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Performance of electrons and photons at CMS and recent developments



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Critical for the success of the CMS physics program, both in terms of standard model measurements and searches



Run2 H $\rightarrow \gamma \gamma$ production cross section and couplings

At ICHEP 2024:

- \square H $\rightarrow \gamma\gamma$ and H $\rightarrow 4\ell$ cross sections and couplings in Run3 (see <u>Jan Lukas' talk</u>)
- H boson mass and width (see <u>Badder's talk</u>)
- Several searches with photons in the final state (see <u>Rocco's talk</u> for SM, <u>Jyoti's talk</u> for BSM)
- Compressed SUSY with electrons in the final state (see Margaret's talk)

etc.



Electrons and photons at CMS – how to







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Bremsstrahlung and photon conversions need to be associated to the same initial particle

\rightarrow Supercluster



Electrons and photons at CMS – how to

CMS-DP-2022/015



Bremsstrahlung and photon conversions need to be associated to the same initial particle

→ Supercluster



Geometrical procedure that considers the slight bending in η of the low- $p_{\rm T}$ constituents of EM showers \rightarrow **moustache**. The moustache was retuned for Run3 to reduce the noise. 5



Photons at CMS – how to





Energy deposit in the ECAL

 \rightarrow photon



Electrons at CMS – how to

CMS-DP-24/052



Energy deposit in the ECAL + track association

\rightarrow electron/positron

Using **special track reconstruction algorithm** (Gaussian-sum filter) to consider energy losses via photon radiation. <u>CMS Note 2005/001</u>





Online reconstruction performance

CMS-DP-23/015 CMS-DP-24/041



The CMS trigger system selects the collision events to save for further analysis. It is two-tiered:



L1: hardware-based, using on-detector electronics. Saves ~ 110 kHz.



HLT: software. Runs a simpler (\rightarrow *faster*) version of the offline reconstruction, allowing us to access data quality and physics content "online", during collisions. Saves $\sim 1.5 \text{ kHz}$.





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High-purity single electron trigger

- Excellent HLT performance in Run3
- Great stability vs pileup
- Backed by good e/γ energy resolution at L1
- Supercluster energy response corrected using a BDT-based regression





HLT scouting: allow larger rate at low- $p_{\rm T}$ at the price of giving up the offline reconstruction.

(Can we produce competitive physics results using objects reconstructed online only and with limited event content information? <u>Run2 answer</u>)



- Large fraction of L1 rate in input to the scouting streams dedicated to single and double e/γ .
- Minimal HLT selection (on supercluster) and thresholds: \circ 30 GeV for single e/ $\gamma \rightarrow \sim$ 9.0 kHz HLT (2023)
 - **12 GeV** for double $e/\gamma \rightarrow \sim 0.5$ kHz HLT (2023)



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arXiv:2403.16134

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- Considerably reduce $p_{\rm T}$ thresholds wrt standard HLT, especially for single photon (200 GeV in 2022 and 2023).
- Additional low $p_{\rm T}$ electrons and photons saved in events recorded using triggers seeded by L1 muons, jets or $H_{\rm T}$.





The **offline reconstruction** is more refined than the online one and allows for further data reprocessing with improved corrections and detector calibration.

For electrons and photons, this means:

- More "layers" of energy corrections (and potentially more complex models \rightarrow timing is not a concern)
 - Notably, combination of ECAL energy and tracker momentum measurements to determine the final electron energy, which does not happen at HLT
- More sophisticated IDs
 - Cut-based
 - Flexible easy to remove unwanted cutoffs at analysis level (e.g., isolation)
 - MVA-based (BDT or DNN)
 - Generally stronger signal/background discrimination



Run3 IDs

In Run3, the cut-based and MVA-based electron and photon IDs were tuned or retrained to ensure a better performance than the Run2 ones within the Run3 data-taking and detector conditions.



MVA-based photon IDs





<u>CMS-DP-24/041</u>

CMS-DP-24/052

The reconstruction, identification, and trigger performance for electrons and photons is studied in collision data and simulation, and we correct them for discrepancies.





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Scale and resolution corrections

- Derived using electrons from $Z \rightarrow ee$ for different:
 - particle's $p_{\rm T}$ or ECAL amplifier gains
 - detector regions
 - data acquisition runs
 - bremsstrahlung emission
- $\circ~$ Scale e/ γ energy in collision data
- $\circ~$ Spread e/ γ energy in simulation





An accurate efficiency measurement for low $p_{\rm T}$ (< 20 GeV) e/ γ requires a standard candle different from the Z resonance

 J/ψ decays provide significantly higher statistics and make the measurement possible even below 10 GeV

• *Tag & Probe*: signal peaks clearly visible in both passing and failing probe m(ee) distributions.





