \times 0.3)$ at $\eta = 3.5 (4.5)$. The total number of read-out cells, including both ends of the calorimeter system, with (without) pre-samplers is 187 652 (178 308).

Calorimeter	Module	Sampling (S_{calo})	N_{cells}	η -coverage	$\Delta\eta \times \Delta\phi$	
<i>Electromagnetic calorimeters</i>	EMB		109 568	$ \eta < 1.52$		
		PreSamplerB	7 808	$ \eta < 1.52$	$0.025 \times \pi/32$	
		EMB1		$ \eta < 1.4$	$0.025/8 \times \pi/32$	
	EMB2			$1.4 < \eta < 1.475$	$0.025 \times \pi/128$	
				$ \eta < 1.4$	$0.025 \times \pi/128$	
				$1.4 < \eta < 1.475$	$0.075 \times \pi/128$	
	EMB3			$ \eta < 1.35$	$0.050 \times \pi/128$	
		EMEC		63 744	$1.375 < \eta < 3.2$	
			PreSamplerE	1 536	$1.5 < \eta < 1.8$	$0.025 \times \pi/32$
	EME1				$1.375 < \eta < 1.425$	$0.050 \times \pi/32$
				$1.425 < \eta < 1.5$	$0.025 \times \pi/32$	
				$1.5 < \eta < 1.8$	$0.025/8 \times \pi/32$	
				$1.8 < \eta < 2.0$	$0.025/6 \times \pi/32$	
				$2.0 < \eta < 2.4$	$0.025/4 \times \pi/32$	
				$2.4 < \eta < 2.5$	$0.025 \times \pi/32$	
				$2.5 < \eta < 3.2$	$0.1 \times \pi/32$	
		EME2			$1.375 < \eta < 1.425$	$0.050 \times \pi/128$
					$1.425 < \eta < 2.5$	$0.025 \times \pi/128$
				$2.5 < \eta < 3.2$	$0.1 \times \pi/128$	
EME3			$1.5 < \eta < 2.5$	$0.050 \times \pi/128$		
	<i>Hadronic calorimeters</i>	Tile (barrel)	2 880	$ \eta < 1$		
						$0.1 \times \pi/32$
					$0.2 \times \pi/32$	
Tile (extended barrel)	2 304	$0.8 < \eta < 1.7$				
				$0.1 \times \pi/32$		
HEC	HEC0/1/2/3	5 632	$1.5 < \eta < 3.2$			
					$0.1 \times \pi/32$	
				$2.5 < \eta < 3.2$	$0.2 \times \pi/16$	
<i>Forward calorimeters</i>	FCAL		3 524	$3.1 < \eta < 4.9$	$\Delta x \times \Delta y$	
		FCAL0			$3.1 < \eta < 3.15$	$1.5 \text{ cm} \times 1.3 \text{ cm}$
					$3.15 < \eta < 4.3$	$3.0 \text{ cm} \times 2.6 \text{ cm}$
				$4.3 < \eta < 4.83$	$1.5 \text{ cm} \times 1.3 \text{ cm}$	
	FCAL1			$3.2 < \eta < 3.24$	$1.7 \text{ cm} \times 2.1 \text{ cm}$	
				$3.24 < \eta < 4.5$	$3.3 \text{ cm} \times 4.2 \text{ cm}$	
				$4.5 < \eta < 4.81$	$1.7 \text{ cm} \times 2.1 \text{ cm}$	
	FCAL2			$3.29 < \eta < 3.32$	$2.7 \text{ cm} \times 2.4 \text{ cm}$	
				$3.32 < \eta < 4.6$	$5.4 \text{ cm} \times 4.7 \text{ cm}$	
				$4.6 < \eta < 4.75$	$2.7 \text{ cm} \times 2.4 \text{ cm}$	

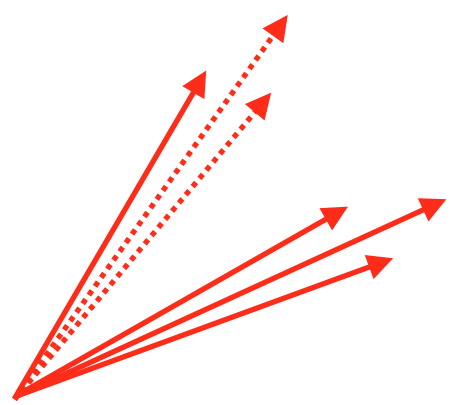


Quark? Gluon?



Higgs boson?

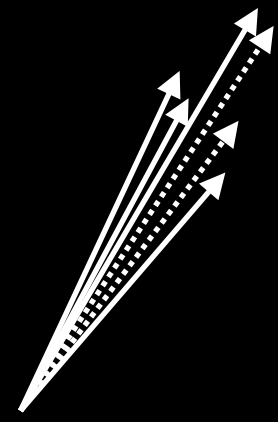
W/Z boson?



