

Particle Flow Performance in CMS

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for the CMS Collaboration

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Menu

Ingrédients

CMS, tracks and clusters

La Recette

hadrons, photons, electrons, muons

Plat Principal

jets, MET, taus, leptons (and dealing with pileup)

Desserts

the future: HGCal, deep learning

Particle-flow reconstruction and global event description with the CMS detector



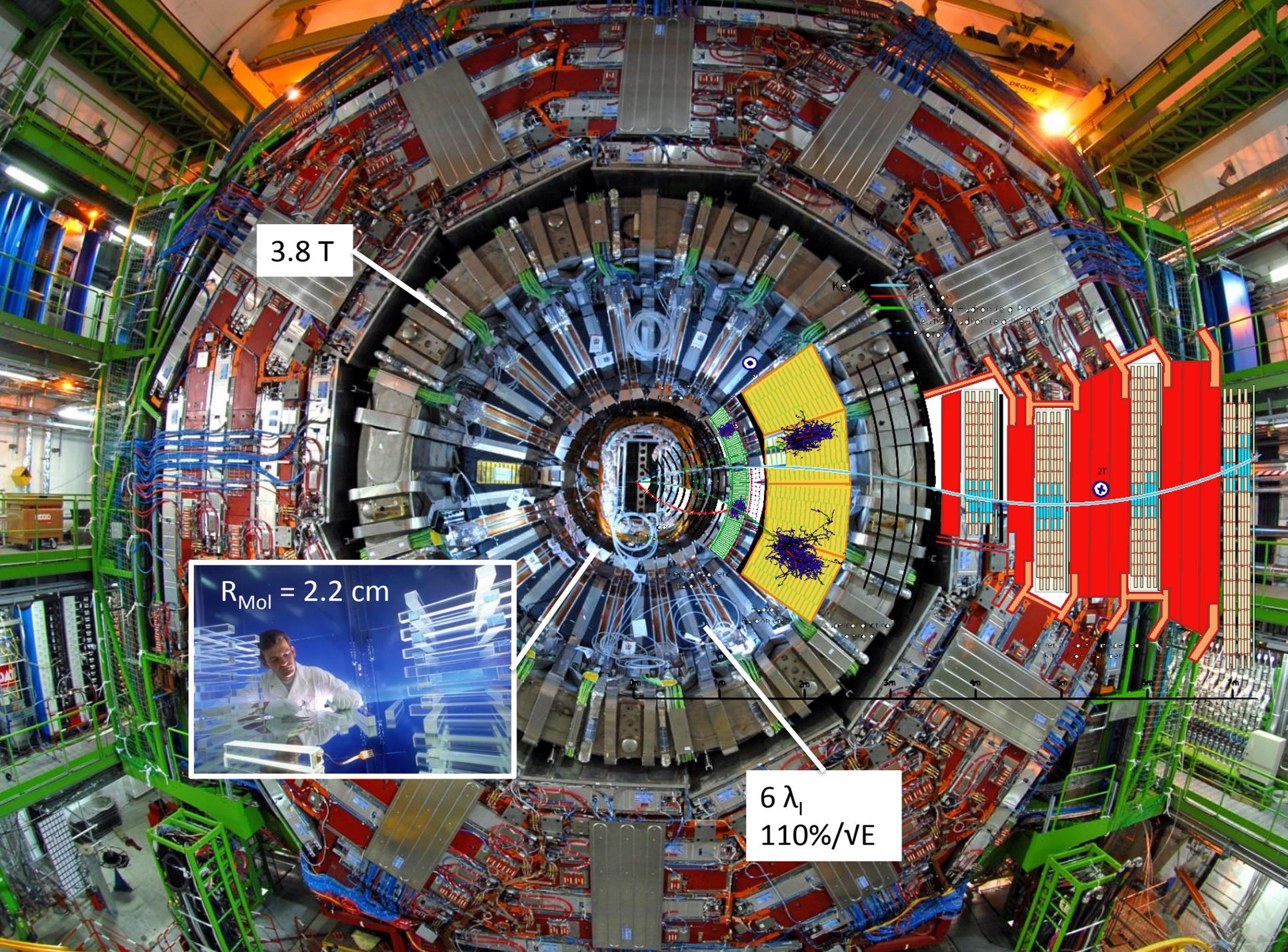
The CMS collaboration

E-mail: cms-publication-committee-chair@cern.ch

ABSTRACT: The CMS apparatus was identified, a few years before the start of the LHC operation at CERN, to feature properties well suited to particle-flow (PF) reconstruction: a highly-segmented tracker, a fine-grained electromagnetic calorimeter, a hermetic hadron calorimeter, a strong magnetic field, and an excellent muon spectrometer. A fully-fledged PF reconstruction algorithm tuned to the CMS detector was therefore developed and has been consistently used in physics analyses for the first time at a hadron collider. For each collision, the comprehensive list of final-state particles identified and reconstructed by the algorithm provides a global event description that leads to unprecedented CMS performance for jet and hadronic τ decay reconstruction, missing transverse momentum determination, and electron and muon identification. This approach also allows particles from pileup interactions to be identified and enables efficient pileup mitigation methods. The data collected by CMS at a centre-of-mass energy of 8 TeV show excellent agreement with the simulation and confirm the superior PF performance at least up to an average of 20 pileup interactions.

KEYWORDS: Large detector systems for particle and astroparticle physics; Particle identification methods

ARXIV EPRINT: [1706.04965](https://arxiv.org/abs/1706.04965)



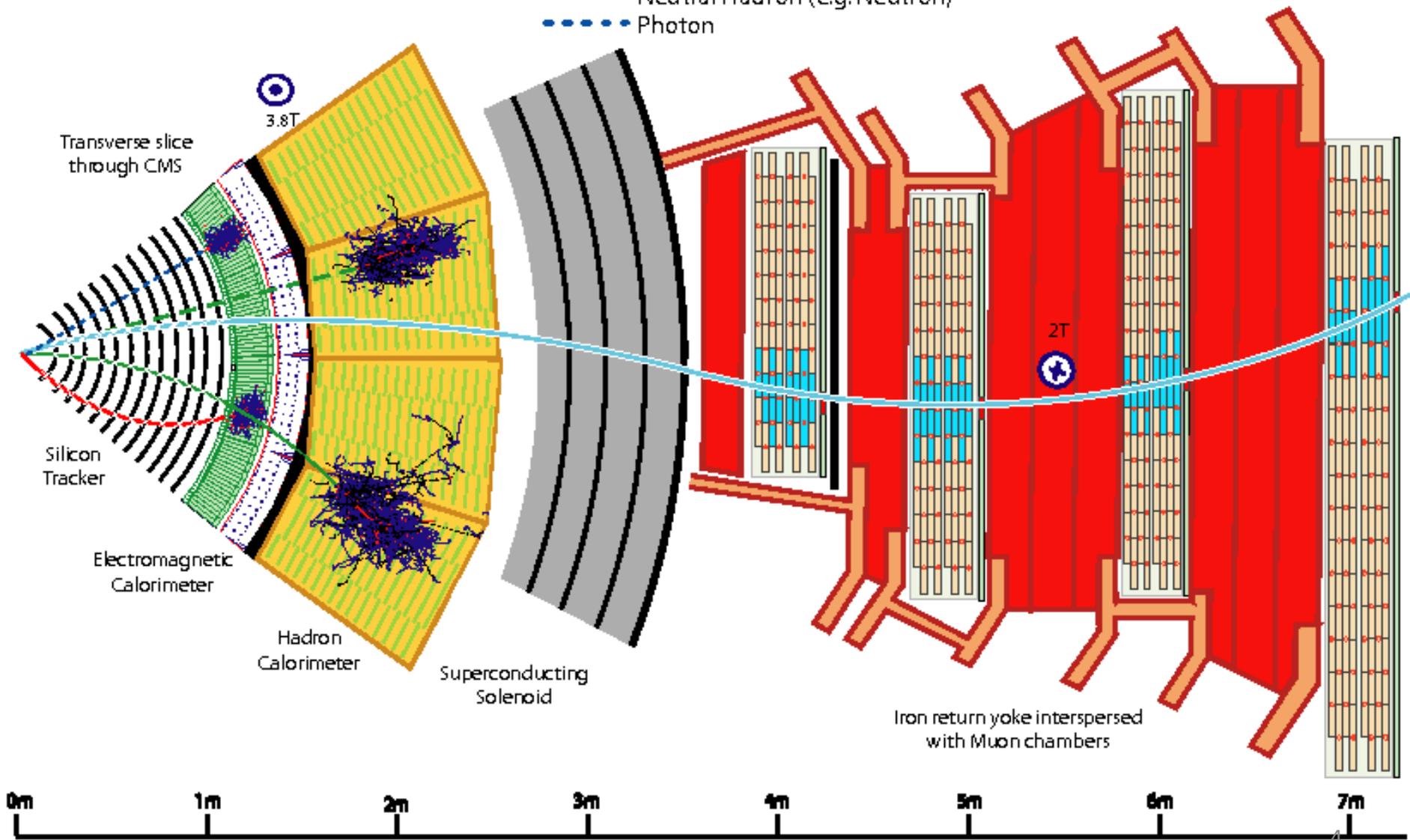
3.8 T

$R_{Mol} = 2.2 \text{ cm}$

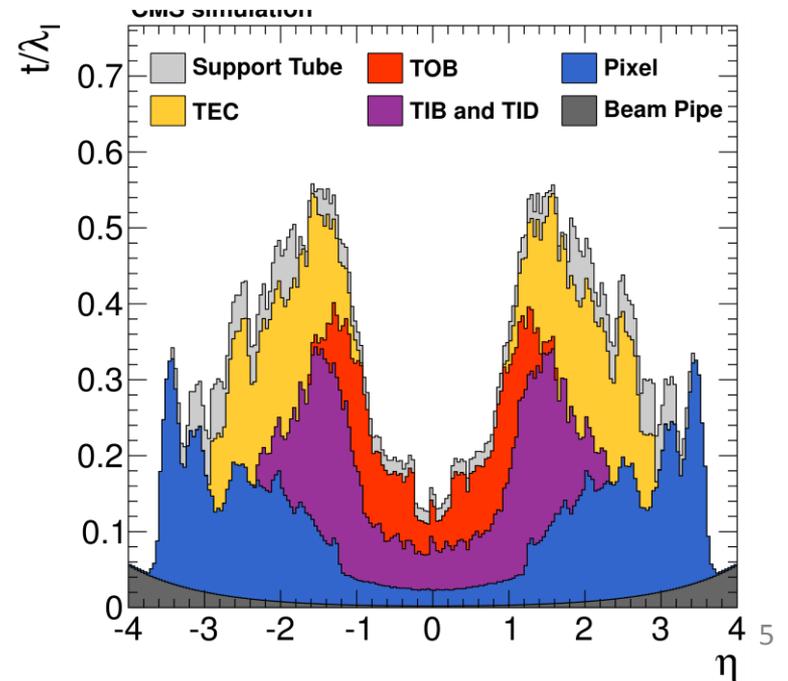
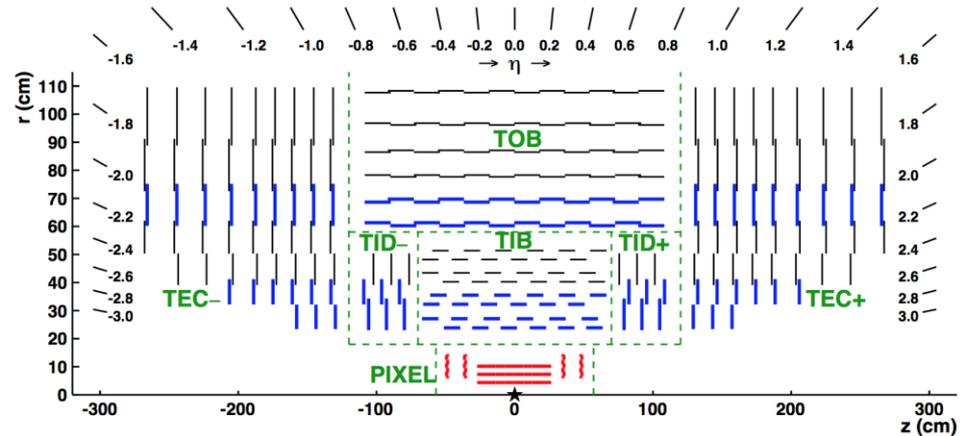
$6 \lambda_1$
110%/VE

2T

- Key:
- Muon
 - Electron
 - Charged Hadron (e.g. Pion)
 - - - Neutral Hadron (e.g. Neutron)
 - Photon



