

Review of CMS HCAL reconstruction performance

Hui Wang (王徽) Nanjing Normal University CLHCP2024, Qingdao

- Introduction
- Reconstruction algorithms
- Reconstruction performance
- Reconstruction with ML

HCAL structure

Reconstruction algorithms

HCAL Energy Reconstruction

- Reco input: digitized charge in 8 LHC bunch crossings (BX) in buffer, called time samples
	- Current BX (BX0): 75-100 ns (Time

sample 3) ~60% total charge

- $BX+1:$ ~20% total charge
- First reco algorithm: Method 0
	- Used in Run1 (50 ns bunch spacing)
	- OOT PU almost negligible
	- $(Q_{\text{BXO}} + Q_{\text{BX+1}})$ x scale factors
- Pulse fitting algorithms
	- In use since Run2 (25 ns bunch spacing)
	- 2016-2017: Method 2 (3) offline (HLT)
	- from 2018: MAHI both offline and HLT

Method 2

- M2 estimates the energy of SOI pulse by minimizing χ^2 using MIGRAD algorithm in Minuit
- Fits up to 3 pulses (SOI 1, SOI and SOI + 1) to QIE digis in 10 TS
- Starts with fitting 1 pulse. If $\chi^2 > 15$ and charge < 100 fC for HPD or 25000 fC for SiPM (both correspond to ~20 GeV), then switches to 3 pulses

 A_i : QIE digi in ith TS μ_i : sum of fitted amplitudes in ith TS $\sigma^2_{p,i}$: quadratic sum of uncertainties (pedestals, QIE granularity, and photostatistics)

 t_i : pulse arrival time

ped: floating baseline

Method 3

- M3 was developed to meet HLT timing requirment
- Compared to M2, M3:
	- Fits 3 pulses (SOI 1, SOI and SOI + 1) to only 3 TS
	- Drops the arrival time term
	- Uses constant baseline term
	- Fitting \rightarrow solving linear equations A_i : QIE digi in ith TS

 $f_{0,1,2}$: pre-measured fractions of the pulse template in +0, +1 and +2 TS, respectively μ_i : amplitudes of ith pulse

B: constant baseline (average of pedestals in

all TS except SOI and SOI+1)

MAHI

 $^{\prime\prime}$ Orma

- MAHI (Minimization At HCAL, Iteratively) estimates the energy of SOI pulse by minimizing χ^2 in an iterative approach, using Non-Negative Least Square (NNLS) algorithm instead of MIGRAD in M2
- Reconstruction speed: MAHI is O(10) faster than M2 and O(10) slower than M3

$$
\mathbf{V} = \sum_{j=0}^{7} \mu_j^2 \mathbf{D}_j^{\text{pulse}} + \mathbf{D}^{\text{noise}}
$$

 μ_i : amplitudes of jth pulse D_j^{pulse} : pulse shape uncertainty D^{noise} : total noise (pedestals, QIE

granularity, and photostatistics)

$$
\chi^2 = \left[\sum_j \vec{P}_j \mu_j - \vec{d}\right]^T \mathbf{V}^{-1} \left[\sum_j \vec{P}_j \mu_j - \vec{d}\right]
$$

 $\overrightarrow{P_i}$: 8x8 matrix contains pulse template \vec{d} : vector contains QIE digis of 8TS

Reconstruction performance

Charged pion resolution in data

- Extrapolate isolated tracks to calorimeter and match to a cone
	- Use track momentum ECAL energy in that cone as "truth" HCAL energy
- MO, M2 and MAHI have similar resolutions, but M0 has high response because of OOT-PU

Response of pions in MC

- Two MC samples from the same GEN step pion gun
	- One has only OOT-PU
	- The other has no PU
- Extrapolate GEN pion tracks to calorimeter and match to a cone
- Response = cone energy / GEN pion energy
- Plot ratios of responses in OOT-PU sample and no PU sample
- Performance: M2/MAHI better than M0, especially in low energy / high eta regions, because of OOT-PU subtractions

- Select events with a well reconstructed Z boson
- Project MET to Z pT, and measure the resolutions of its parallel and perpendicular components
- Performance: M2/MAHI better than M3 than M0

Reconstruction with ML

Limitation of analytical algorithms

- Reconstructed energy resolution in each channel
	- MAHI: not fitting pulse arrival time Bad performance at high energy
	- M2: too slow only fits up to 3 pulses Bad performance at low energy
- Is there an algorithm that has better resolution at both low and high energy?
- Machine learning can achieve this!

DLPHIN

- Deep Learning Processes for HCAL INtegration
- Novel architecture based on 2D CNN
	- Dim. 1: digitized charge in 8 BX
	- Dim. 2: depth \rightarrow exploit correlations among channels in an HCAL tower
- More than 3 times faster than MAHI
- Better perform from upstream to downstream Channel-level \rightarrow single particle-level \rightarrow jet-level
- Will benefit almost all physics analyses

DLPHIN performance

Backup Slides

Trigger System and Pileup

- Two-level trigger system
	- Reduce the event rates from 40 MHz to ~1kHz
	- While keeping most of the interesting events
- Level-1 trigger (L1T)
	- Custom ASIC, FPGA, etc
	- Reduce rate to 100 kHz
- High-level trigger (HLT)
	- Commercial CPU + GPU
	- Rate reduce to ~1k Hz

pileup (PU)

- In-time PU: current bunch crossing (BX)
- Out-of-time PU: other BX, very important for calorimeter reconstruction

Event Reconstruction

- Particle Flow (PF) Algorithm
	- Runs on HLT and offline reconstruction
	- Synthesizes information from all subdetectors and reconstructs particles based on their signatures
		- 1. Muon
		- 2. Electron and Photon
		- 3. Charged and Neutral Hadron
- Then PF particles are clustered as jets
	- Usually anti- k_T algorithm in CMS
- Last global quantities of an event
	- e.g. missing transverse momentum p_T^{miss} , aka MET usually a manifest of neutrinos, but may also from BSM :P

Parton Shower vs Hadron Shower

r= Molière radius

Parton shower (+ hadronization that form a jet) typically in a cone

Shower length

Hadron shower (interacting with detector material)

typically in a cylinder

- longitudinal development: radiation or interaction length
- lateral development: Molière radius (a cylinder containing on average 90% of the shower's energy deposition)
- Typical Molière radius for a pion is an HCAL tower (0.087 x 0.087 rad.)