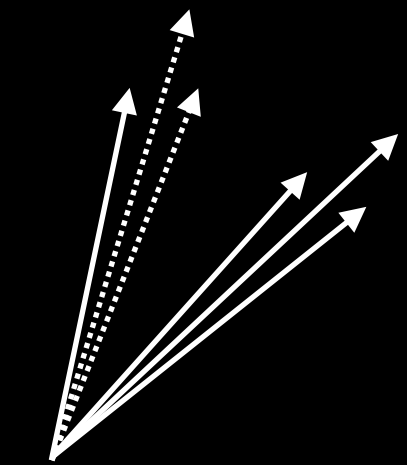
Top quark?

Tagging / Classification

Let's play a game!

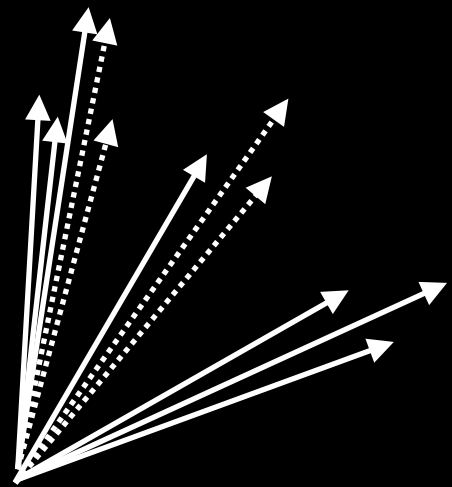


Quark? Gluon?

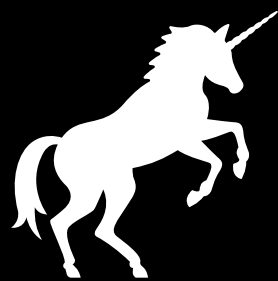


Higgs boson?

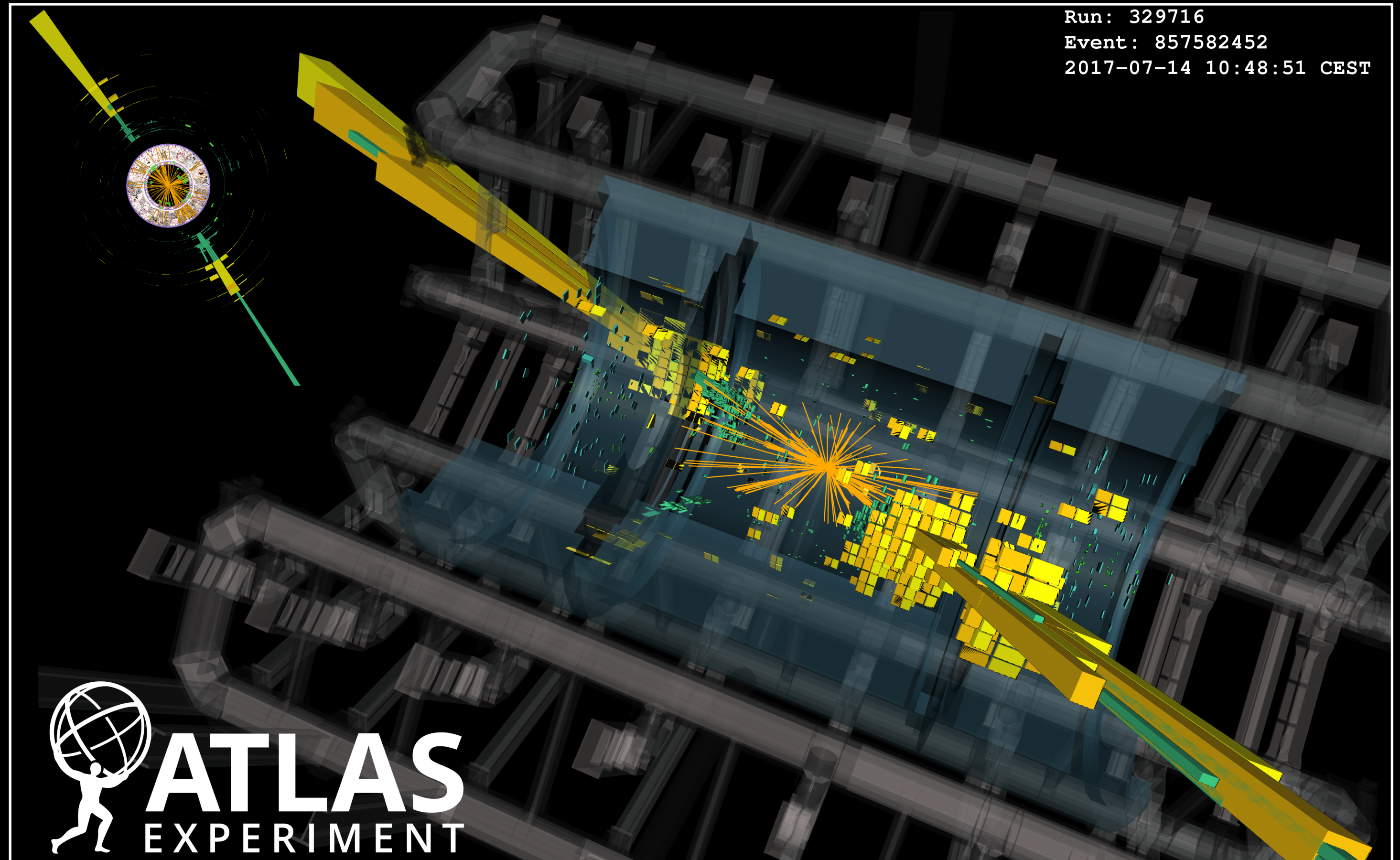
W/Z boson?