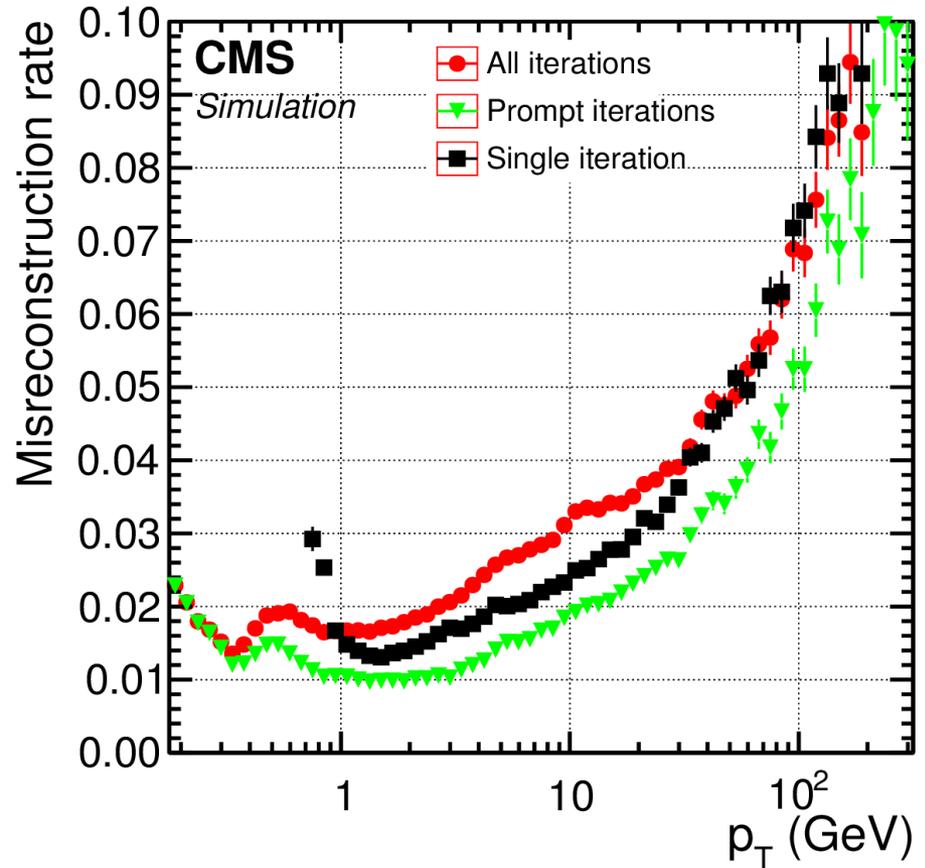
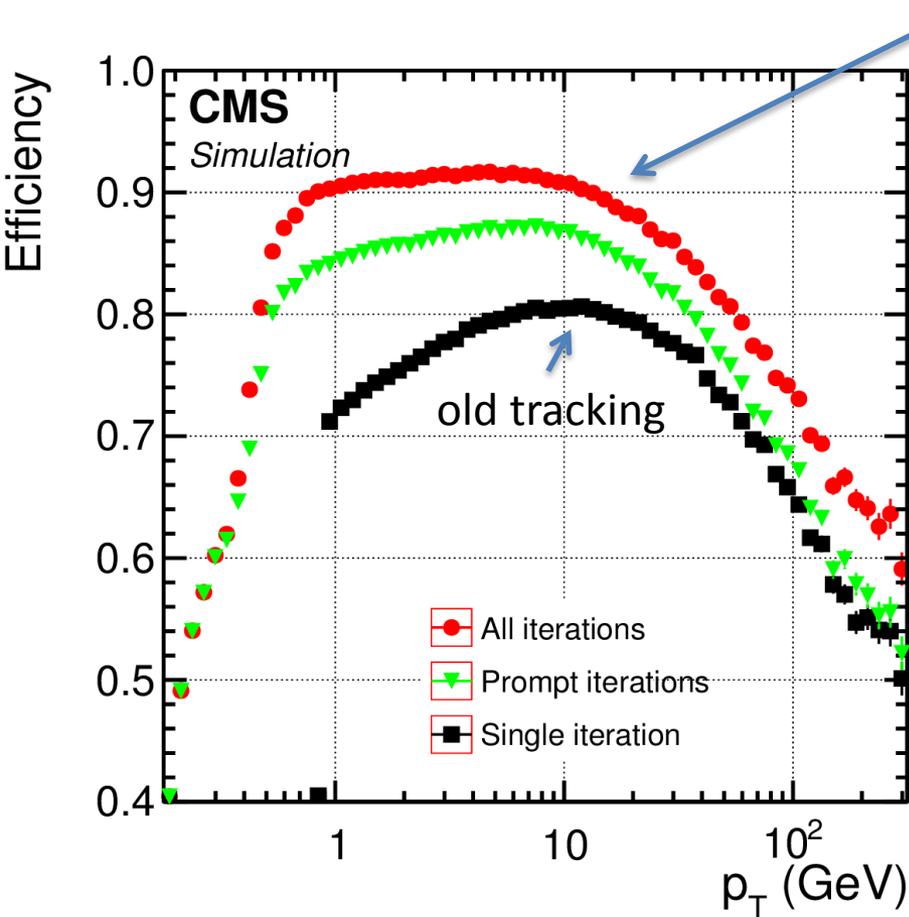
The CMS Tracker: Big! (and Thick)



Hadrons: nuclear interactions

e/γ : bremsstrahlung,
conversions

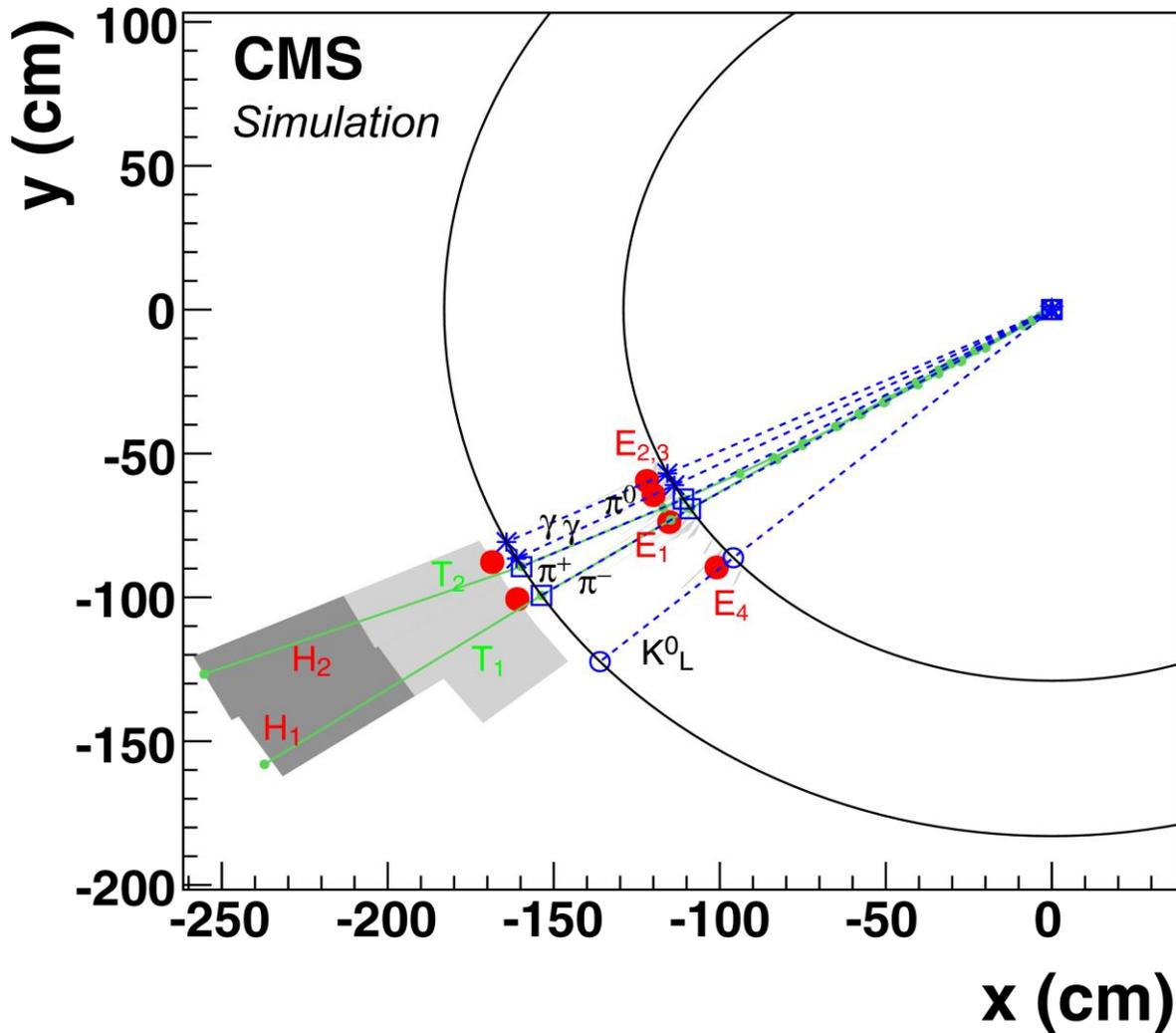
Iterative tracking



10 iterations:

- 1- Reconstruct easy tracks
- 2- Remove their hits
- 3- Reconstruct more difficult tracks

A Simple Jet

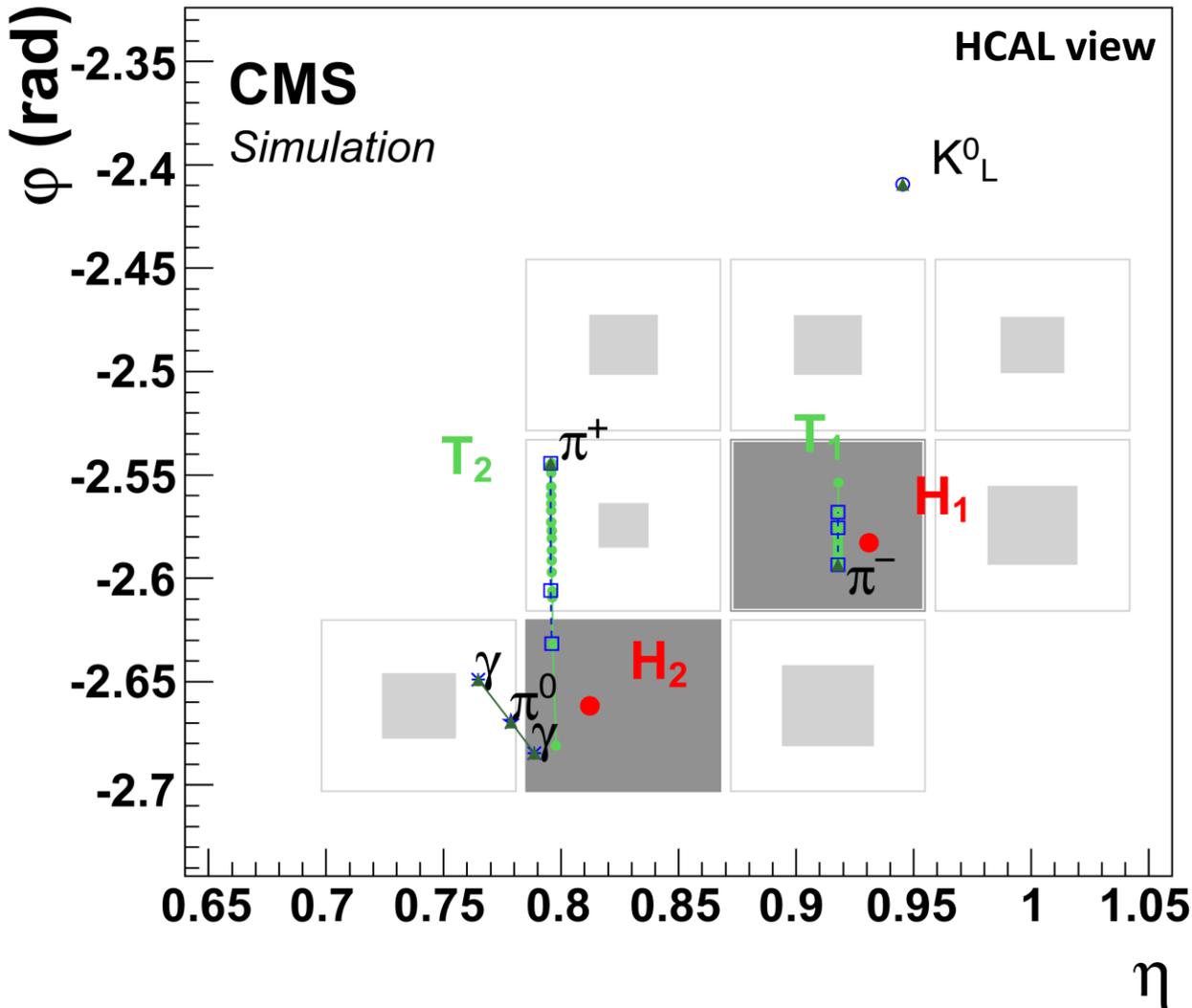


$p_T = 65$ GeV
only 5 particles

Illustrate:

- calo clustering
- particle flow

Calorimeter Clustering



Seeds

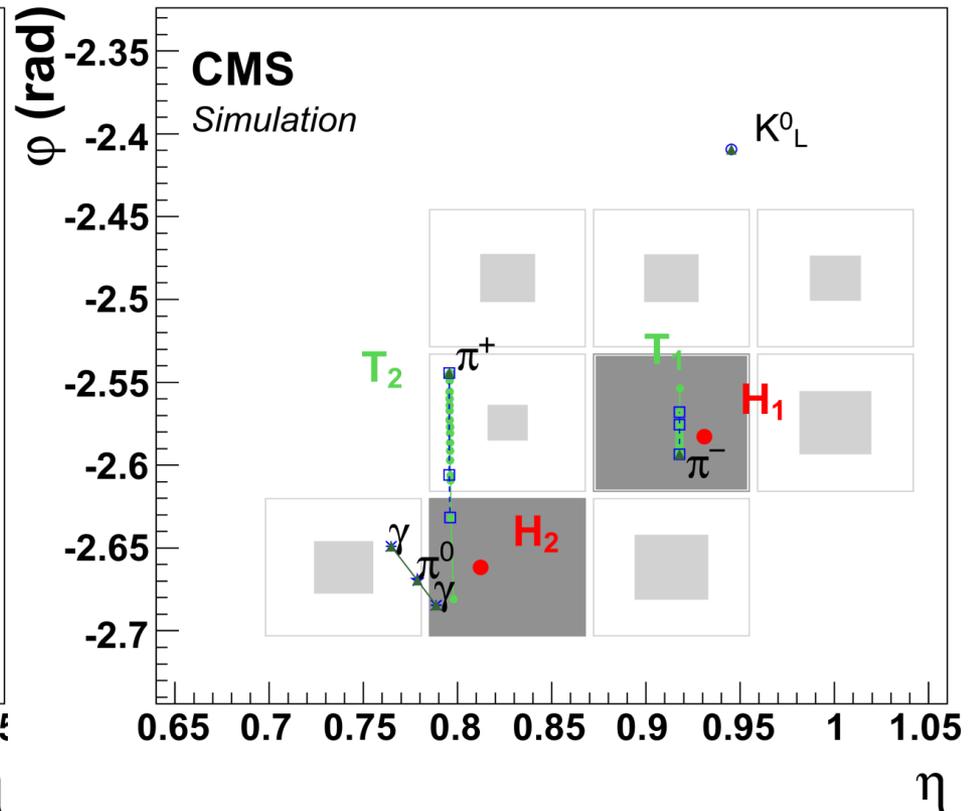
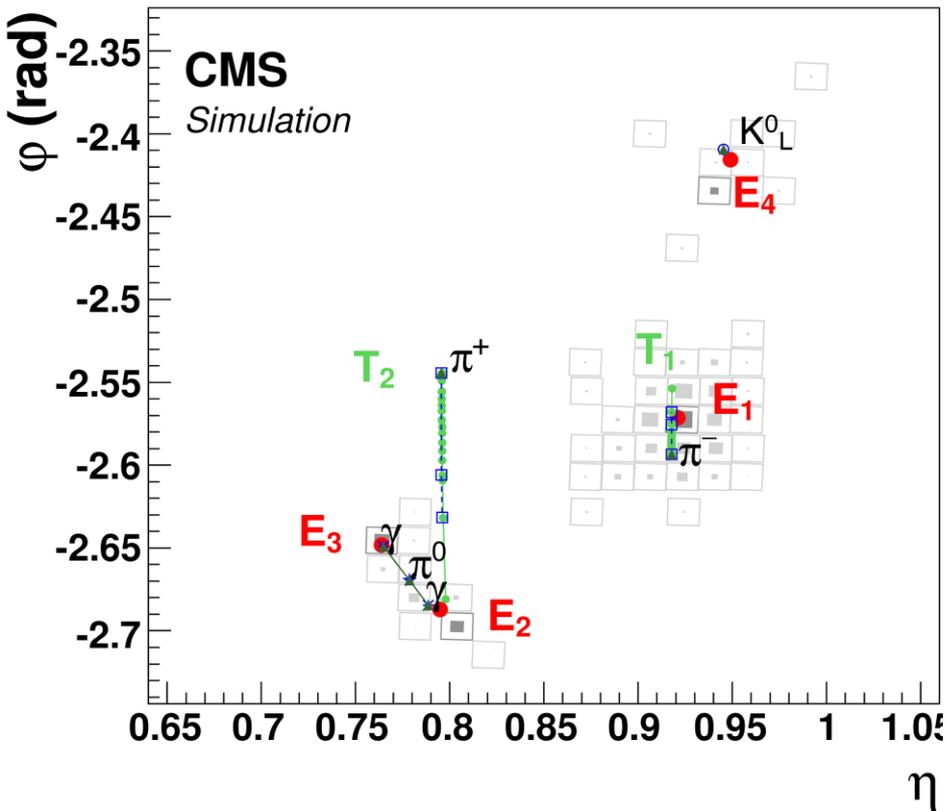
- local energy maxima

Connected cells

Share energy

- iteratively
- assuming Gaussian shower profile

Particle Flow



Clusters not linked to a track \rightarrow photon (ECAL) or neutral hadron (HCAL)

Tracks \rightarrow charged hadrons

- compatible energy in calos: energy from a fit of track and cluster measurements
- excess: additional neutrals
- deficit: muon, fake track

-- iSpy -- <http://iguana.cern.ch/ispy>
Data recorded 1970-Jan-01 00:13:55 GMT
Run number 1
Event number 667
Lumi section 666668
Orbit number -1
Beam crossing -1

Electrons

L1 Triggers:

L1_DoubleEG1
L1_DoubleEG5
L1_SingleEG1
L1_SingleEG2
L1_SingleEG5
L1_SingleIsoEG5
L1_SingleJet15
L1_ZeroBias

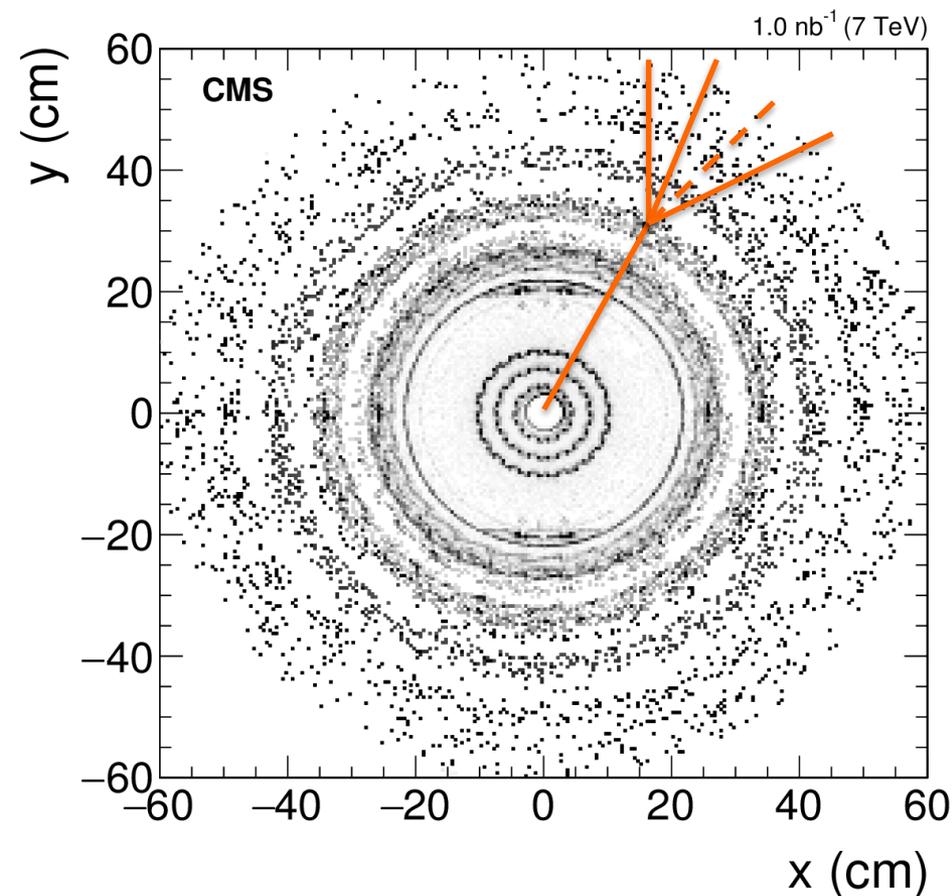
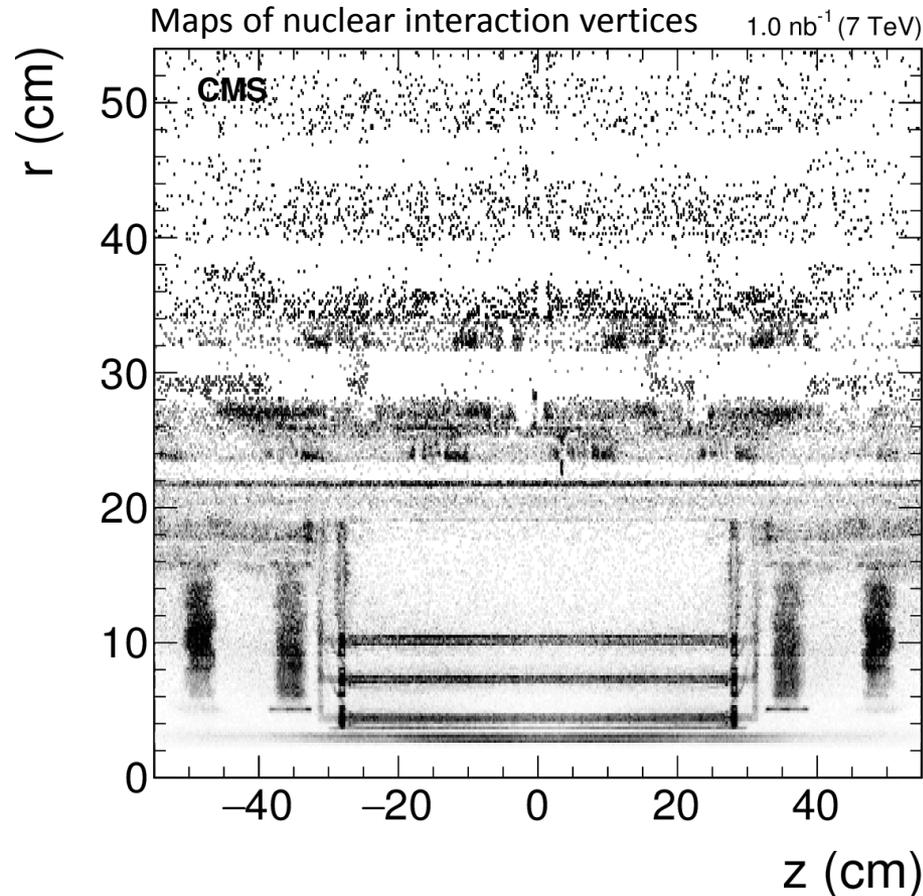
They brem
Brem photons convert

Conversion tracks
collect secondary
electron clusters

Track momentum
change followed by
Gaussian Sum Filter

Brem clusters collected
by « track tangents »

