

# Calorimetry Meeting: July 13, 2017

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# Omega decay

- $\gamma p \rightarrow \omega p$

- $\omega \rightarrow \pi^+ \pi^- \pi^0$   
 $\rightarrow \pi^+ \pi^- \gamma \gamma$

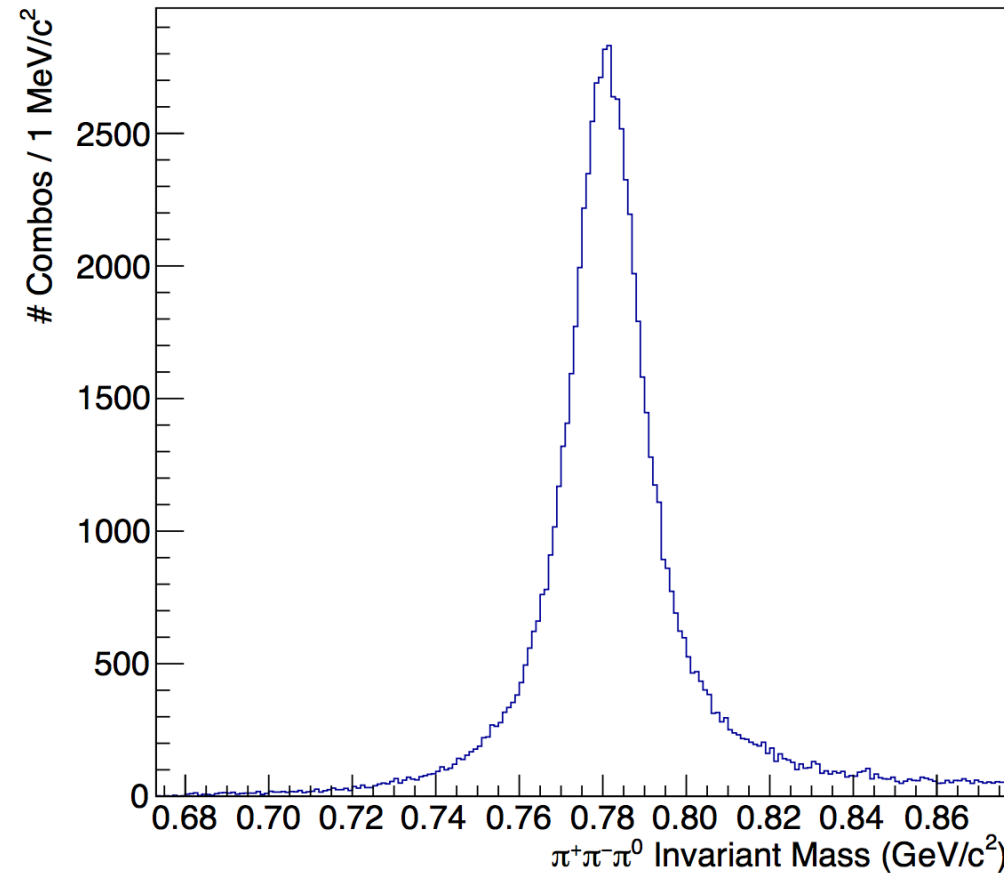
# Definitions

- Type 0 showers:
  - Photon showers originating from a  $\pi^0$  decay
- Type 1 showers:
  - Showers originating from a charged particle for which there is a track
  - Easy to identify by simple geometry
- Type 2 showers:
  - All other showers
  - Noise, split-offs from Type 1 showers, etc.
  - These cause backgrounds for true photons

# Project Background

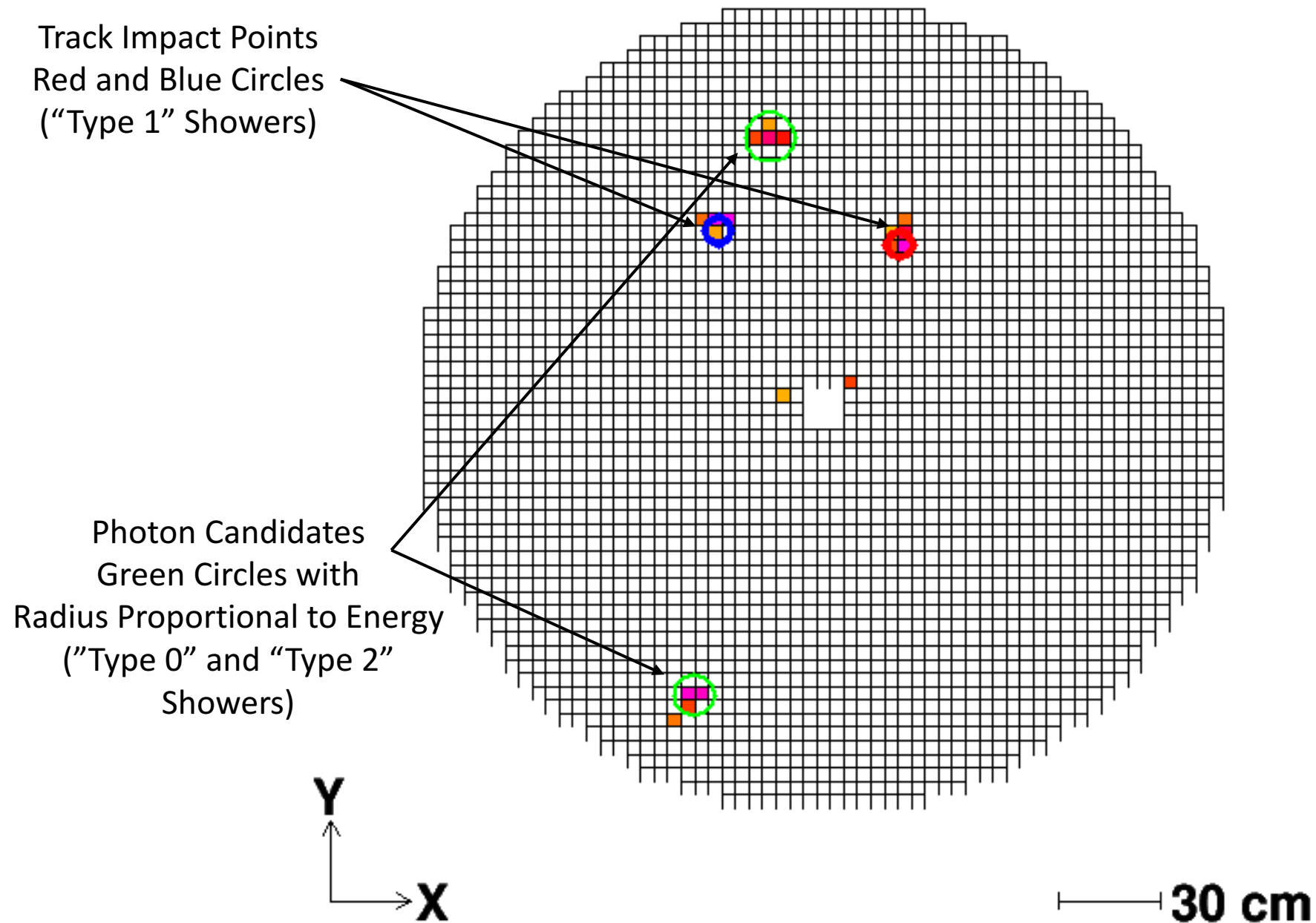
- **Goal:**  
Algorithmically differentiate between Type 0 and Type 2 showers
- **Zeroth Stage:**  
Create a set of data cuts to isolate omega decays to identify Type 0 showers
- **First Stage:**  
Isolate variables that can be used to differentiate between Type 0 and Type 2 showers
- **Second Stage:**  
Identify algorithms that can use these variables to create clear distinctions
- **Third Stage:**  
Implement algorithm to differentiate between shower types