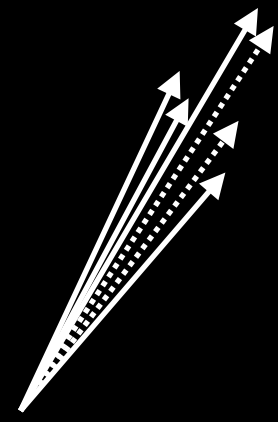


Top quark?

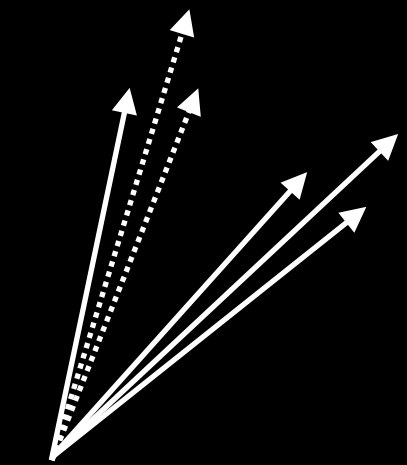


Something else?



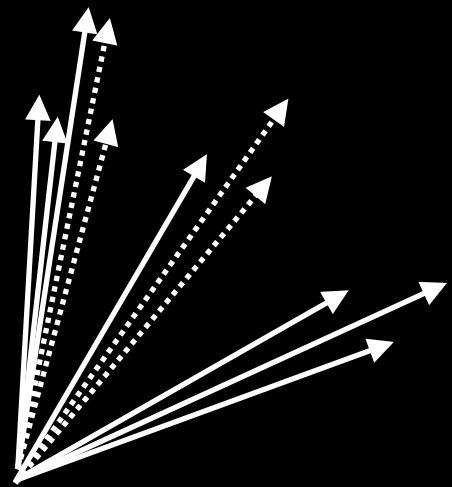


Quark? Gluon?

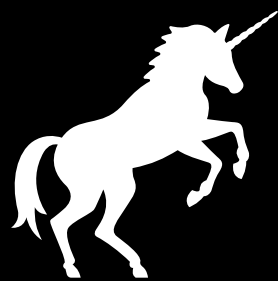


Higgs boson?

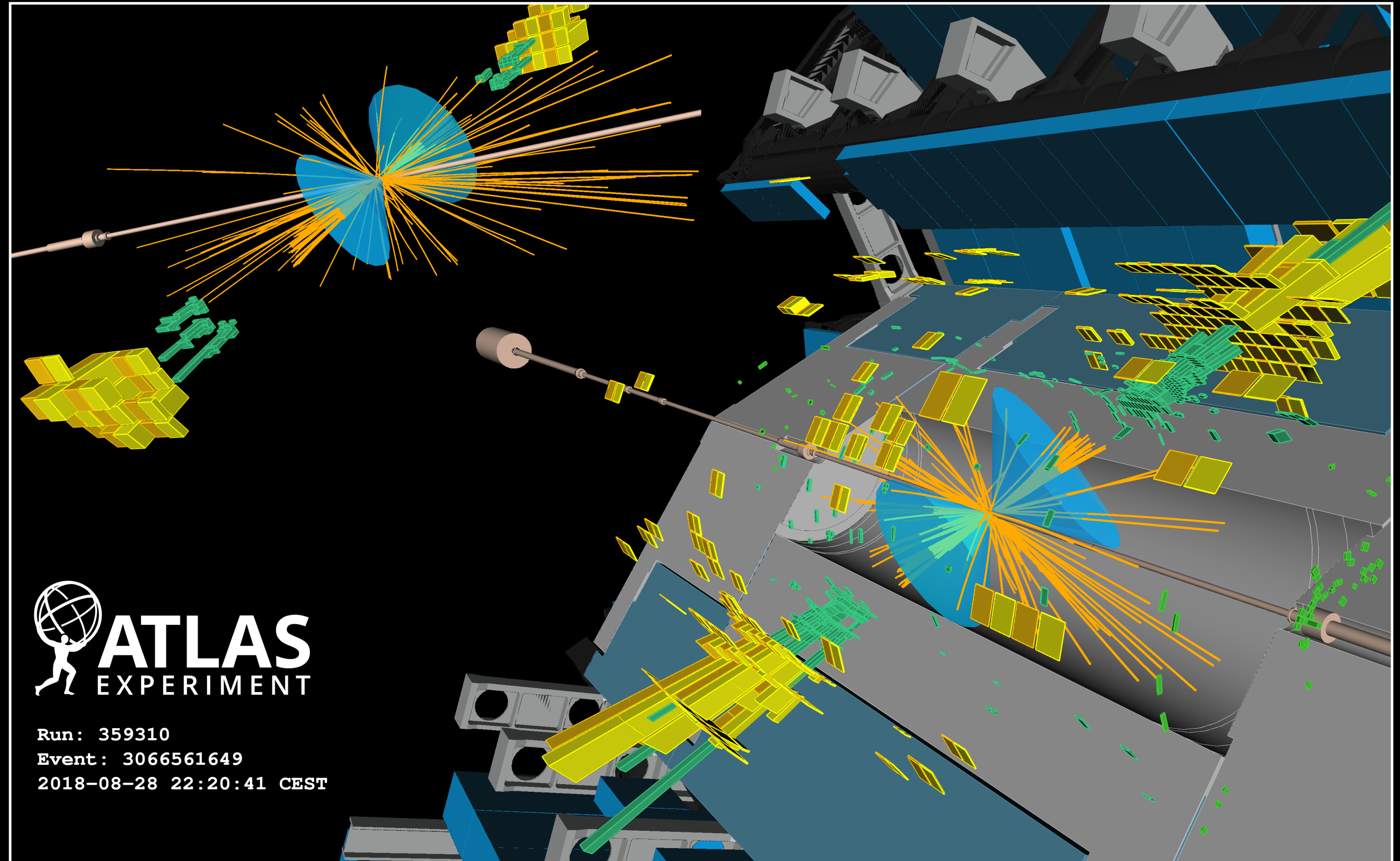
W/Z boson?

