

Implementation of π/γ separation in SPDROOT

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Outline

Motivation

Design of the neural network for π/γ separation

Implementation design

Performance and tests

Conclusions

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Motivation

Prompt photons:

- ▶ interested in $p_T > 3\text{-}4$ GeV, reject background from π^0

Online polarizability measurement:

- ▶ select high-purity π^0 sample

Idea:

- ▶ separate clusters produced by a photon and neutral pion, where the two decay photon showers merged into one cluster, based on the shape of the cluster;
- ▶ here, cluster means a group of neighboring cells, where in each cell the deposited energy is above threshold.

Input parameters

- ▶ X/Y for endcaps or Z/ϕ for barrel, inputs shown in red

Energy distribution

- ▶ S_1, M_2 – cells with first and second largest energies
- ▶ S_9, S_{25} – sum of energies in $3 \times 3, 5 \times 5$ regions around cell with highest energy
- ▶ S_6 – maximum energy in 3×2 region containing both first and second largest energy cells

$$\frac{S_1}{S_9}, \frac{S_9 - S_1}{S_{25} - S_1}, \frac{M_2 + S_1}{S_4}, \frac{S_6}{S_9}, \frac{M_2 + S_1}{S_9}, \frac{S_4}{S_{25}}$$

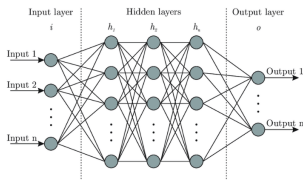
Size/shape

- ▶ $|x_{cog}|_{25} = \left| \frac{\sum_{i=1}^{25} E_i X_i^{rel}}{S_{25}} \right|$,
 $|y_{cog}|_{25} = \left| \frac{\sum_{i=1}^{25} E_i Y_i^{rel}}{S_{25}} \right|$
- ▶ $S_{\alpha\beta} = \frac{\sum_{i=1}^N e_i (\alpha_i - \alpha_c)(\beta_i - \beta_c)}{\sum_{i=1}^N e_i}$,
 $\alpha, \beta : X, Y$
- ▶ $\rightarrow S_{XX}, S_{YY}, S_{XY}$
- ▶ $r^2 = \langle r^2 \rangle = \frac{S_{XX} + S_{YY}}{\sum_{i=1}^N e_i ((x_i - x_c)^2 + (y_i - y_c)^2)}$
- ▶ $\kappa = \sqrt{1 - 4 \frac{S_{XX} S_{YY} - S_{XY}^2}{(S_{XX} + S_{YY})^2}} = \sqrt{1 - 4 \frac{\det S}{\text{Tr}^2 S}}$
- ▶ Angle θ of incidence

Topology of the network

Attempt at using a more complex NN

Inspired by the work of Dimitrije Maletic (thanks!) and <https://cds.cern.ch/record/2042173>



$$O_i = f(W_{i0} + \sum_{j=1}^N W_{ij} O_j) \rightarrow \text{weighted sum + bias for each node}$$

- **f: ReLU** $f(x) = \begin{cases} 0 & \text{for } x < 0 \\ x & \text{for } x \geq 0 \end{cases}$ sigmoid for output: $f(x) = \frac{1}{1 + e^{-x}}$

- **Dropout** ($p=0.1$),
- **batchnorm** for each layer (before activation)

- Binary cross entropy loss (**BCE**):

$$H_p(q) = -\frac{1}{N} \sum_{i=1}^N y_i \cdot \log(p(y_i)) + (1 - y_i) \cdot \log(1 - p(y_i))$$

- Optimizer: **Adam**
(stochastic gradient descent +
adaptive moment estimation)
($\text{lr} = 0.001$, $\beta_1 = 0.9$, $\beta_2 = 0.999$, $\epsilon = 1e-8$)

2 hidden layers, 64 neurons each

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SpdEcalRCParticle

- ▶ SpdEcalRCParticle class → fPID field: 22 (γ)/111 (π^0)

SpdEcalPiGammaSeparator

- ▶ idea: create a separate class which contains:
 - ▶ function to predict particle ID from pre-calculated input parameters (implementation of the NN)
 - ▶ array of parameters of the NN as static members of the class
 - ▶ interface that accepts SpdEcalRCCluster
- ▶ also needs some geometry parameters as input (number of cells in barrel/endcap)
- ▶ single instance is stored in SpdEcalRCMaker (reconstruction algorithm)