# Data cuts on $\omega \rightarrow \pi^+ \pi^- \pi^0$

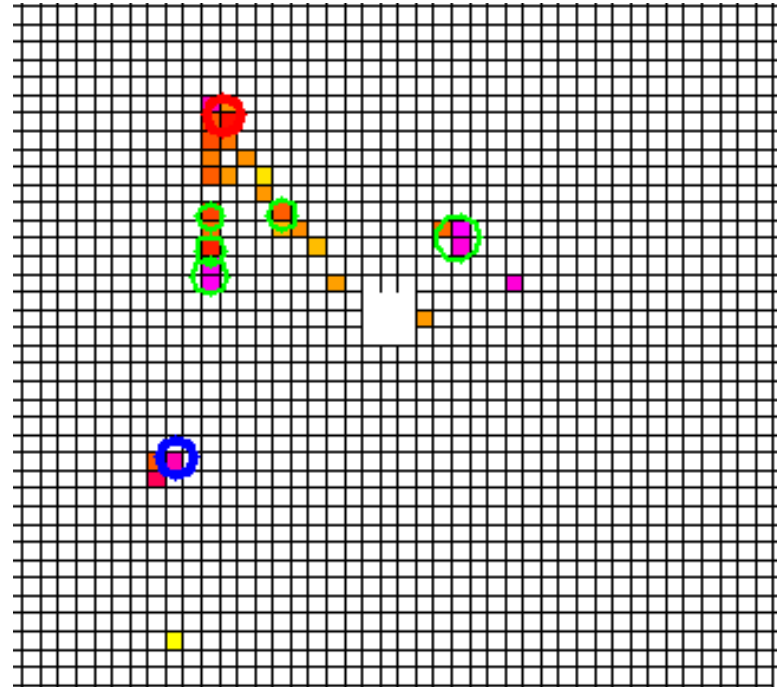
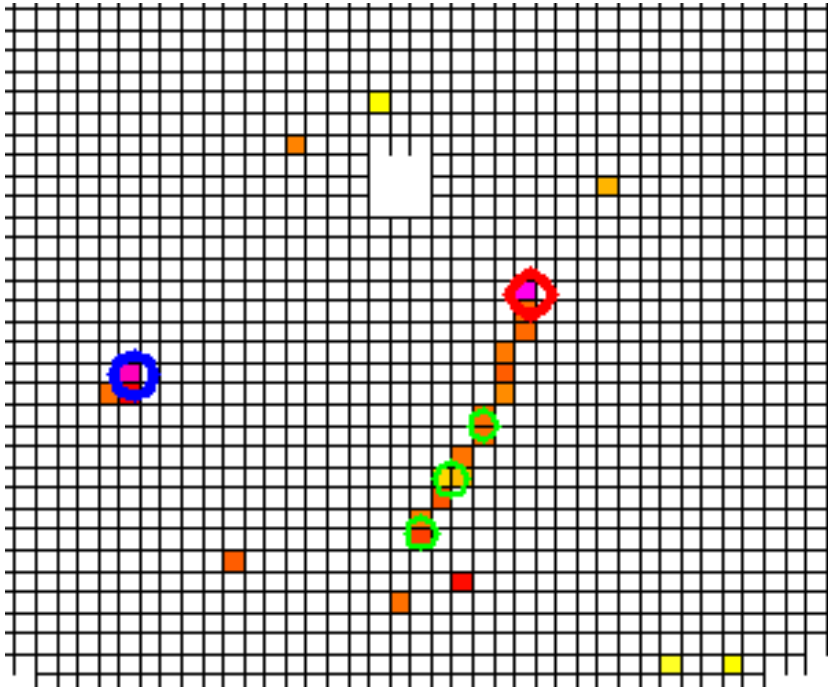


Cuts made on kinematic chi-squared value, beam energy

# FCAL view from downstream looking upstream

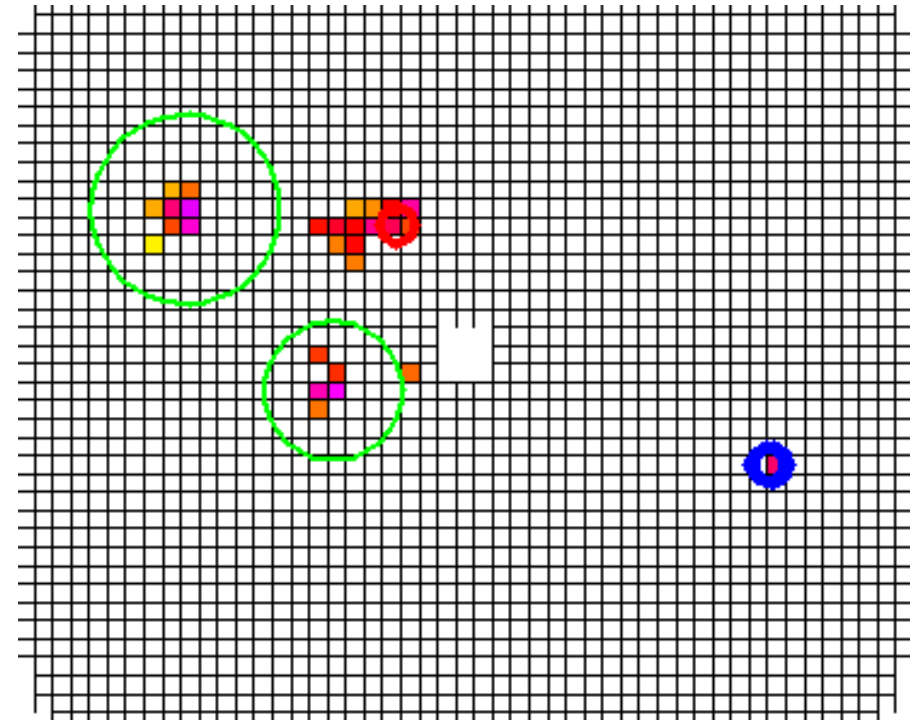
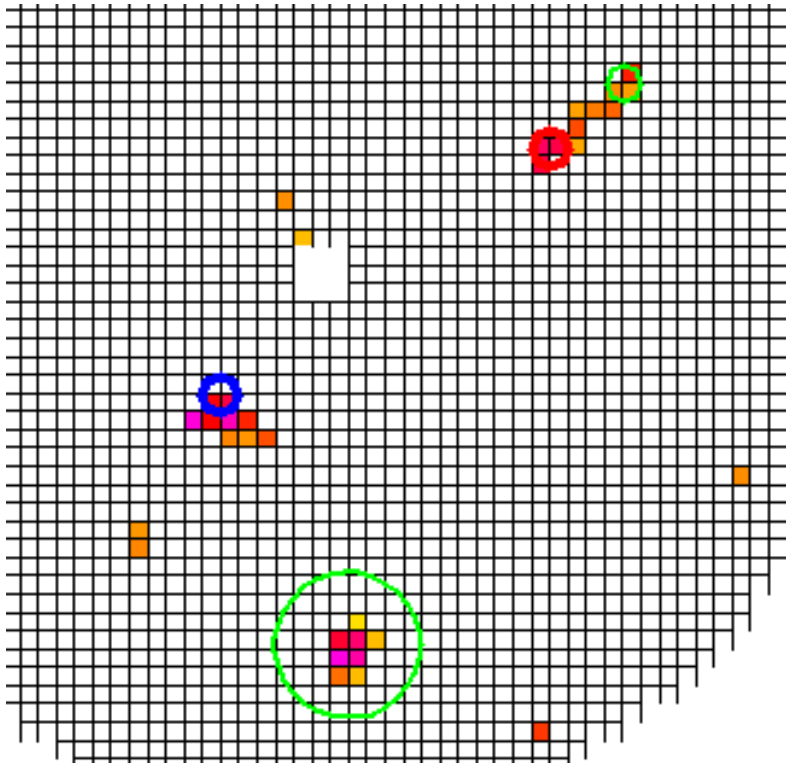