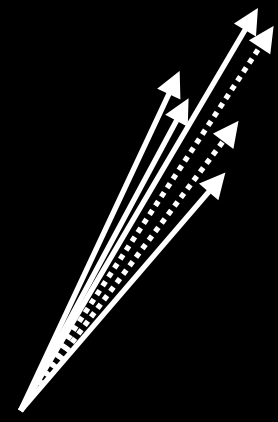


Top quark?

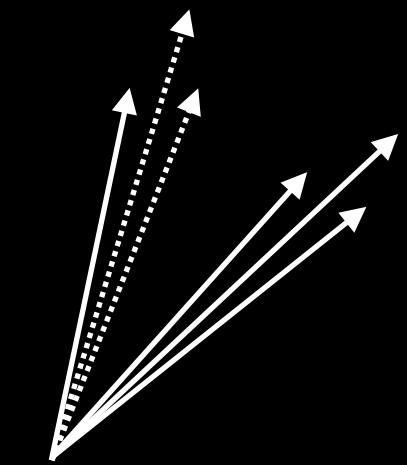


Something else?



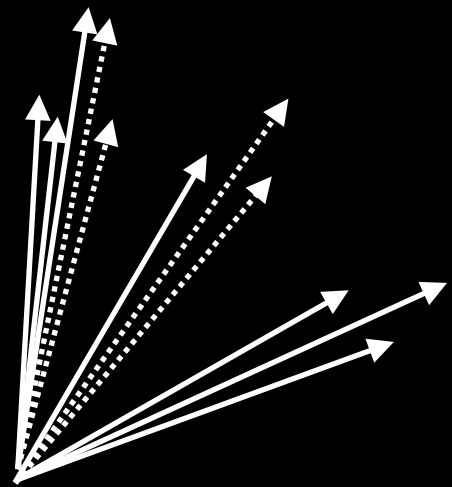


Quark? Gluon?

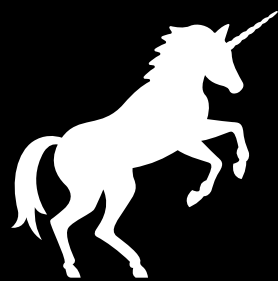


Higgs boson?