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 J/ψ s electrons tend to be less isolated than Z electrons

• Extrapolate efficiency at the Z relative isolation value from a fit to the trend of the $J/\psi \rightarrow ee$ efficiency for different relative isolation bins.





ONGOING STUDY

Supercluster

- Base component in the reconstruction of photons and electrons
- Essential for the ECAL energy response calibration
- One of the inputs to the <u>Particle Flow</u> global event reconstruction

Moustache

 \circ selects clusters defining a parabolic $\eta - \phi$ region parametrized by the seed position and the cluster transverse energy

very efficient but subject to PU/noise contamination – resolution degrades with PU

GNN supercluster \rightarrow "DeepSC"

 uses small clusters and single crystal hits in a window around the seed

 filters noise/PU on a cluster-by-cluster basis, improving the "raw" resolution





> Build detector windows around each cluster with $p_{\rm T} > 1 \text{ GeV}$ (seed).







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- Apply a cut of p_T > 4 GeV to remove supercluster formed mainly by detector noise and very low energy pileup.







Energy scale

ONGOING STUDY

comparable to the moustache



Energy resolution

improvements observed across the detector, in particular in the most challenging regions



Region with the largest tracker material budget in front of the ECAL → largest secondary emissions





Energy loss in the tracker Leakage into the HCAL



Clustering inefficiency Finite noise thresholds





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DRN energy regression



• Allows for variable size input

ONGOING STUDY

- Can deal with unordered inputs
- Can handle complicated geometries
- Can handle representation in 4D (x, y, z, energy)



a few higher-level variables (median energy density, H/E) ONGOING STUDY

CMS

DRN energy regression





1. Hits are mapped to a multidimensional space

- 2. GNN learns high-level characteristics from the hits by grouping them based on their similarities
- 3. The new object is passed through a further NN which maps the learned features to the regression target



a few higher-level variables (median energy density, H/E)







- Excellent electron and photon reconstruction performance in Run3 offline and online
 - ✓ Online, the "scouting" trigger extends the reach to lower momentum phase spaces wrt the standard trigger
 - ✓ Can be an asset for several statistically limited searches
- Electron and photon IDs were tuned or fully retrained in Run3
 - \checkmark Comparable or better identification performance than in Run2
- Several types of calibrations and corrections are adopted to ensure the best precision
 - ✓ New methods have been studied that can improve the effectiveness of such corrections, or make the performance measurements more robust or even possible in certain phase spaces (e.g., at low $p_{\rm T}$)
- Several ongoing studies using novel techniques are paving the way to more efficient electron and photon reconstruction and more robust energy measurements

ADDITIONAL MATERIAL



The goal of the scale and smearing corrections is to reduce the residual differences in the electron and photon energy scale and resolution between collision data and simulation. They are derived using electrons from $Z \rightarrow e^+e^-$ and they:

- scale the electron energy in data: $M_{ee}^{scaled} = M_{ee}\sqrt{(1 + \Delta P_{e1})(1 + \Delta P_{e2})}$, where ΔP_{ei} is the scale shift with respect to MC for electron "i";
- smear the electron energy in the simulation: $M_{ee}^{smeared} = M_{ee}\sqrt{Gaus(1, \Delta C_{e1})Gaus(1, \Delta C_{e2})}$, where ΔC_{ei} is the additional smearing for electron "i".

The variables ΔP_{ei} and ΔC_{ei} are determined by minimizing a global binned negative log-likelihood of the invariant mass for each di-electron category. The categories are defined by subdividing events according to the data acquisition run number, the ECAL amplifier gain, and the electron η , p_{T} , and R_{9} .

The derivation of scale and smearing corrections for photons uses electrons from Z boson decay reconstructed as photons by the CMS software, and to which the photon energy regression was applied.



Can it (easily) handle	BDT	MLP	CNN	RNN	GNN
Variable-size input	X	X	\checkmark	\checkmark	
Complicated geometries	\checkmark	\checkmark	X	\checkmark	\checkmark
4D inputs	\checkmark	\checkmark	X	\checkmark	\checkmark
Unordered inputs	X	X	\checkmark	X	\checkmark



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Measuring efficiencies at low pT

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 Studied a method to make reliable measurements of ID efficiencies when the IDs contain isolation requirements



- I. Measure the electron ID efficiencies from J/ψ events in different $(p_{\rm T}, \eta, relative isolation)$ bins.
- II. Fit the trend of efficiency measured in step I. vs relative isolation for each $(p_{\rm T}, \eta)$ bin.
- III. Take the efficiency value obtained by interpolating the measurements done in step I. with the fit done in step II. at the average Z MC relative isolation value in that particular ($p_{\rm T}$, η) bin.