## Shower examples: Type 2



Photon candidates emanating from charged track impacts

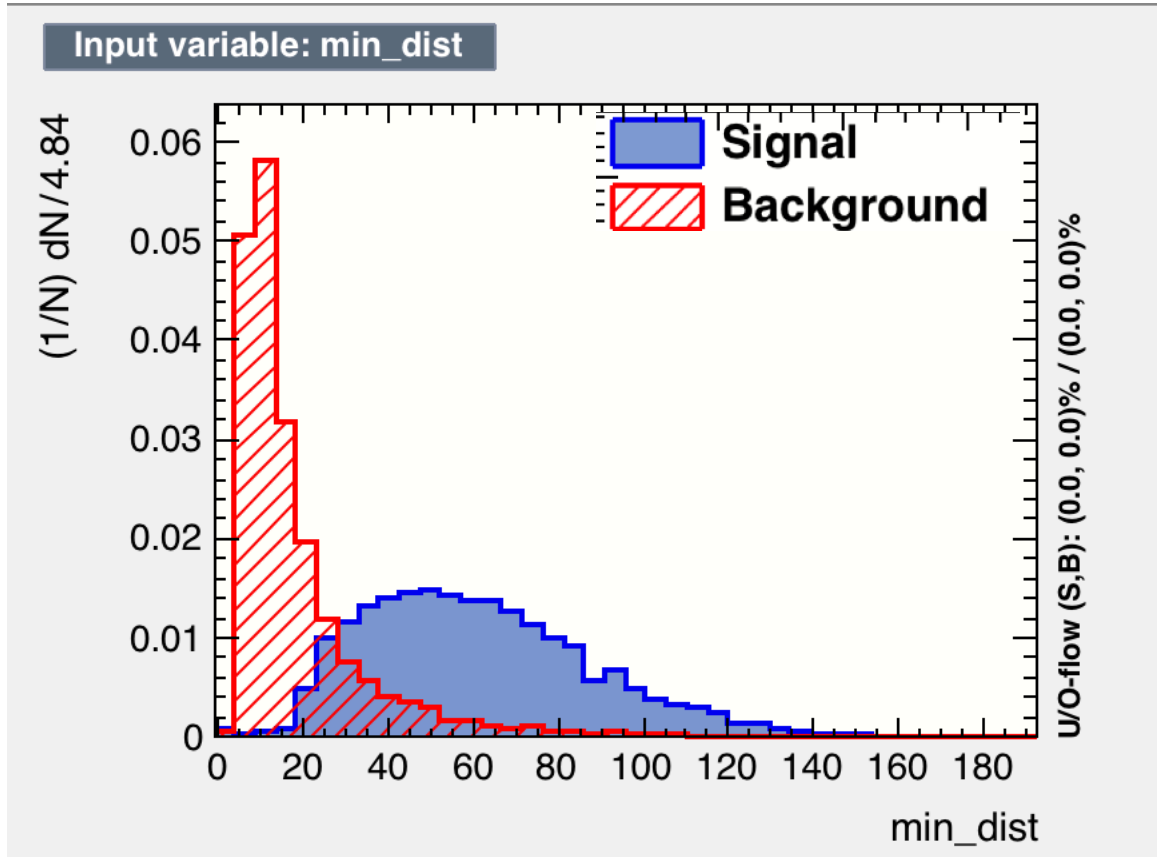
# Shower examples: Type 0



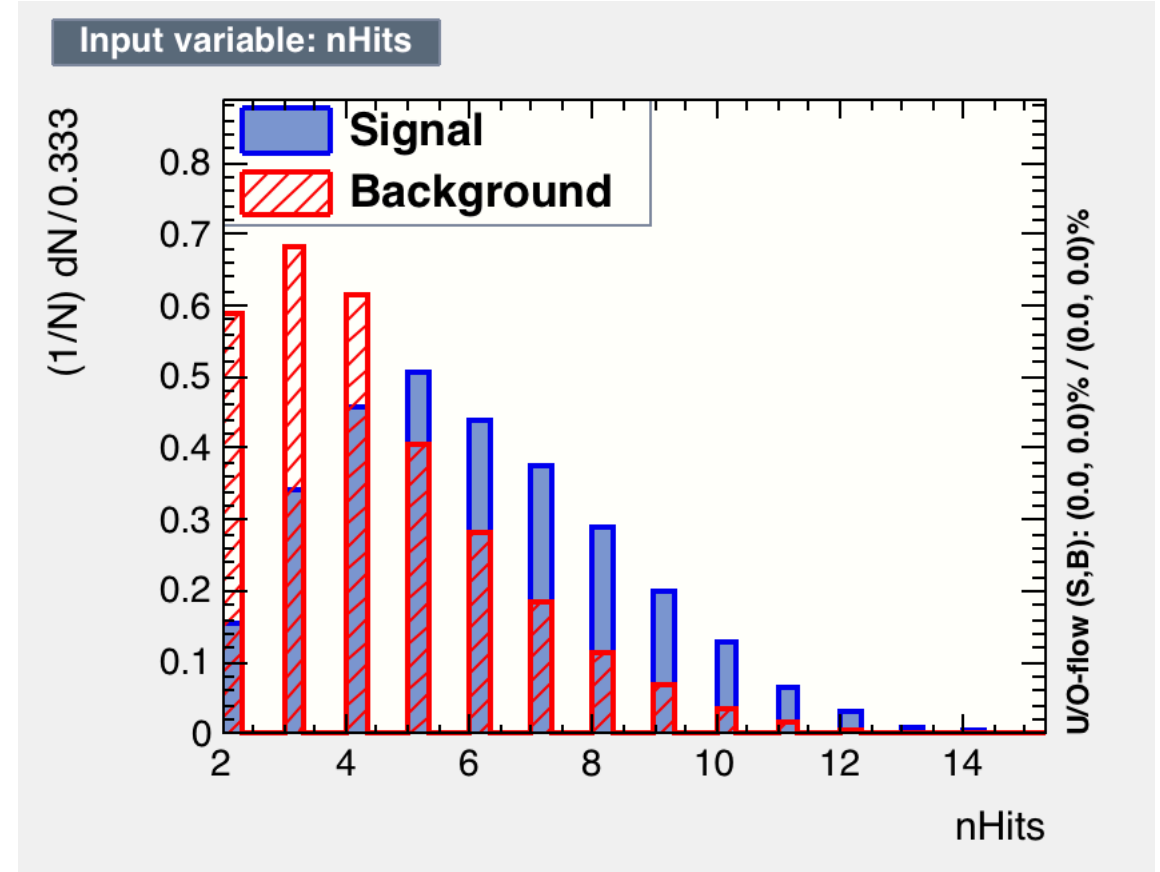
True photon showers: compact in size and more uniformly distributed