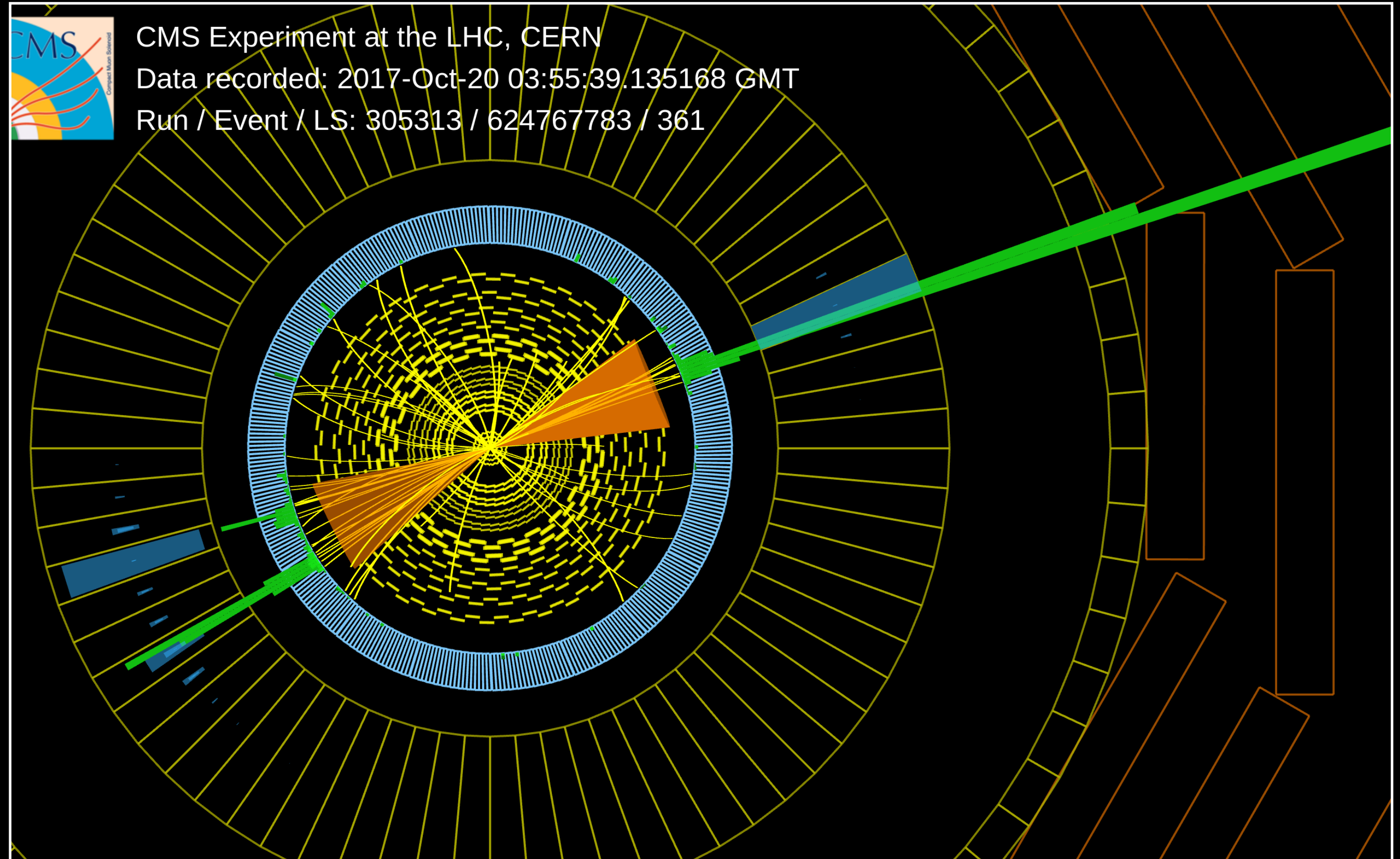
W/Z boson?

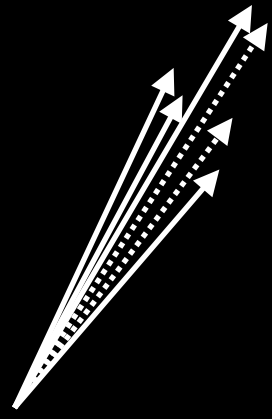


Top quark?

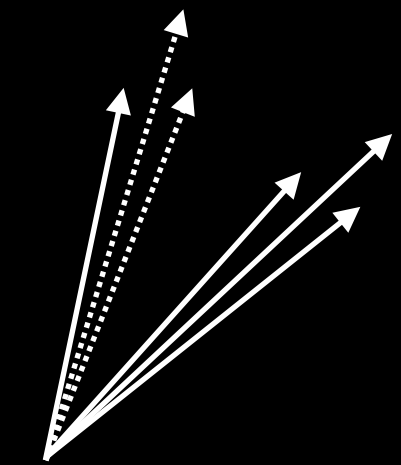


Something else?



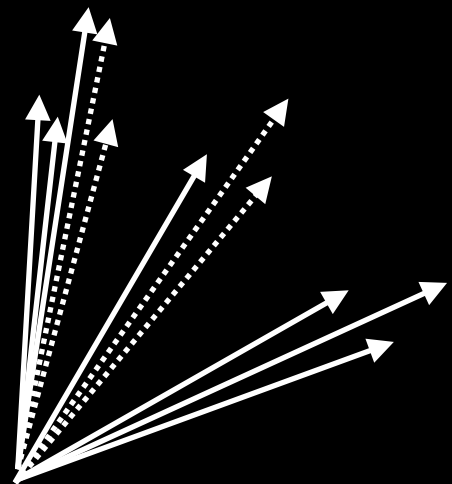


Quark? Gluon?

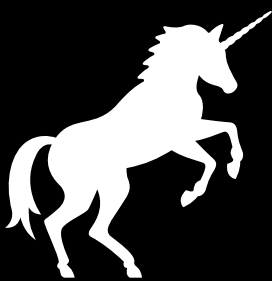


Higgs boson?

W/Z boson?

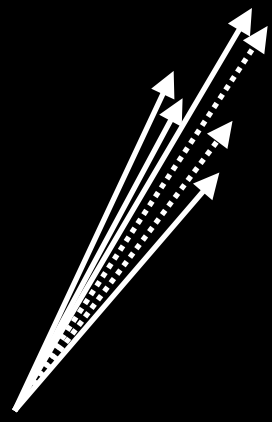


Top quark?

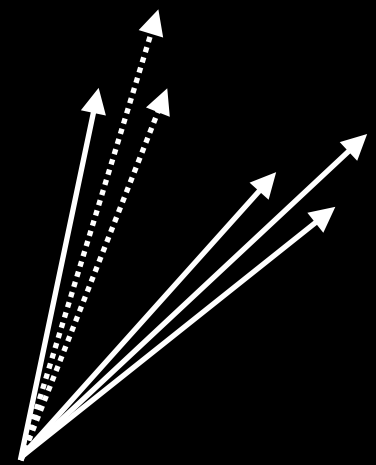


Something else?