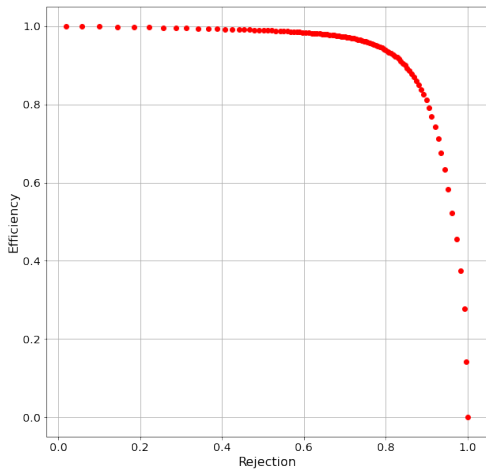
SpdEcalRCMaker

Class for cluster reconstruction is mostly unchanged, except:

- ▶ initializes SpdEcalPiGammaSeparator in the Init() step
- ▶ added one more function to predict particle ID (γ / π^0)

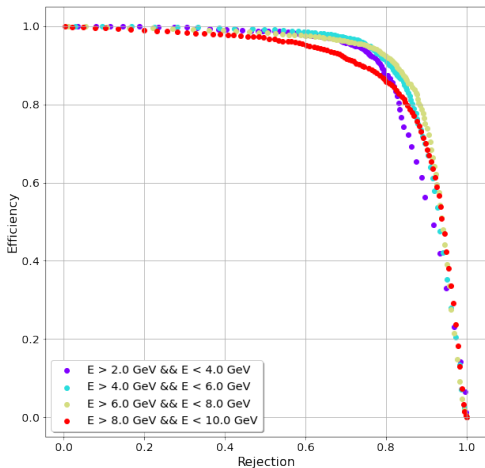
Efficiency and rejection

All angles and energies



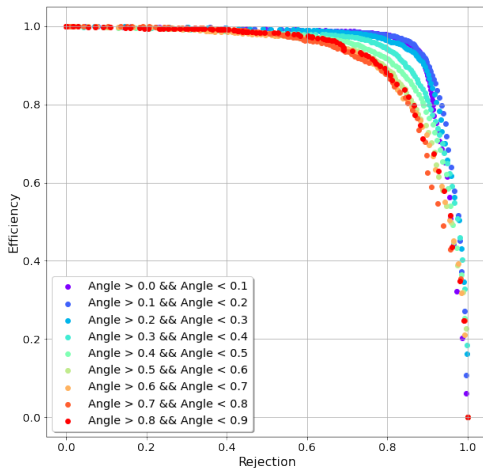
Efficiency and rejection

Efficiency/rejection for different energies



Efficiency and rejection

Efficiency/rejection for different angles

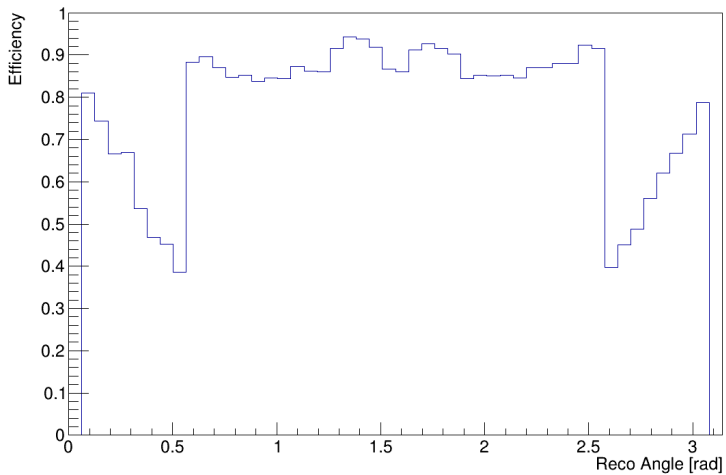


Test conditions

- ▶ π^0/γ generated isotropically;
- ▶ energies of particles from 0 to 10 GeV (discrete distribution);
- ▶ cell energy threshold is 20 MeV;
- ▶ identified as a photon if both conditions are true:
 - ▶ only one cluster in event;
 - ▶ the cluster is tagged as a photon;
- ▶ identified as a pion if either:
 - ▶ two clusters in event;
 - ▶ one cluster, tagged as a pion.