# Input variables

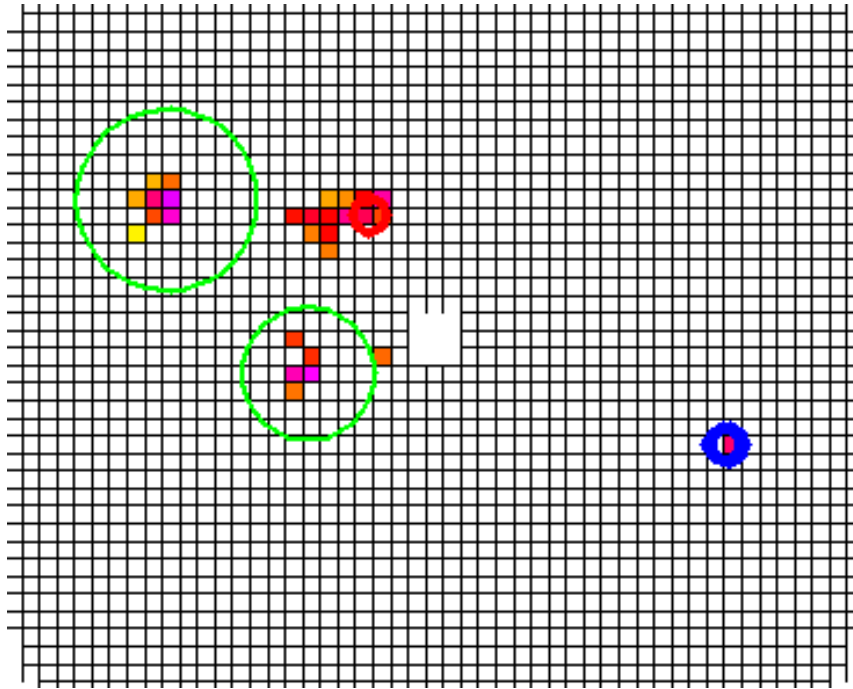


Shortest distance to nearest Type 1 Shower

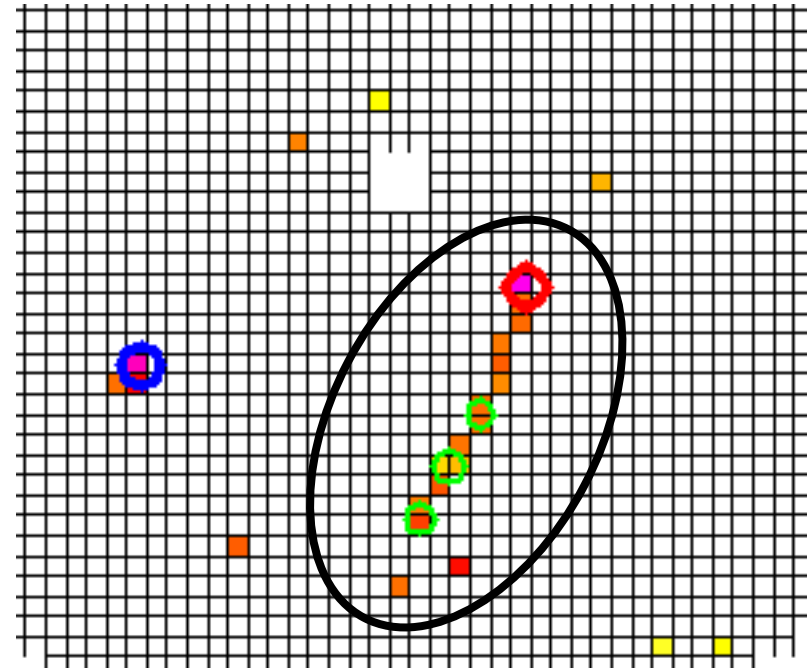


Number of Hits per shower

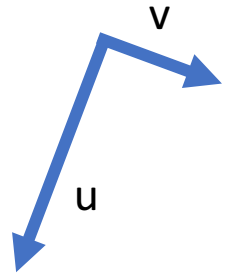
# Principal axes of a shower



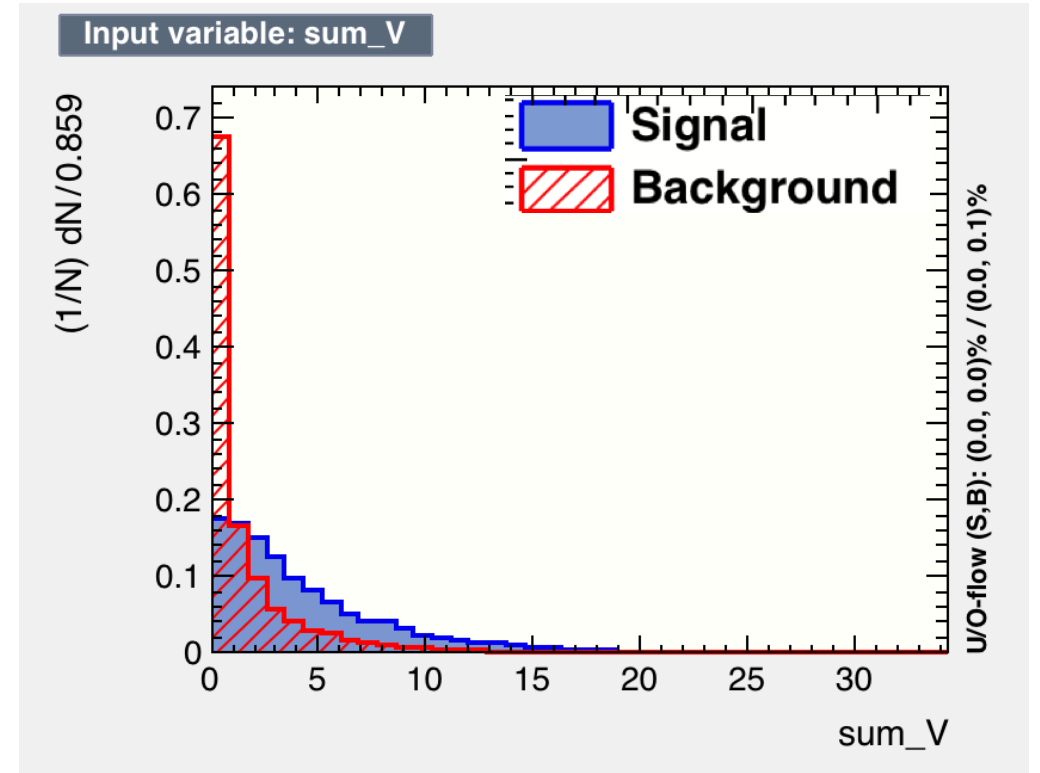
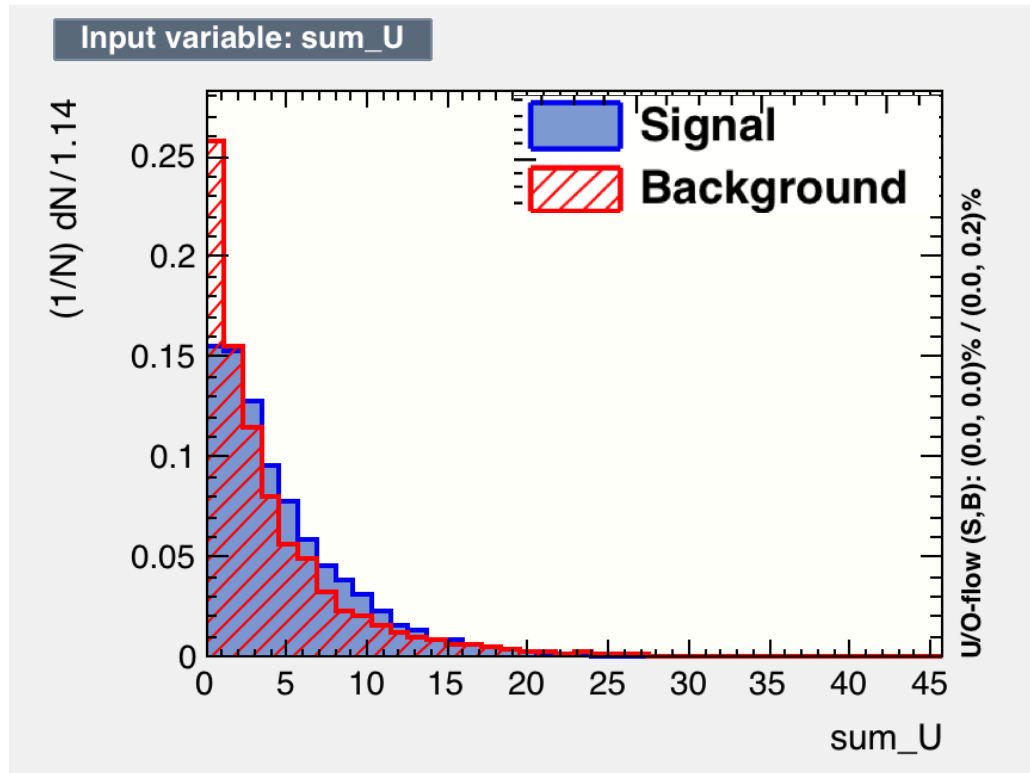
Type 0 shower