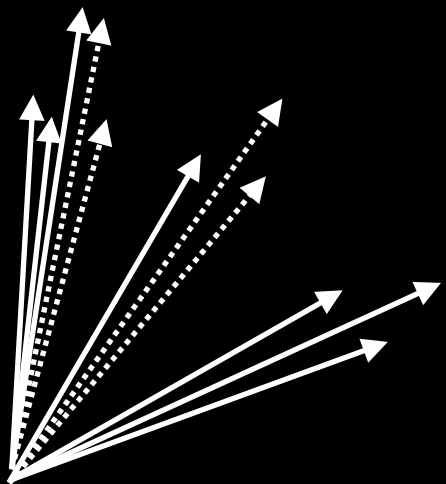


Quark? Gluon?

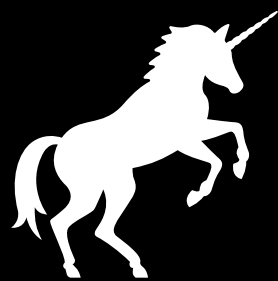


Higgs boson?

W/Z boson?



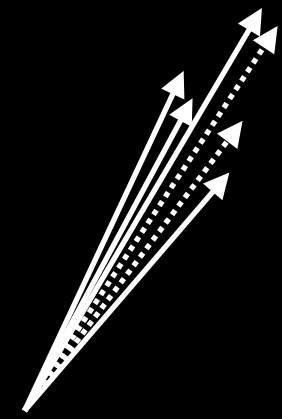
Top quark?



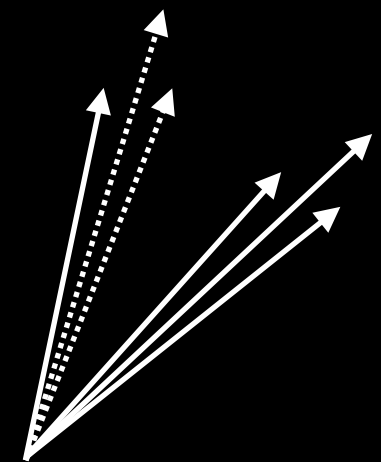
Something else?



(Dark Shower)

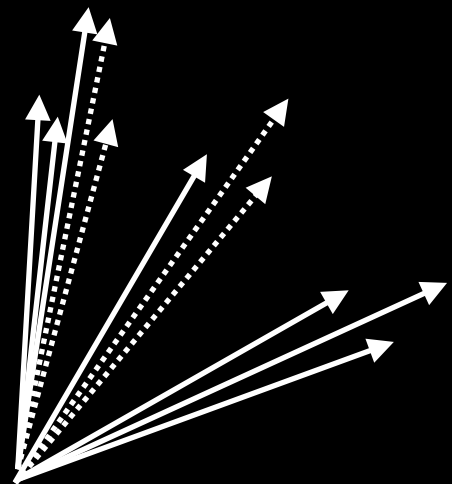


Quark? Gluon?

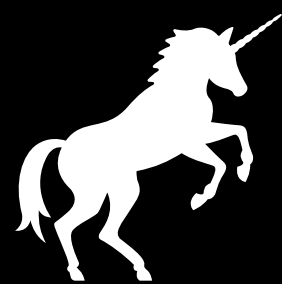


Higgs boson?

W/Z boson?



Top quark?