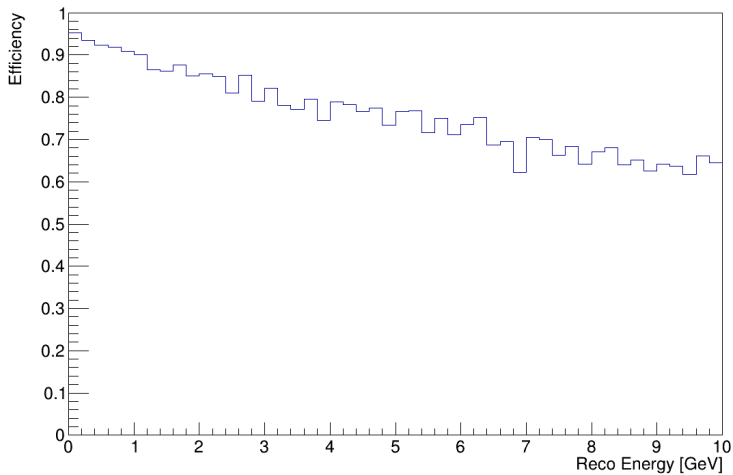
γ identification

Gamma ID efficiency: events with 1 cluster, PID=22

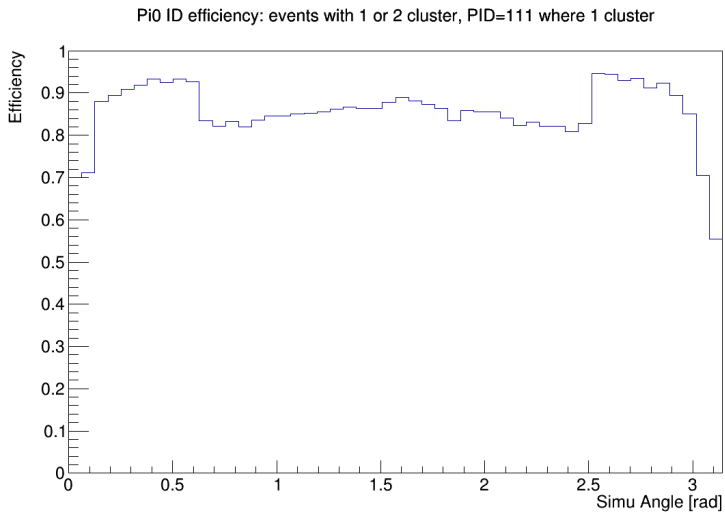


γ identification

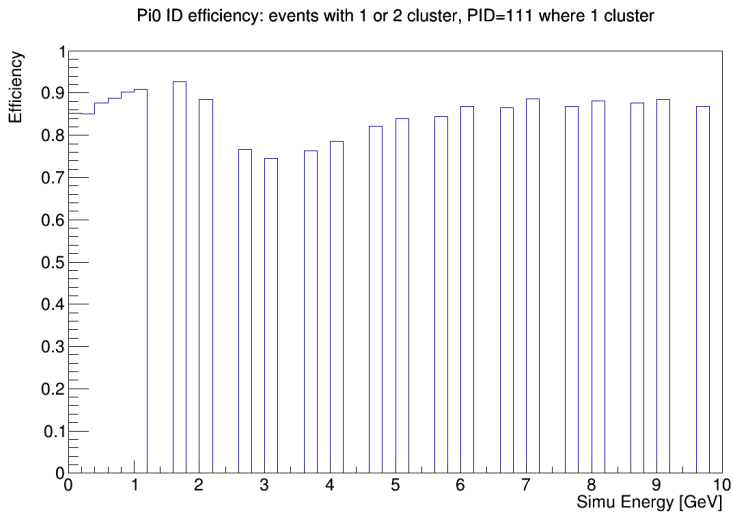
Gamma ID efficiency: events with 1 cluster, PID=22



π^0 identification

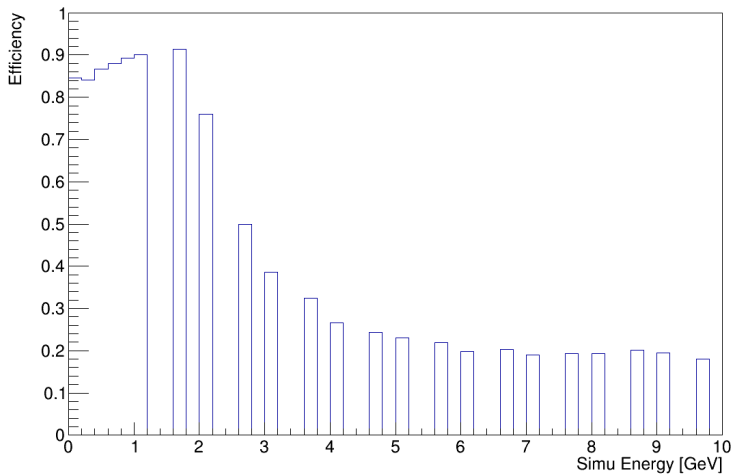


π^0 identification



π^0 identification

Pi0 ID efficiency: events with 2 cluster



Conclusions

- ▶ After being implemented into SPDROOT, the separation network gives 70-90% photon detection efficiency (depending on energy) and 80-90% pion detection efficiency
- ▶ Bad performance in the vicinity of the gap between barrel and endcap

Future steps

- ▶ Add the macros to produce pi/gamma database and train the network to git
- ▶ Bridge clusters in barrel and endcap
- ▶ Investigate the possibility of using R-CNN for cluster reconstruction