Type 2 shower



# Input variables



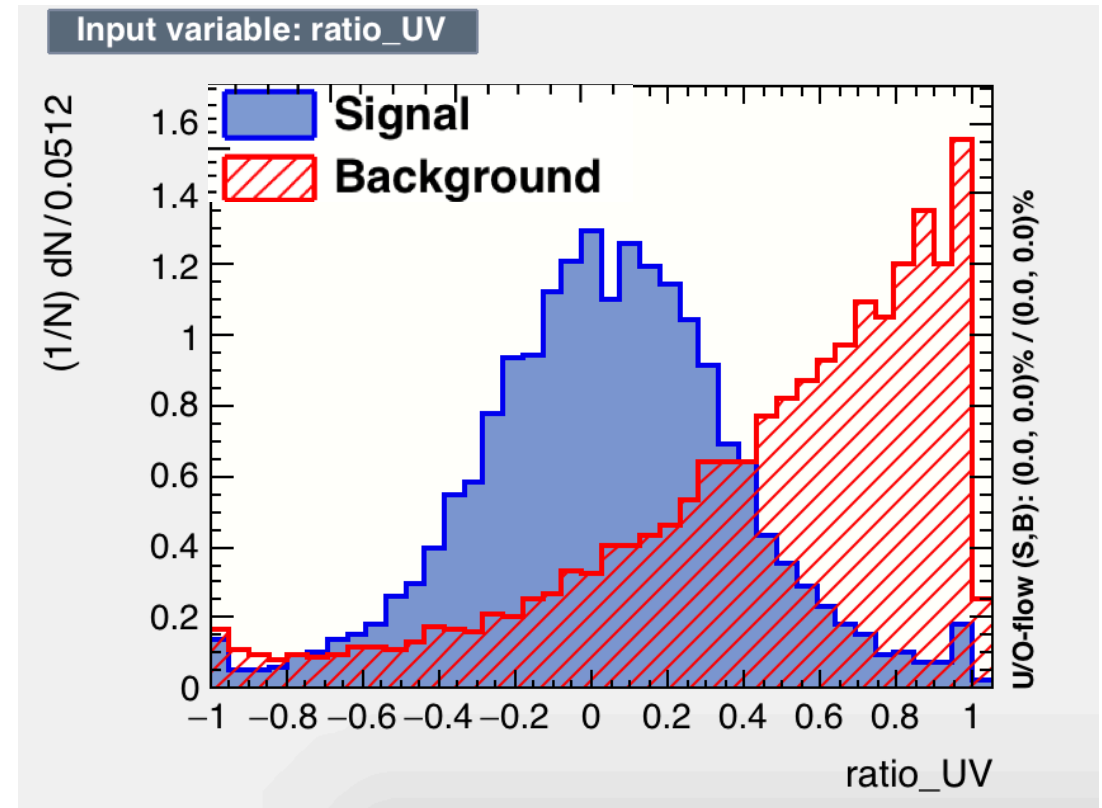
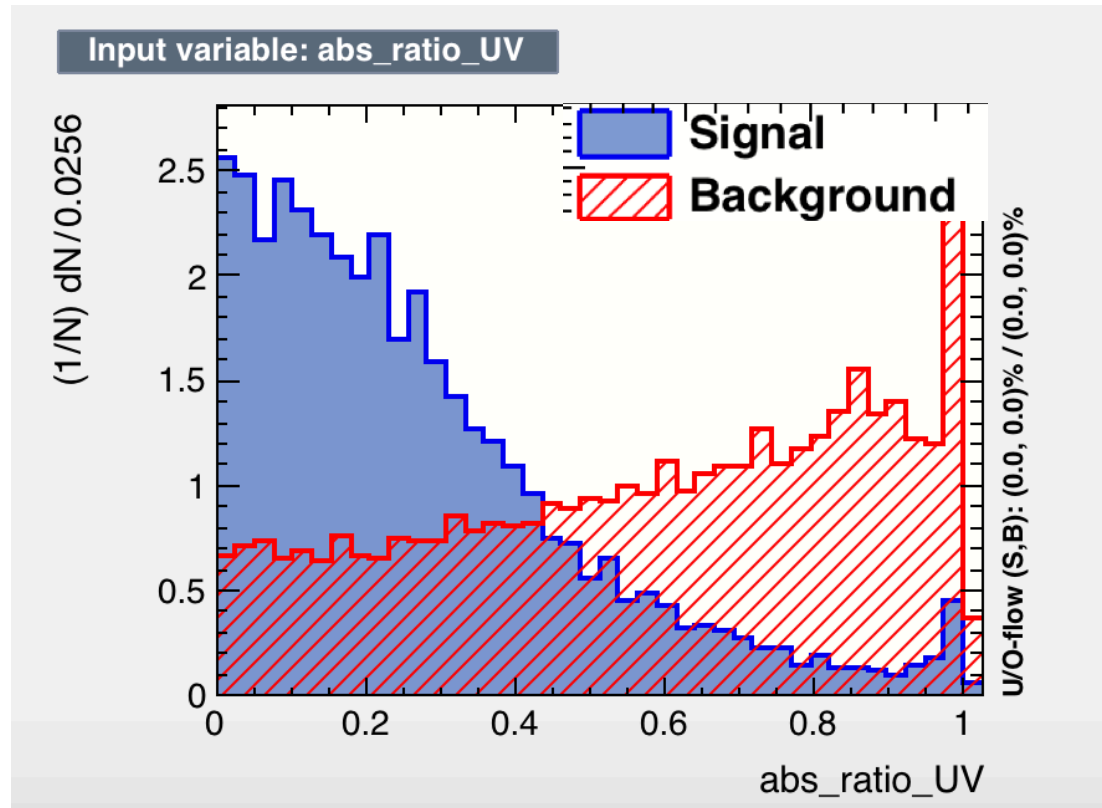
Energy weighted width of the shower along principal shower axes

$$\text{sum\_U} = \sum_i E_i (\mathbf{h}_i \cdot \mathbf{u})^2 / \sum_i E_i$$

$\mathbf{h}$  is the vector from the shower center to the hit block in the shower