Nuclear Interactions

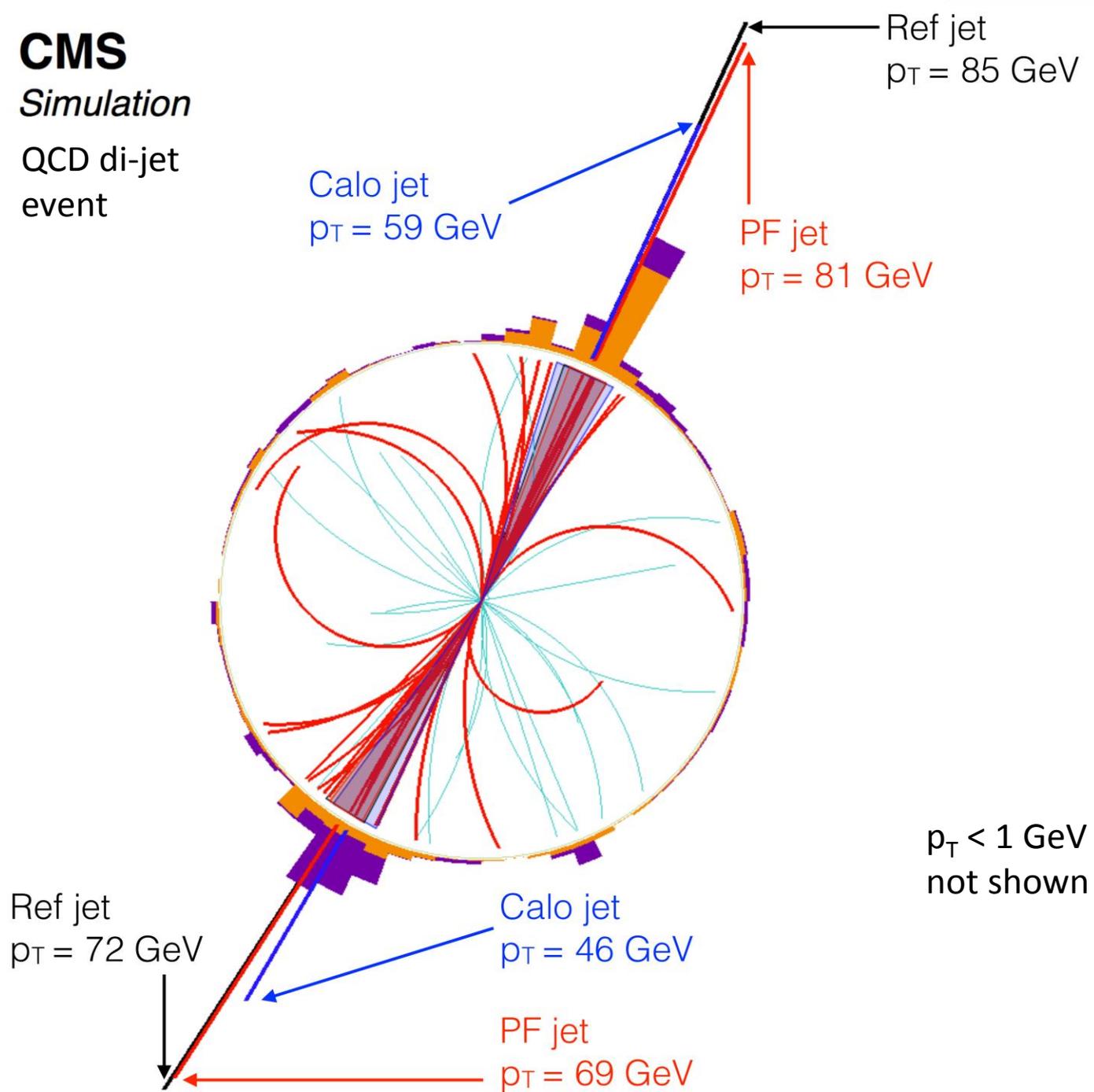


- Secondary vertices reconstructed from displaced tracks (+ 0 or 1 incoming track)
- Secondary charged particles reconstructed by PF \rightarrow single charged hadron
- Secondary neutrals reconstructed by PF as usual
- Incoming track if any is discarded

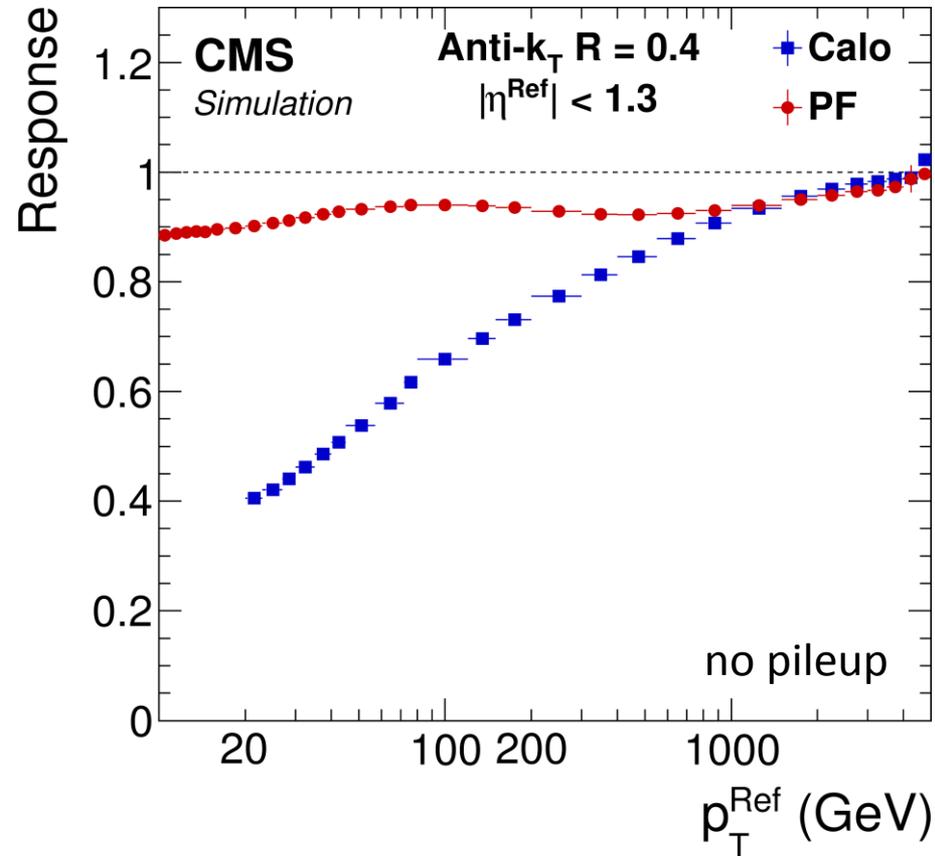
CMS

Simulation

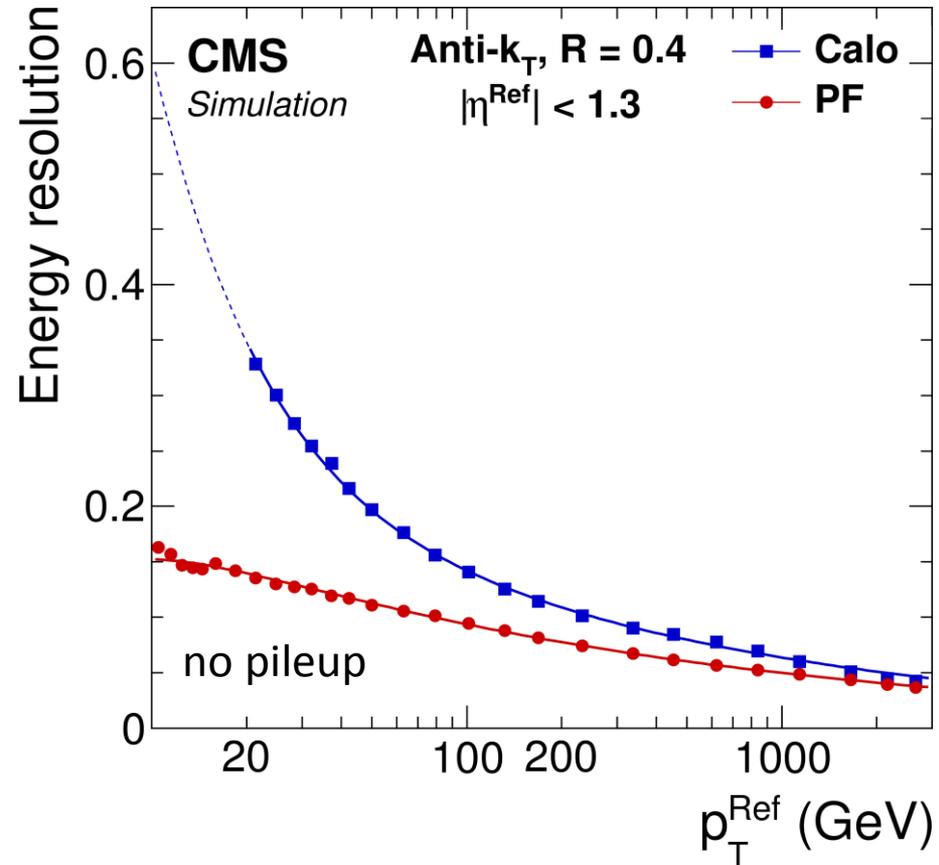
QCD di-jet
event



Jet Response and Resolution



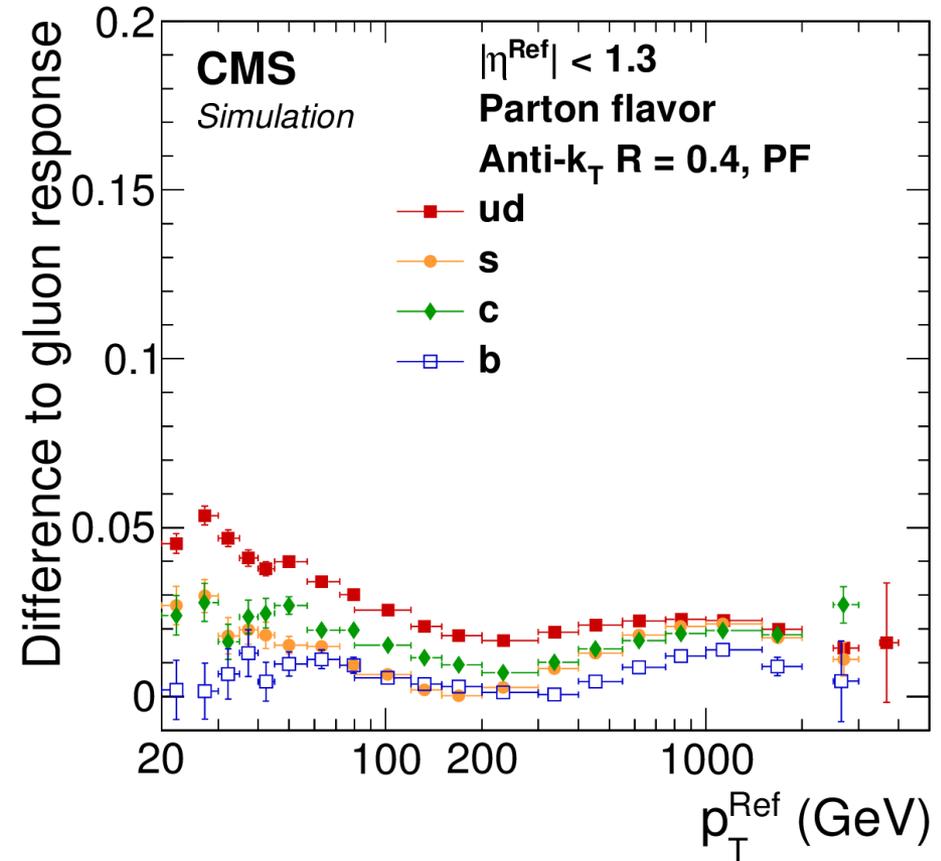
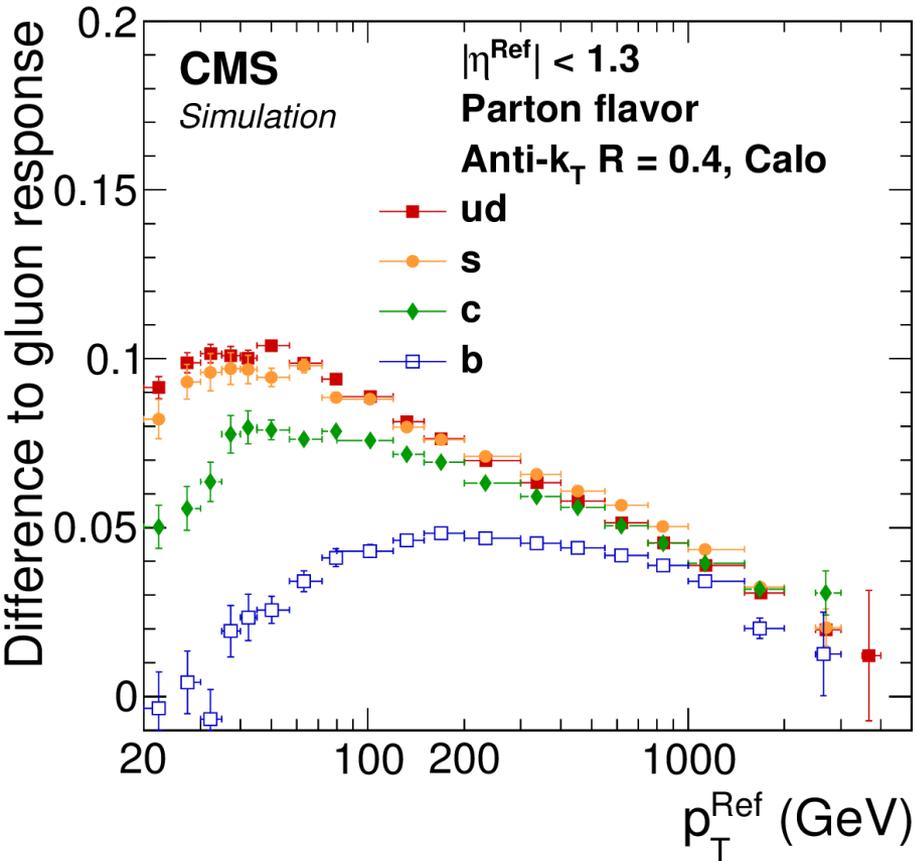
Response closer to unity
 \sim linear



Resolution always better than for calo jets:

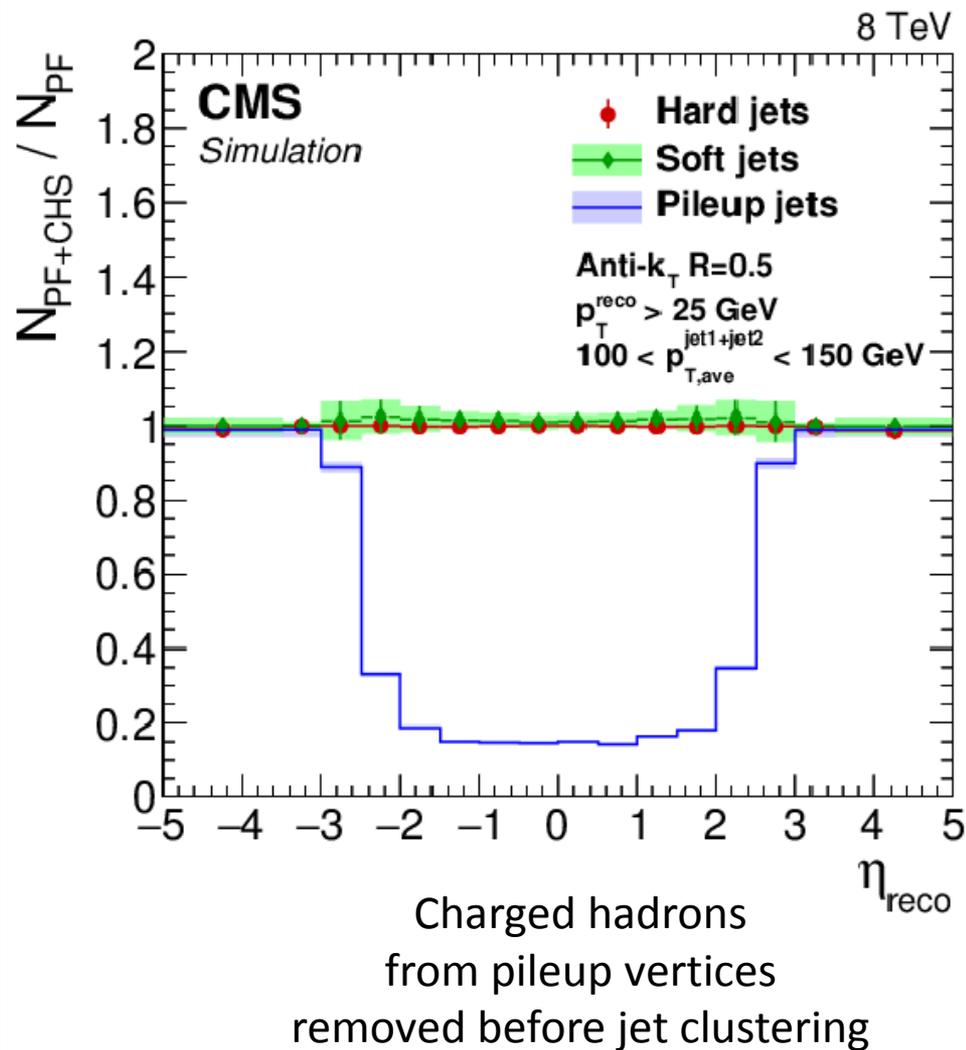
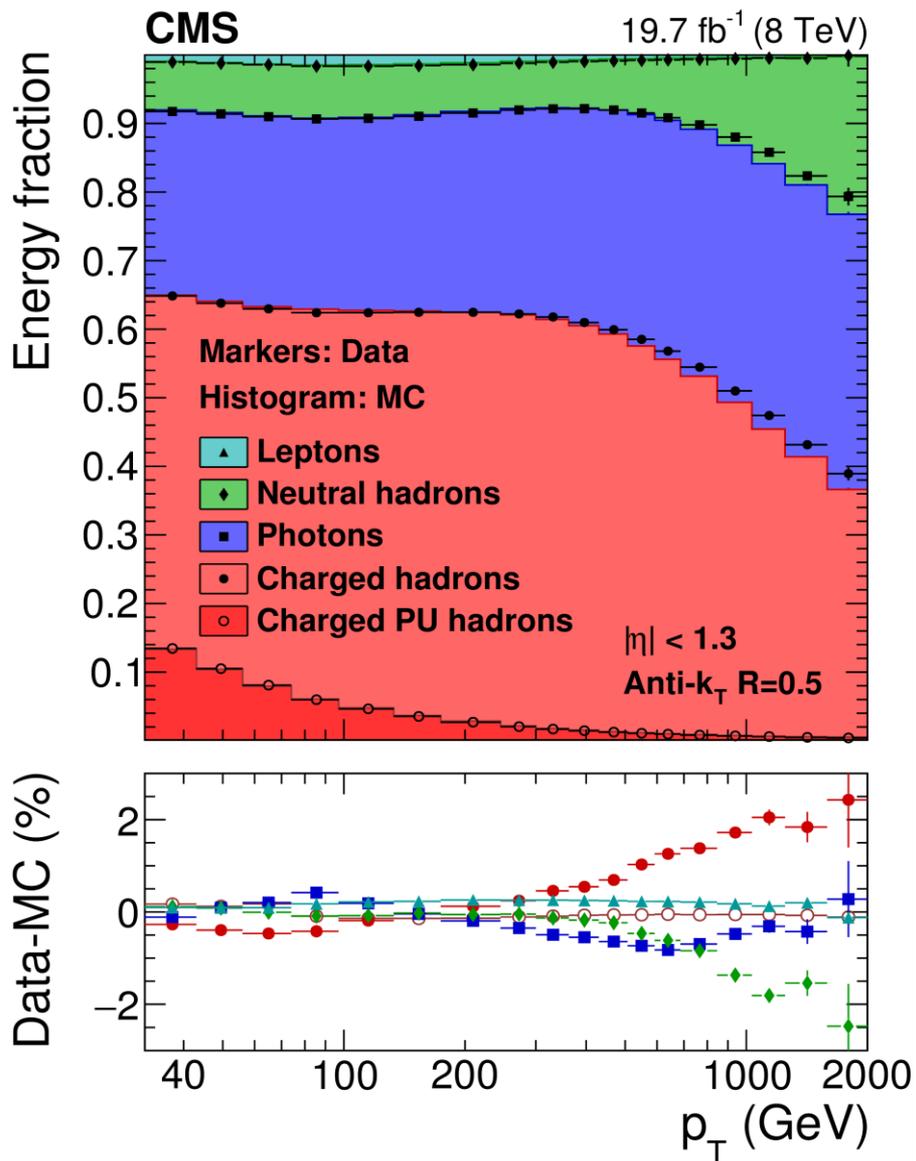
- PF works when particles are close
- charged hadron energy from fit of track and cluster measurements

Response for different flavours



Sensitivity of the response to the parton flavour reduced
→ Jet energy scale systematic uncertainty reduced

Jet Composition, Pileup



Particle Flow and Jets @HLT

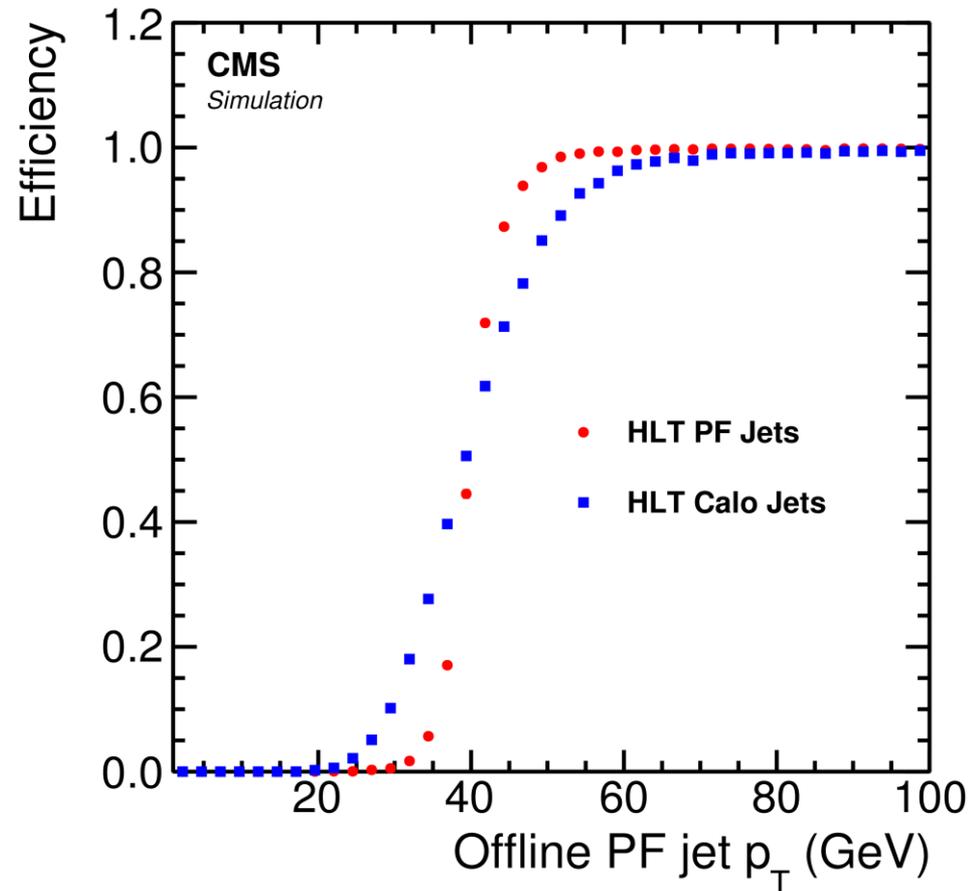
In the High-Level trigger, timing is crucial! (~ 140 ms / evt)

	Tracking	PF
Offline	600 ms	70 ms
HLT	60 ms	30 ms

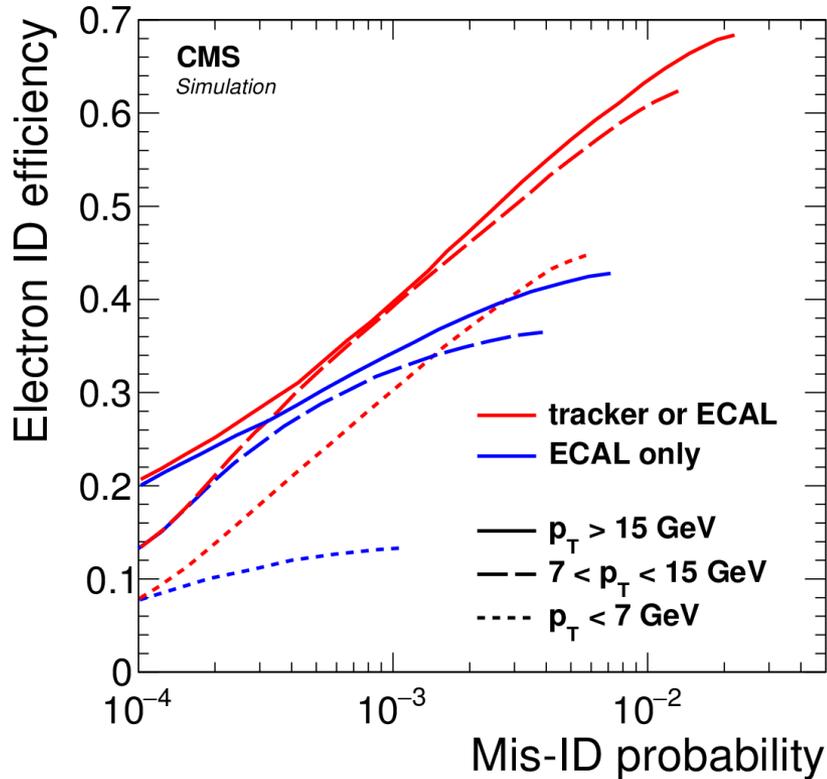
no pileup, assuming tracking and PF are performed for every event

Fraction of total time at HLT @45 pileup:

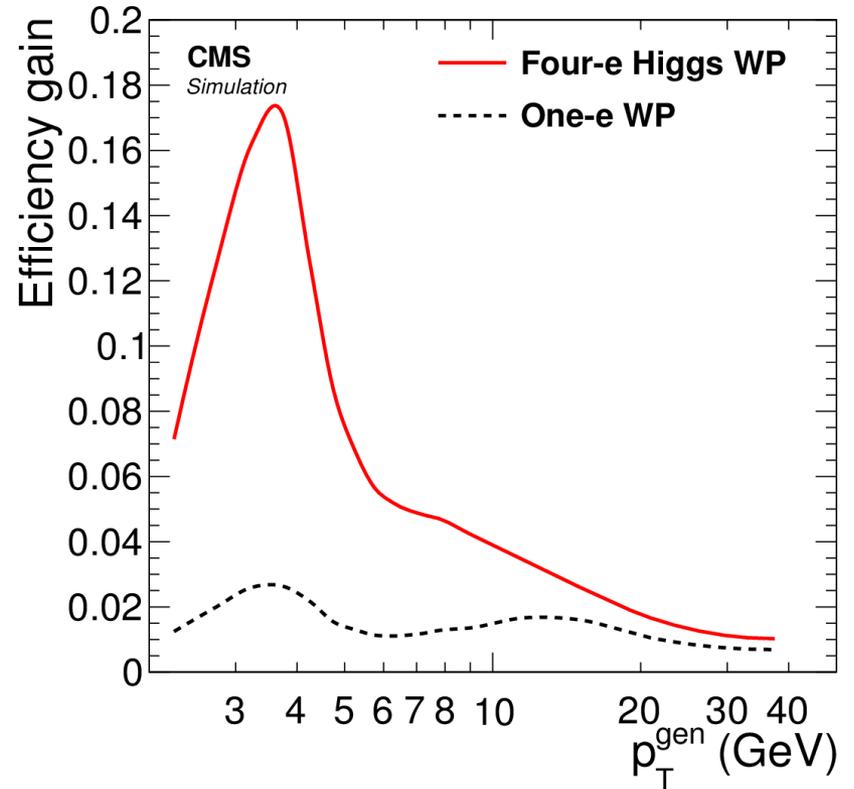
	Tracking	PF
HLT	< 20%	< 10 %



Electrons

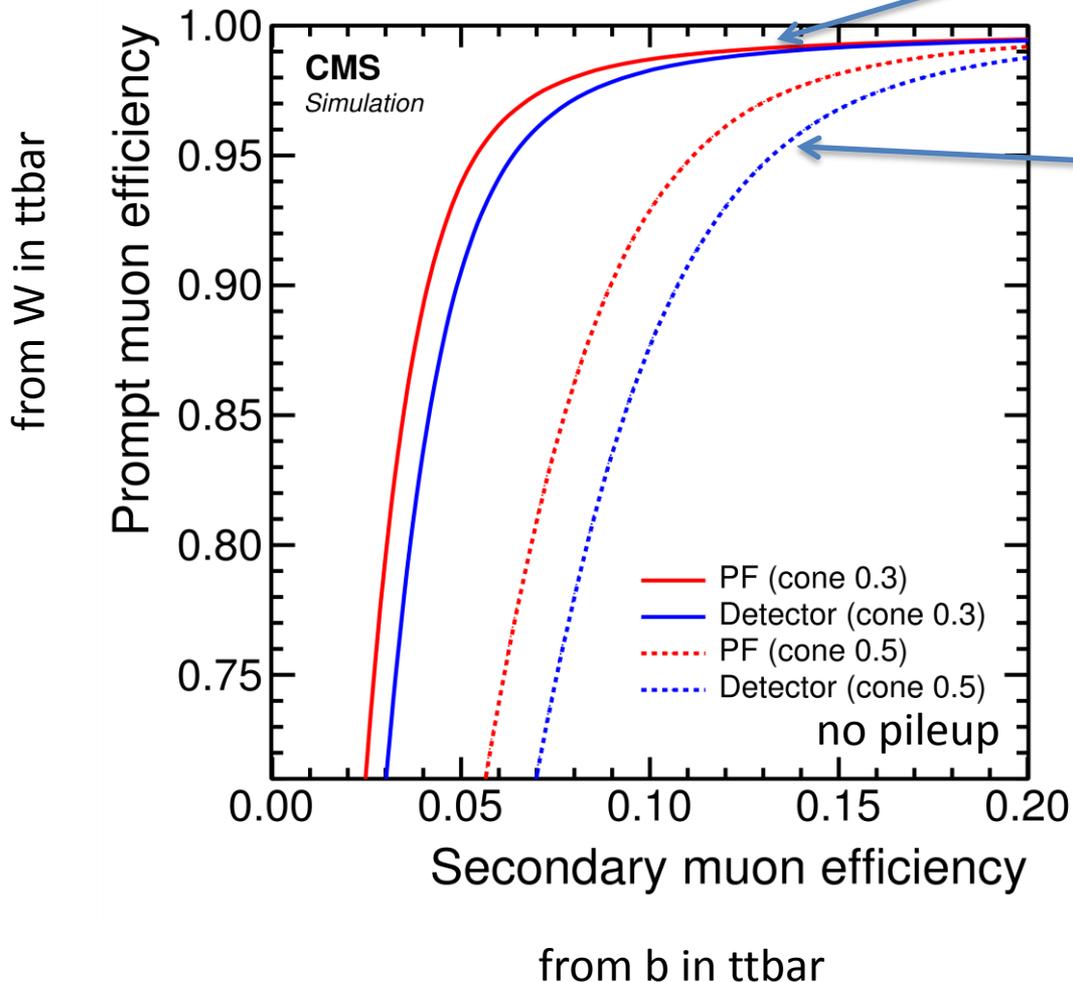


Electrons in jets
b tagging with electrons possible



Prompt electrons
Big efficiency gains at low p_T :
+7% more $H \rightarrow ZZ \rightarrow 4e$ events

Isolation



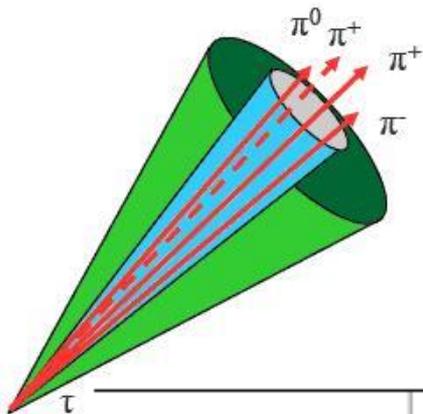
$$I_{\text{PF}} = \frac{1}{p_T} \left(\sum_{h^\pm} p_T^{h^\pm} + \sum_{\gamma} p_T^{\gamma} + \sum_{h^0} p_T^{h^0} \right)$$

$$I_{\text{det}} = \frac{1}{p_T} \left(\sum_{\text{tracks}} p_T^{\text{track}} + \sum_{\text{ECAL}} E_T^{\text{ECAL}} + \sum_{\text{HCAL}} E_T^{\text{HCAL}} \right)$$

Detector-based isolation:

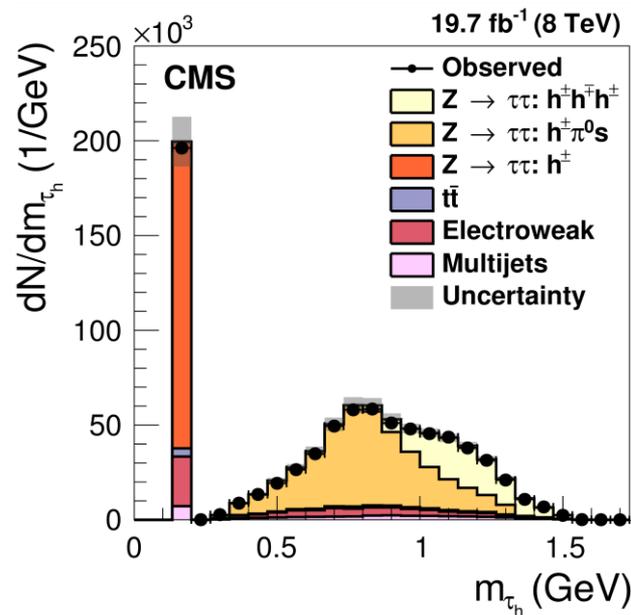
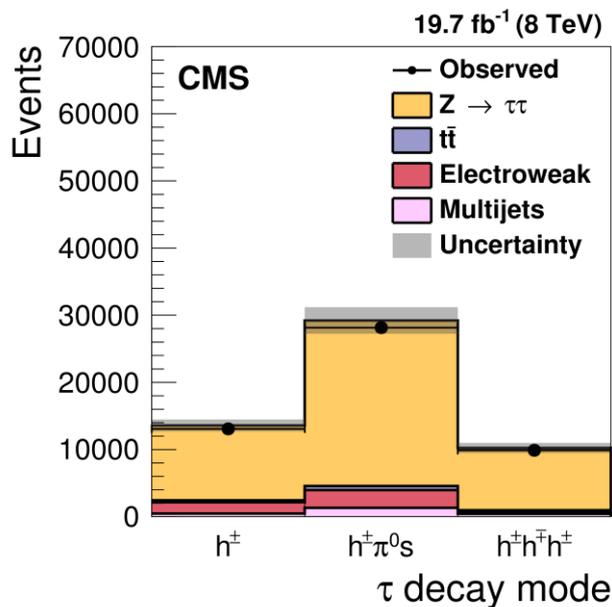
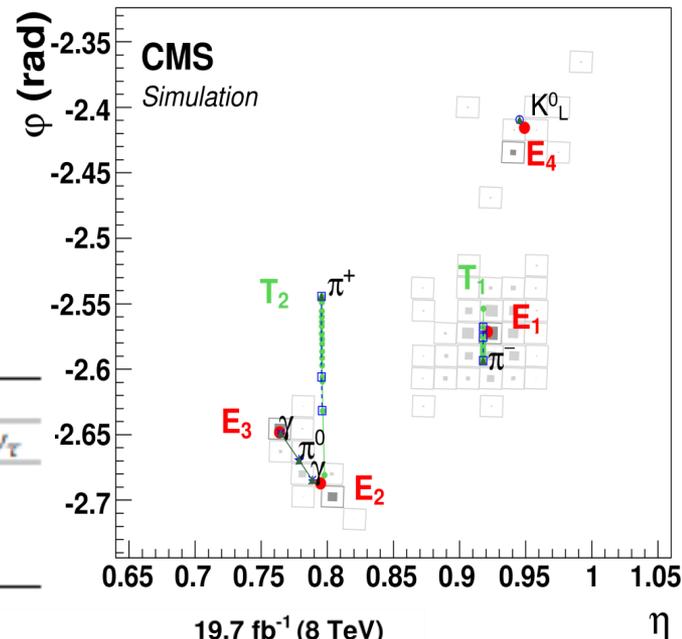
- Charged-hadron energy double counted
- Cannot easily remove pileup calorimeter deposits

Tau Reconstruction & Identification

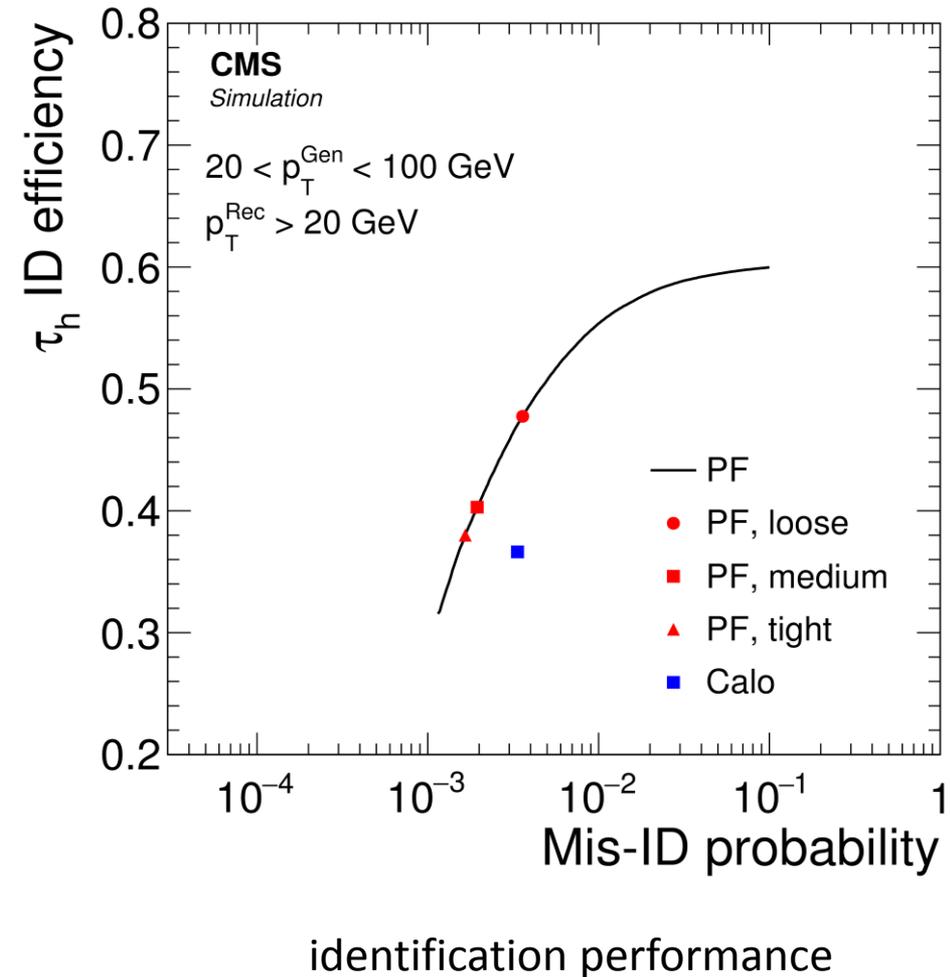
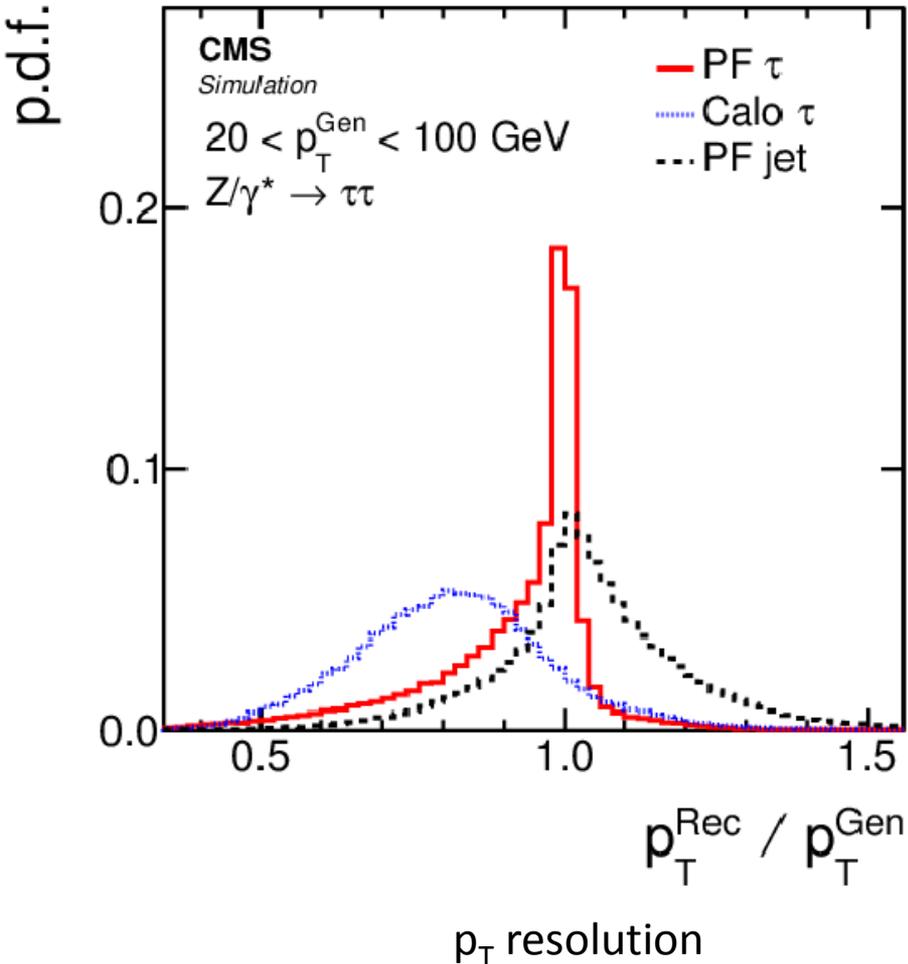