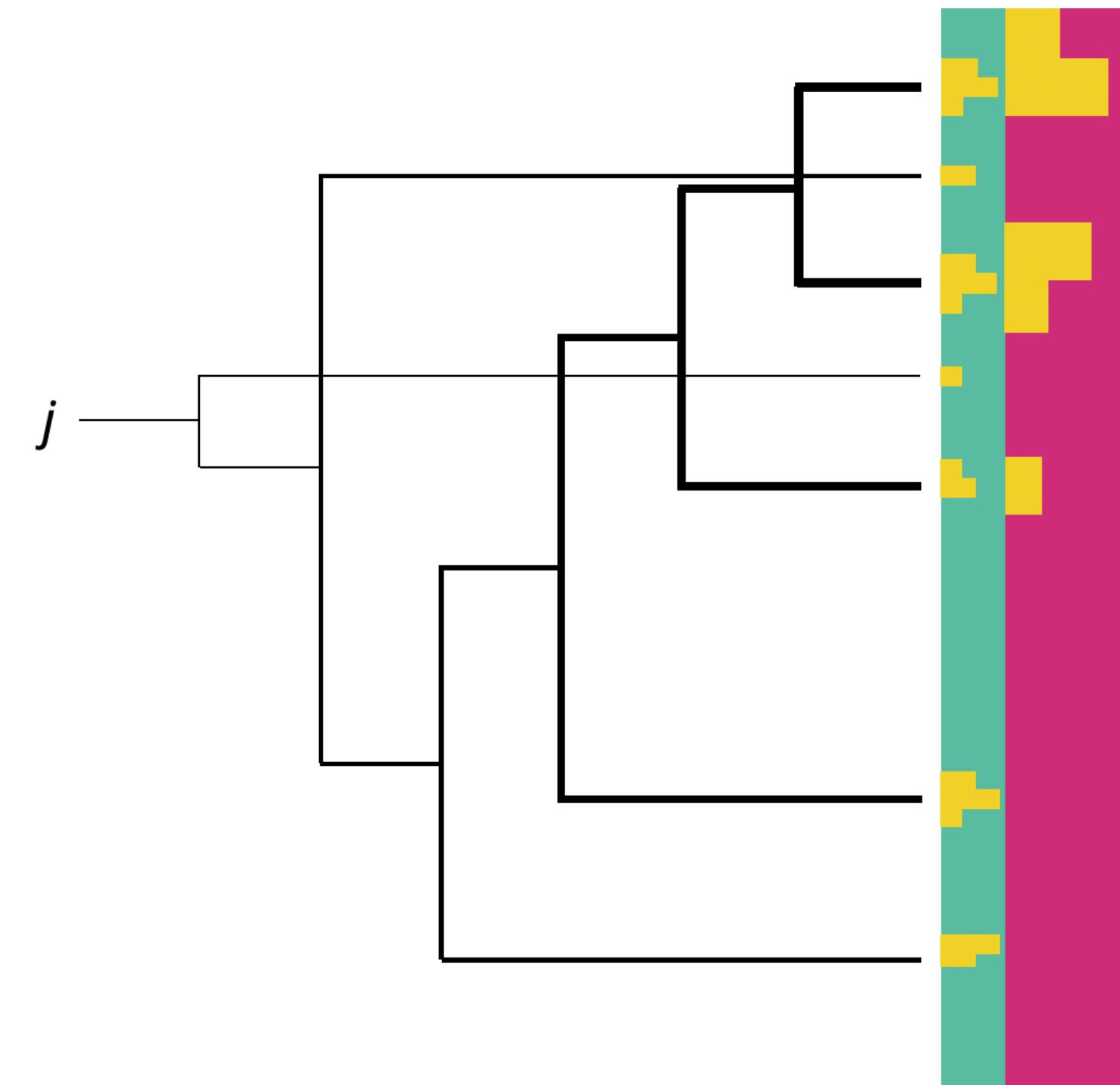
Something else?



(Dark Shower)

(Pause for laughter)

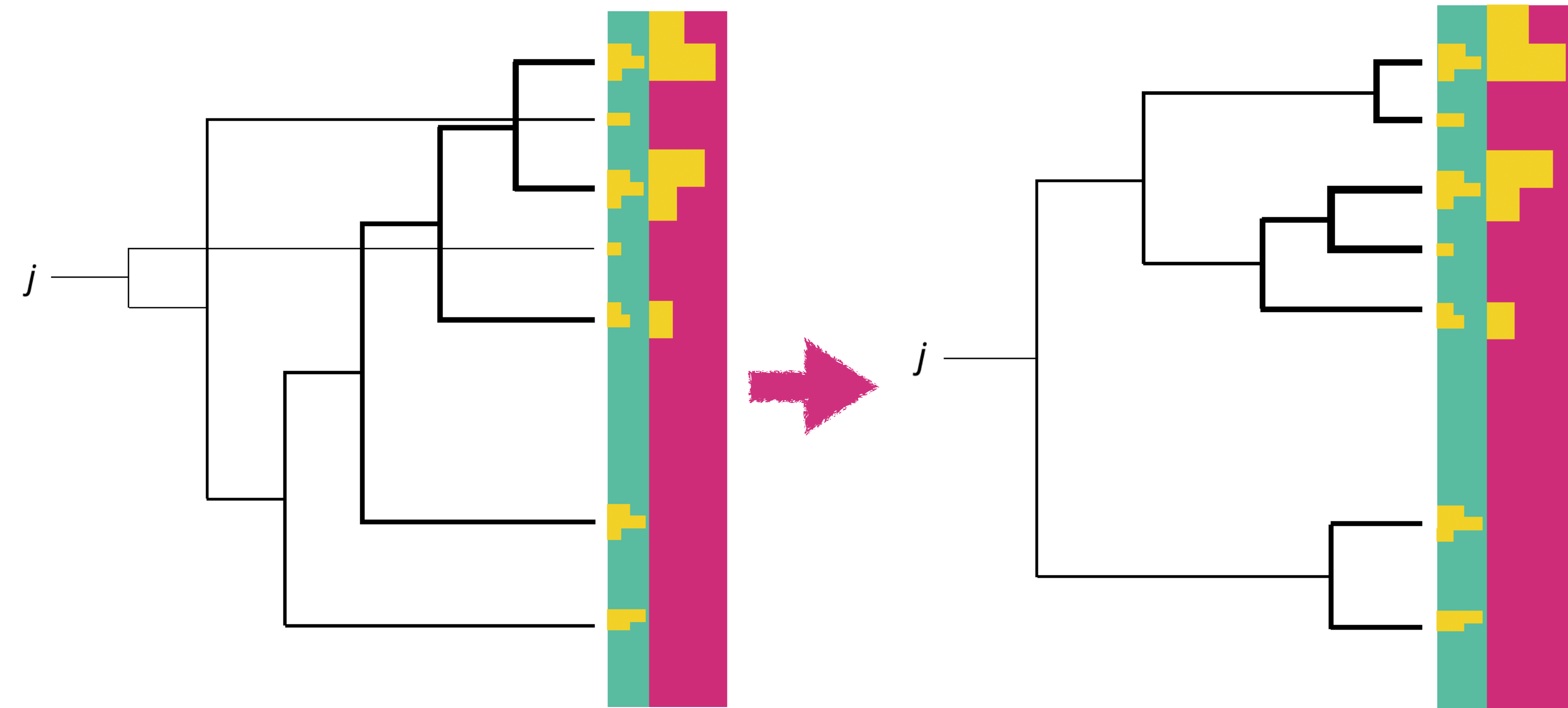
The Soft-Drop / modified Mass-Drop Algorithm