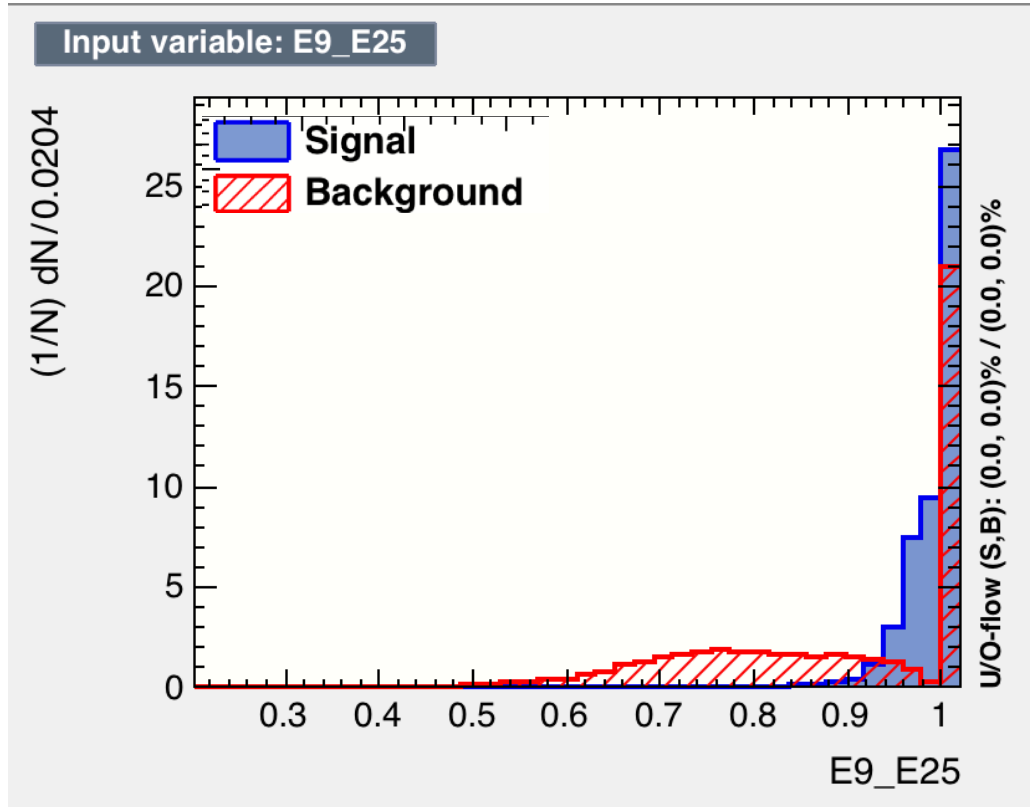
$\mathbf{u}$  is the principal axis

# Input variables

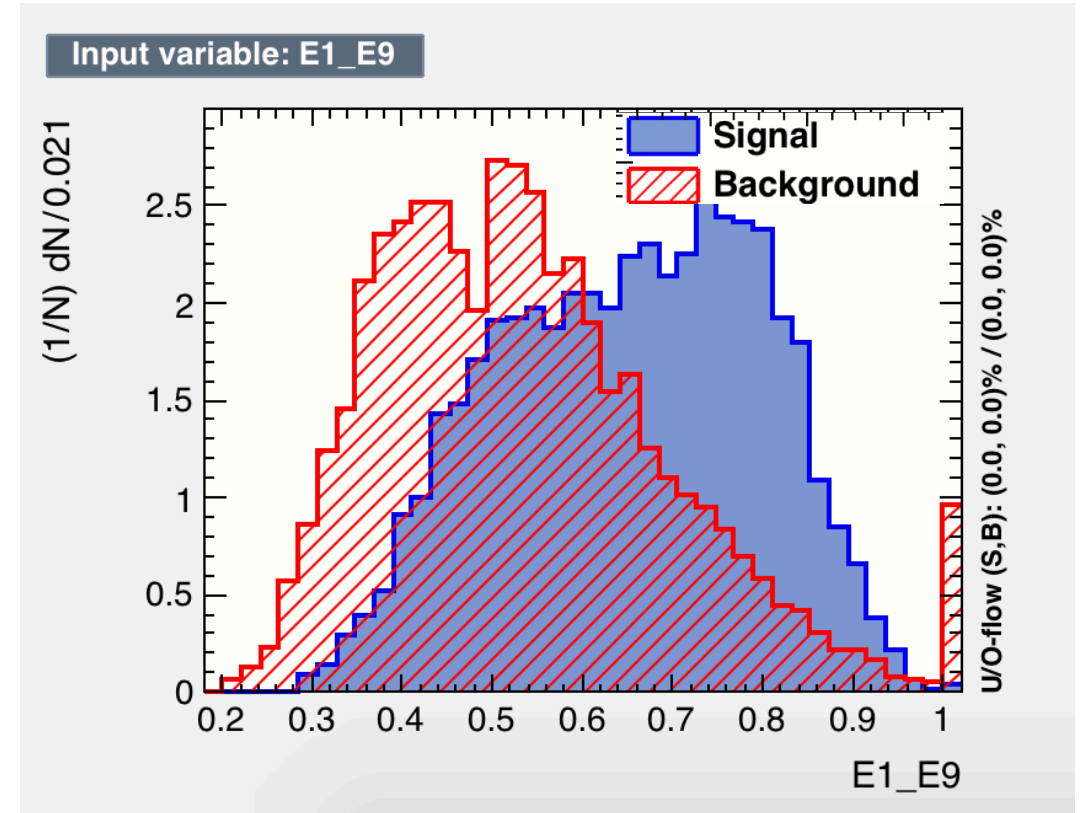


Ratio of energy weighted shower widths along principal shower axes

# Input variables



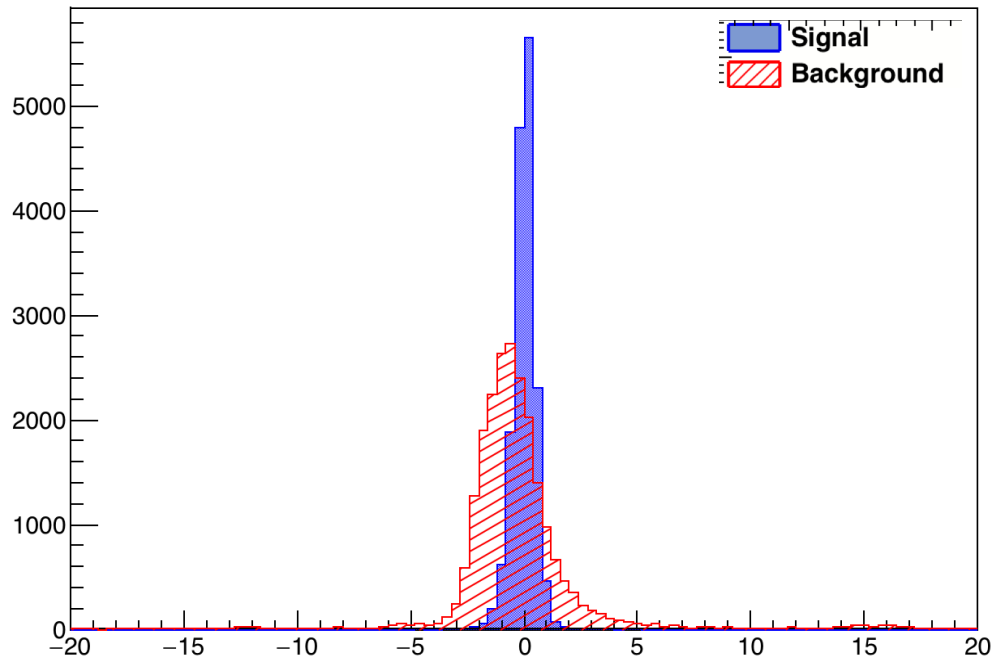
Energy in the nearest 9 blocks  
over the energy in nearest 25