1 or 3 π^\pm
photons from π^0

Reconstructed	Generated		
	$\tau^- \rightarrow h^- \nu_\tau$	$\tau^- \rightarrow h^- \geq 1\pi^0 \nu_\tau$	$\tau^- \rightarrow h^- h^+ h^- \nu_\tau$
$\tau^- \rightarrow h^- \nu_\tau$	0.89	0.16	0.01
$\tau^- \rightarrow h^- \geq 1\pi^0 \nu_\tau$	0.11	0.83	0.02
$\tau^- \rightarrow h^- h^+ h^- \nu_\tau$	0.00	0.01	0.97



Tau Reconstruction & Identification



HGCAL

<https://cds.cern.ch/record/2293646?ln=fr>

200 pileup collisions / beam crossing

Boosted jets

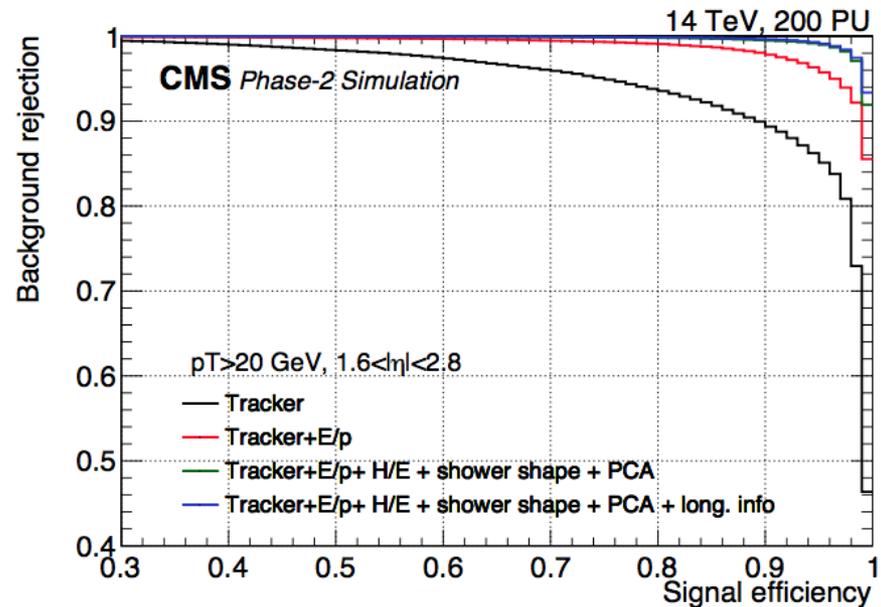
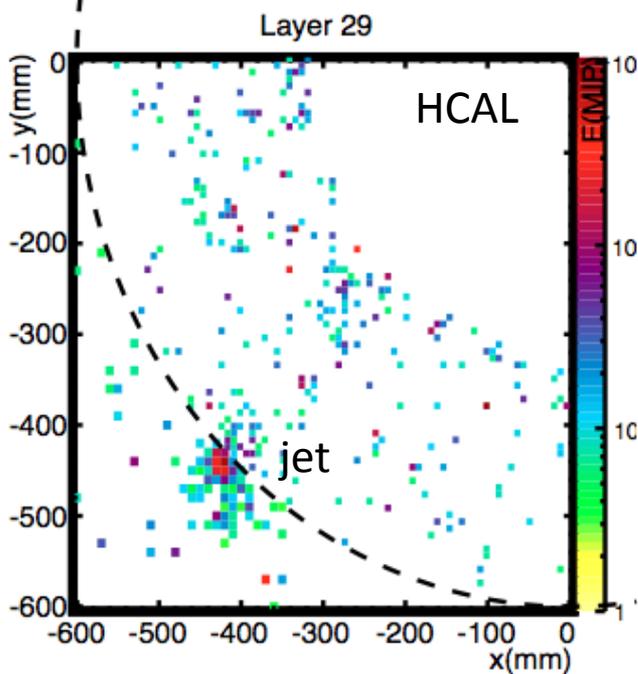
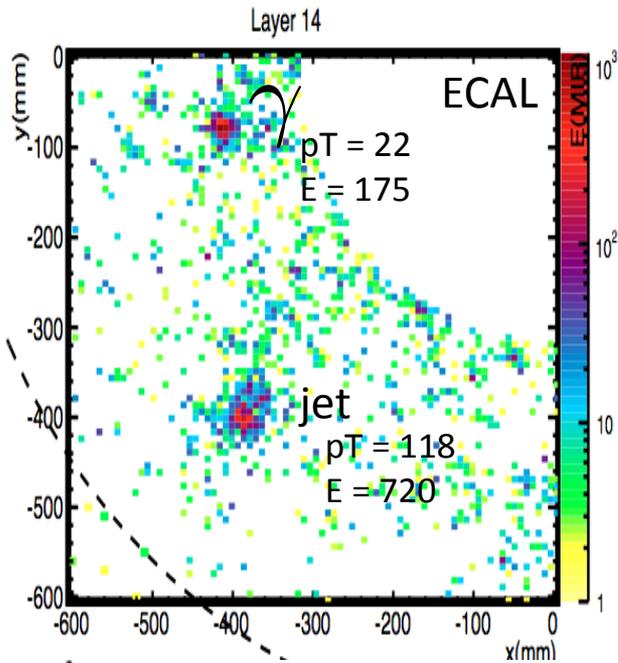
Pandora and Arbor:

- aggregate many unrelated hits

Tried a different approach:

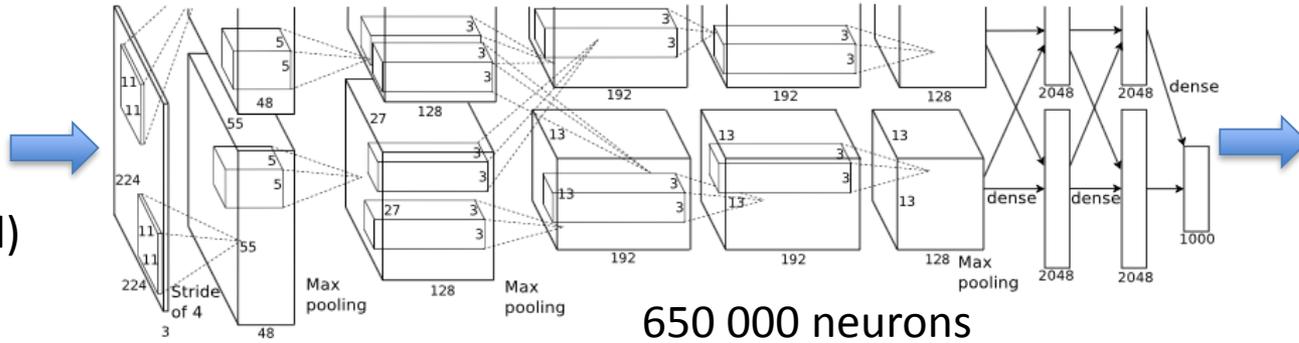
- fast 2D clustering in each layer
- projective association into 3D clusters
- works only for electrons so far

Setting up a full PF algo will take time and effort



Deep Learning

Input:
raw image
(color levels
for each pixel)



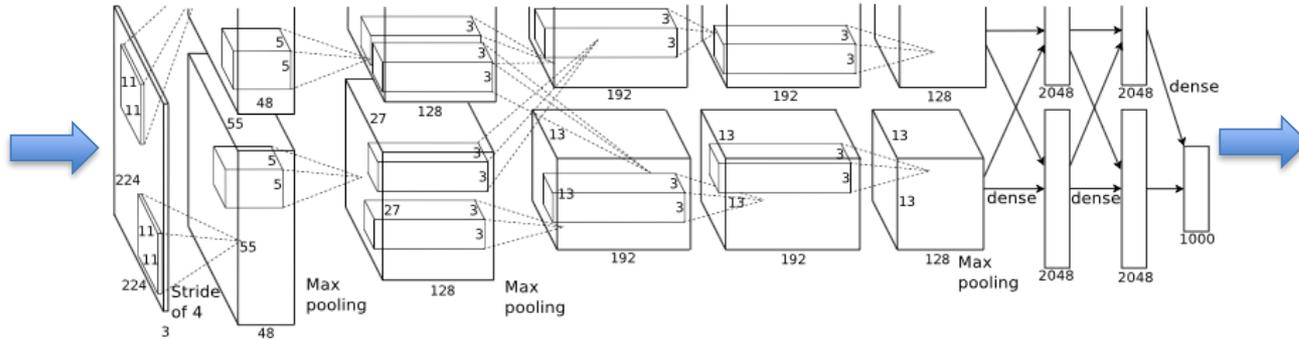
Output:
score for
each
category

[AlexNet](#)