1. Start with anti- k_t jet.

*Dasgupta, Fregoso, Marzani, Salam, JHEP09 (2013) 029,
Larkowski, Marzani, Soyez, Thaler, JHEP 1405 (2014) 146*

The Soft-Drop / modified Mass-Drop Algorithm

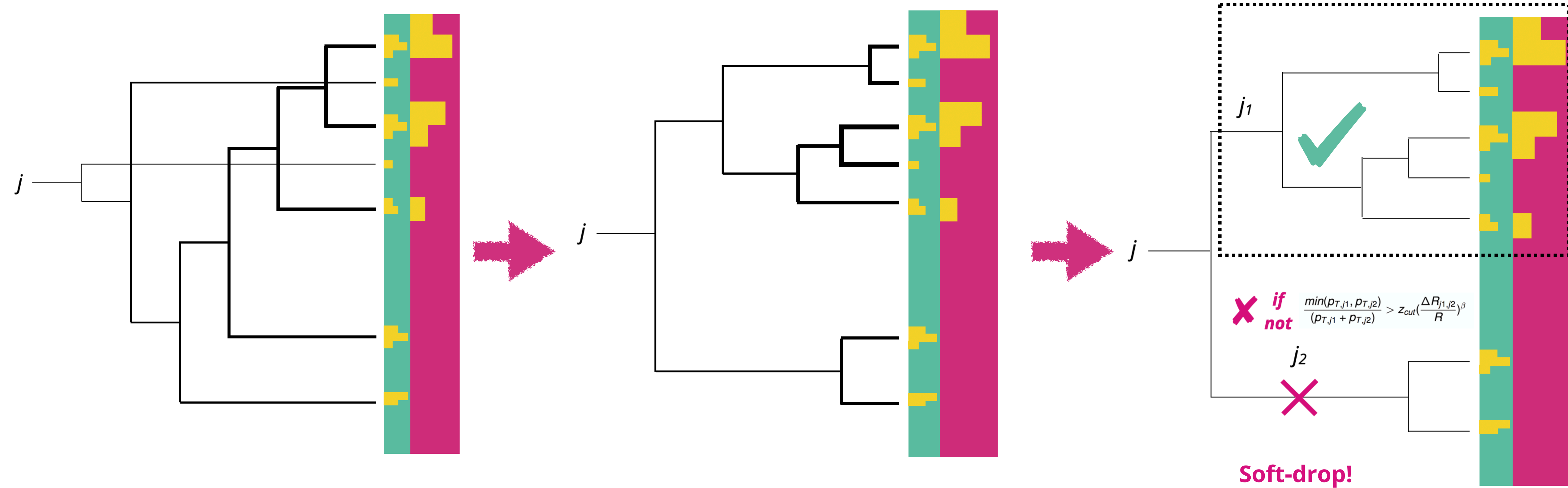


1. Start with anti- k_t jet.

**2. Recluster with C/A algorithm.
(Angle-ordered!)**

*Dasgupta, Fregoso, Marzani, Salam, JHEP09 (2013) 029,
Larkowski, Marzani, Soyez, Thaler, JHEP 1405 (2014) 146*

The Soft-Drop / modified Mass-Drop Algorithm



1. Start with anti- k_t jet.

2. Recluster with C/A algorithm.
(Angle-ordered!)

3. Check soft-drop condition at each node, starting with the widest-angle emission.

Stop when one passes!

*Dasgupta, Fregoso, Marzani, Salam, JHEP09 (2013) 029,
Larkowski, Marzani, Soyez, Thaler, JHEP 1405 (2014) 146*

Lund jet plane

Dreyer, Salam & Soyez [JHEP 12 \(2018\) 064](#)

- **Lund Plane** : tool used by Parton Shower community for >30 years (Andersson et al. [Z.Phys.C 43 \(1989\) 625](#))
- Newly applied to JSS by **Dreyer et al.**
 - **Key concept:** probe entire angle-ordered emission history of originating parton.
 - Recluster jet with C/A algorithm
 - Parameterise emissions in terms of their **relative energies (z)** and **angles (ΔR)**.
- Powerful, physics-forward representation of JSS:
 - **ML/AI** [1903.09644](#), [2012.08526](#),
q/g tagging [2112.09140](#),
PSMC development [1805.09327](#), [2205.02861](#),
analytics [2007.06578](#),
heavy-flavour [2106.05713](#)
dead-cone *ALICE, Nature* 605, 440–446 (2022)
 - **[We had a whole LJP workshop at CERN in July!](#)**