Energy in central block over  
energy in the nearest 9 blocks

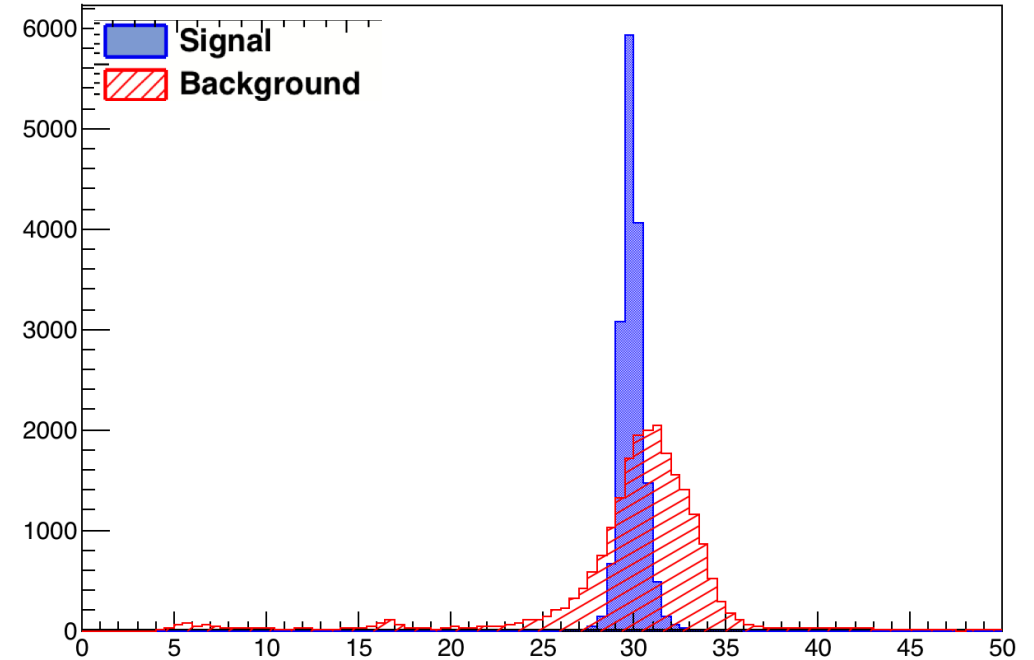
# Input variables

Input variable: tShowTrDiff



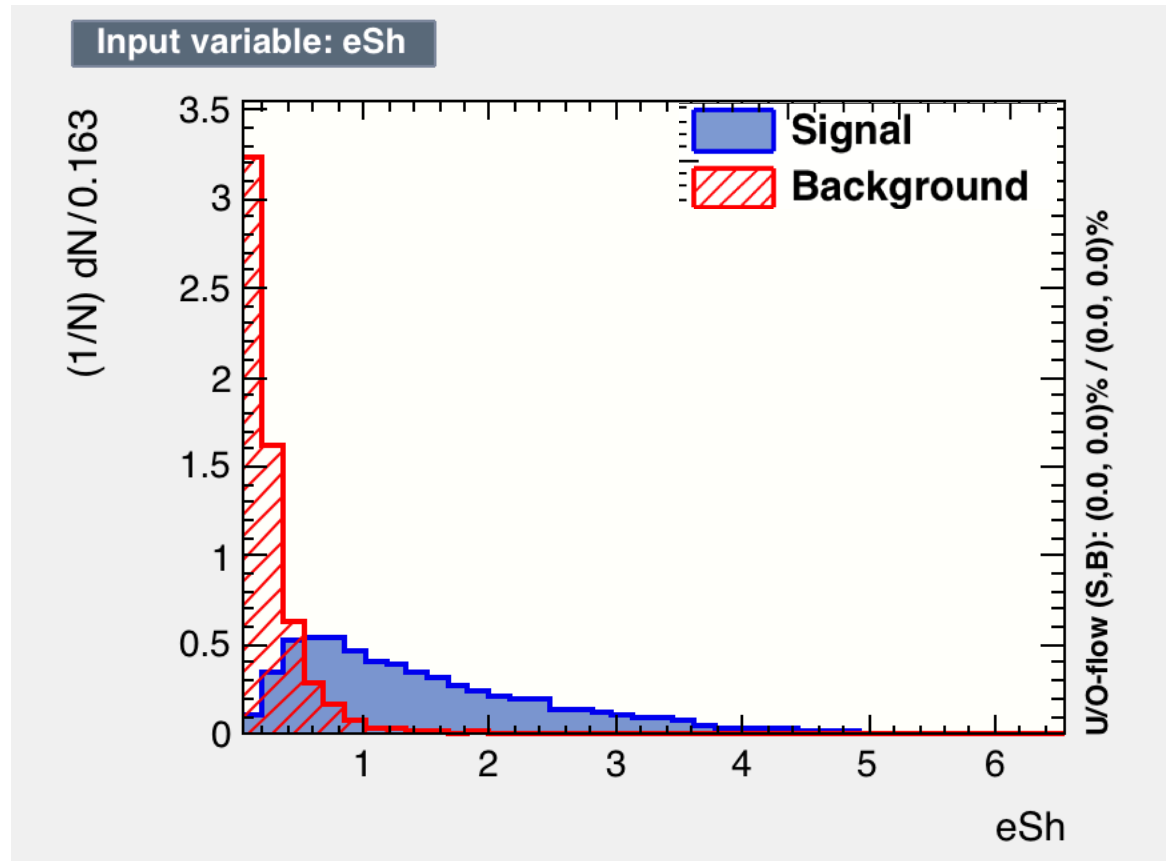