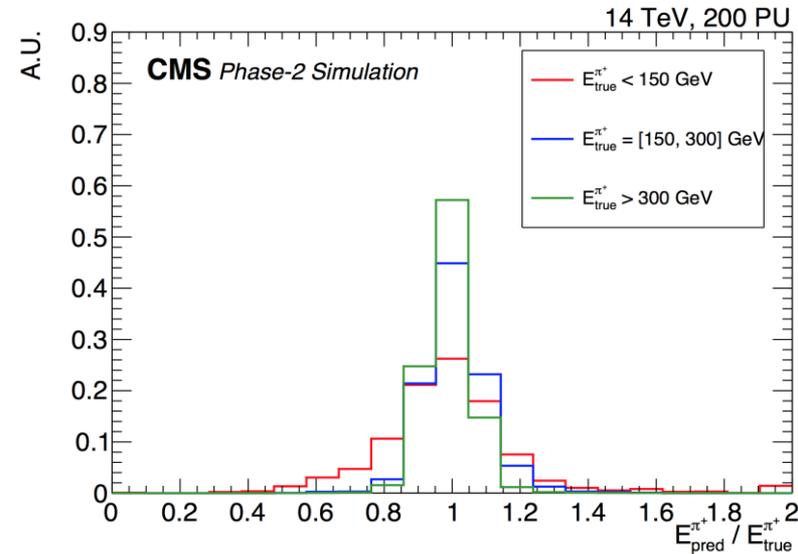
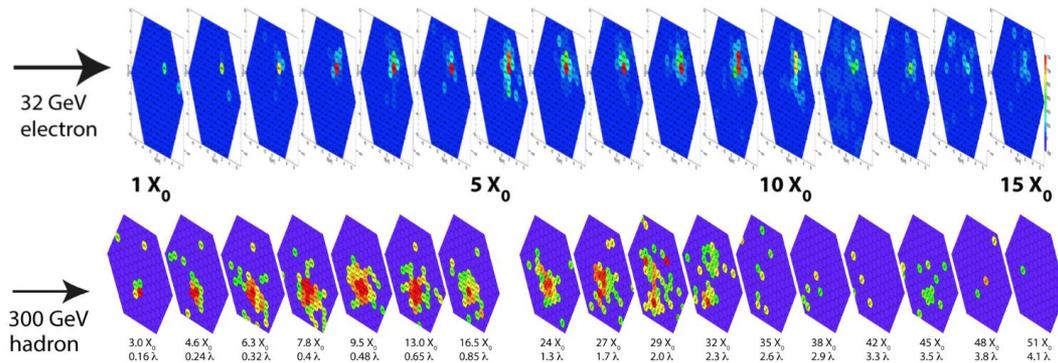
			
mite	container ship	motor scooter	leopard
<ul style="list-style-type: none"> mite black widow cockroach tick starfish 	<ul style="list-style-type: none"> container ship lifeboat amphibian fireboat drilling platform 	<ul style="list-style-type: none"> motor scooter go-kart moped bumper car golfcart 	<ul style="list-style-type: none"> leopard jaguar cheetah snow leopard Egyptian cat
			
grille	mushroom	cherry	Madagascar cat
<ul style="list-style-type: none"> convertible grille pickup beach wagon fire engine 	<ul style="list-style-type: none"> agaric mushroom jelly fungus gill fungus dead-man's-fingers 	<ul style="list-style-type: none"> dalmatian grape elderberry ffordshire bullterrier currant 	<ul style="list-style-type: none"> squirrel monkey spider monkey titi indri howler monkey

Deep Learning for HGCAL reconstruction

Input:
raw image



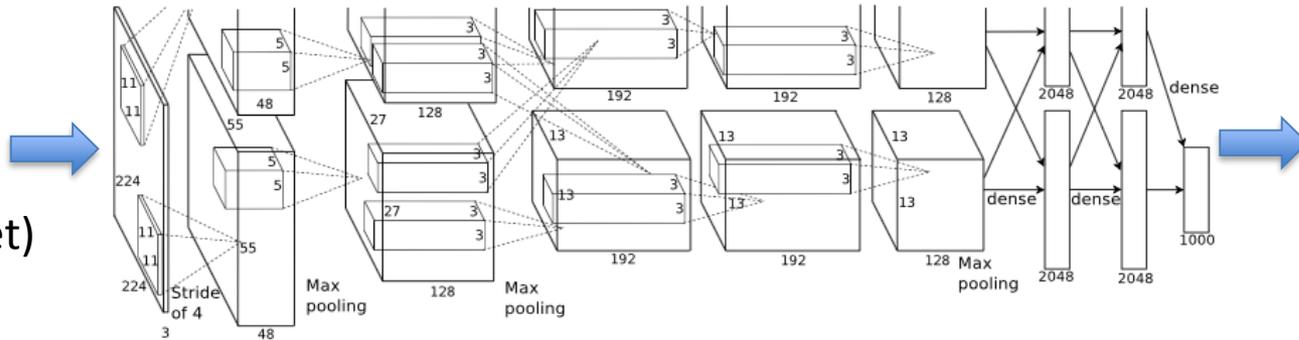
Output:
shower energy



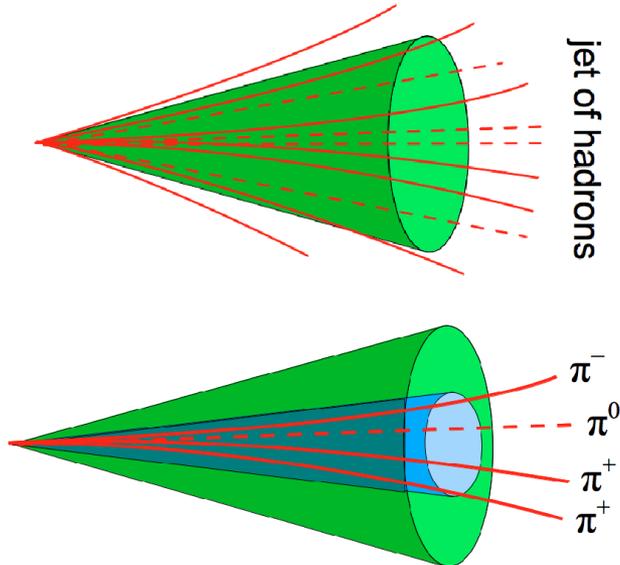
Deep Learning on PF output

Input:
raw jet
(e.g. 1st 100
particles in jet)

or other
collection of
particles



Output:
score for
each
category
or
regression
value



e.g. :

- τ identification
- b jet tagging
- jet energy correction
- MET

Our advantage:

Monte Carlo Simulation

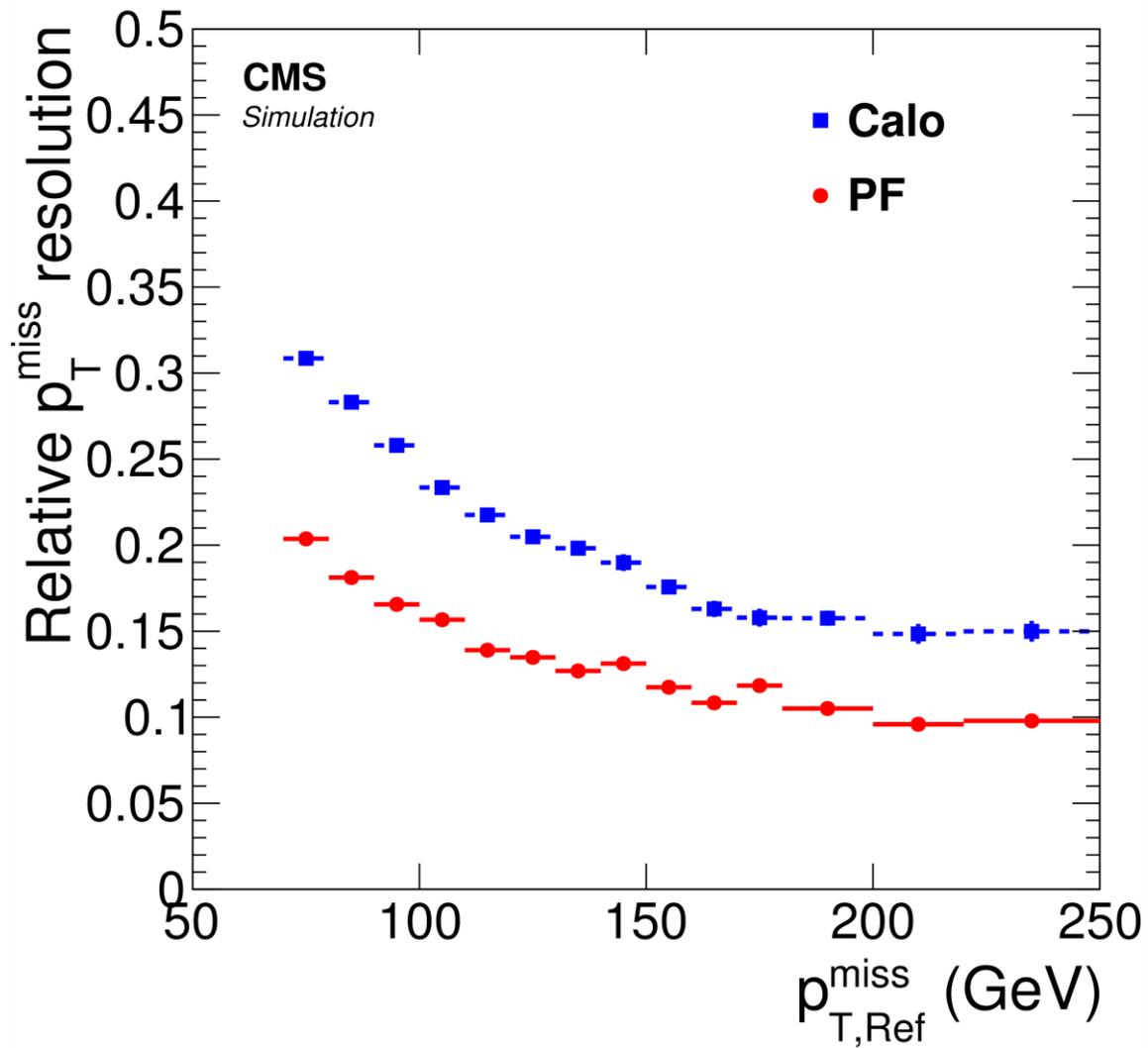
- large number of events
- we know the truth
(NN target)

Conclusion

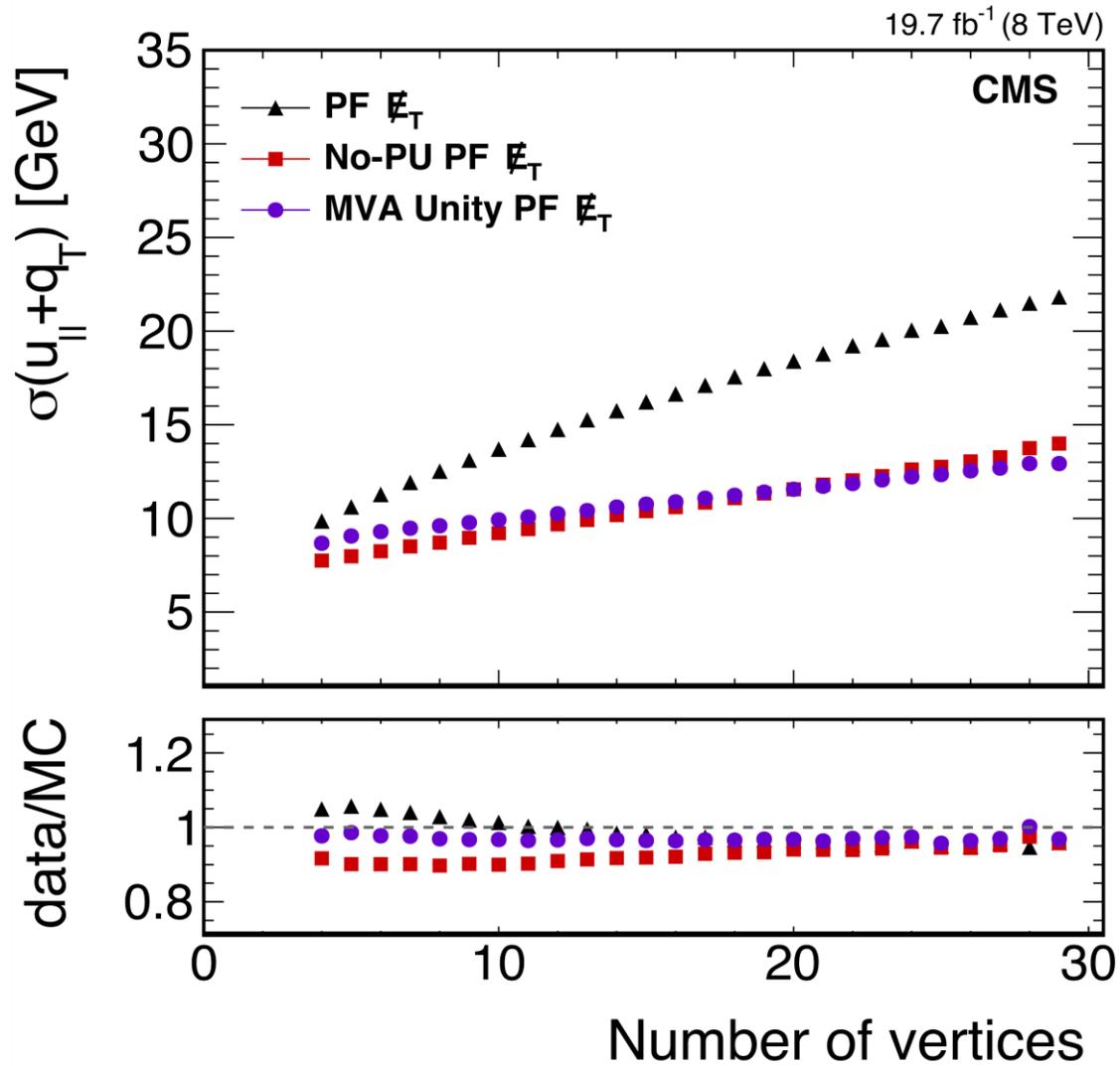
- The foundation of 99.5% of our analyses
 - all physics objects come from PF
- > 50 000 lines of code
 - to deal with the gory « details » :
 - fake tracks, muons, e/γ , secondary interactions, noise, ...
- Algorithm unchanged since 2009
 - 0 \rightarrow 45 pileup collisions
- But HL-LHC data will be a big challenge

Backup

MET



MVA MET



Cells