Difference between Time of Shower and Time of Track

Input variable: dist\_Time



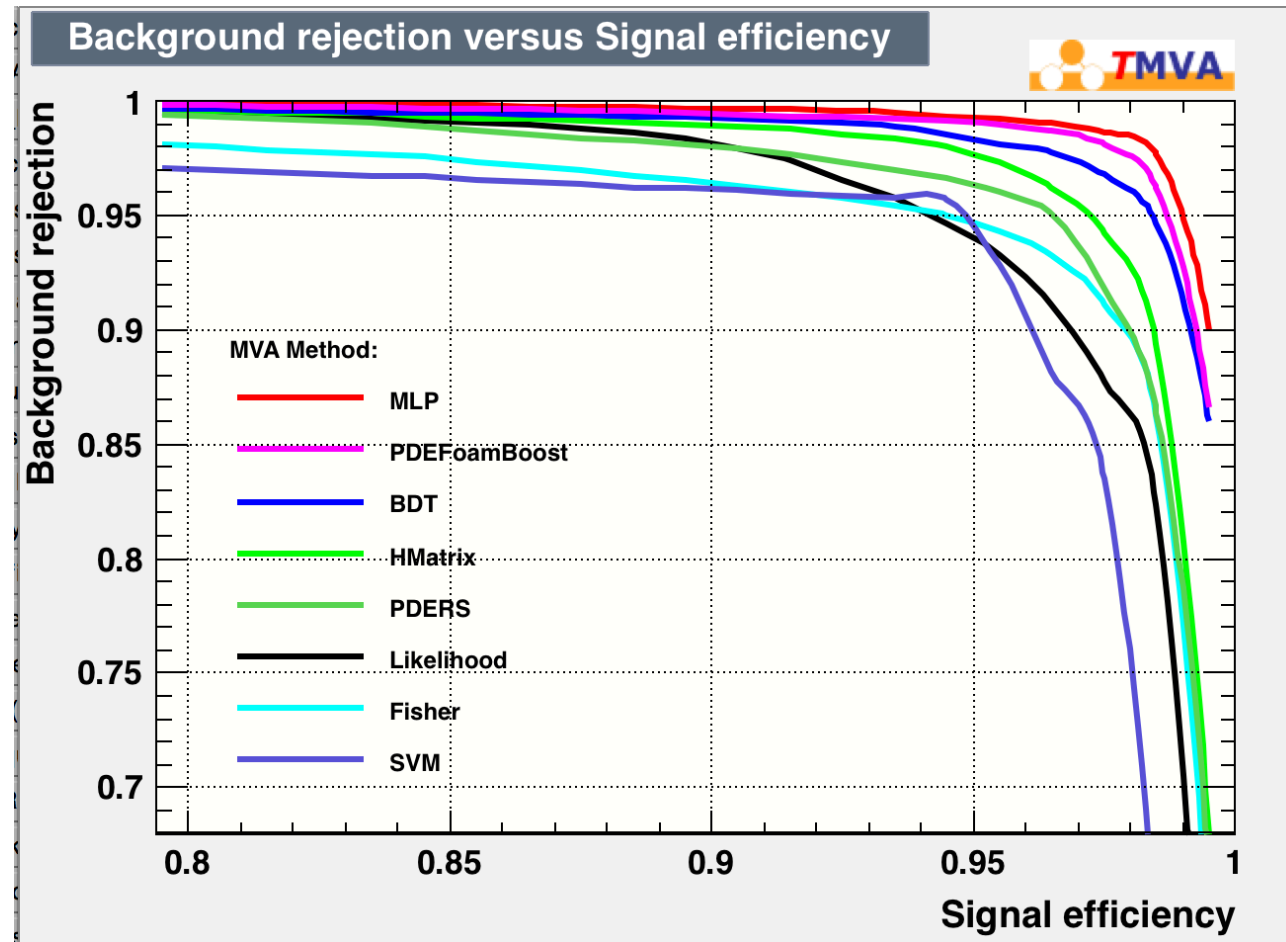
Effective velocity

# Input variables



Energy of the shower

# Algorithms: first glance

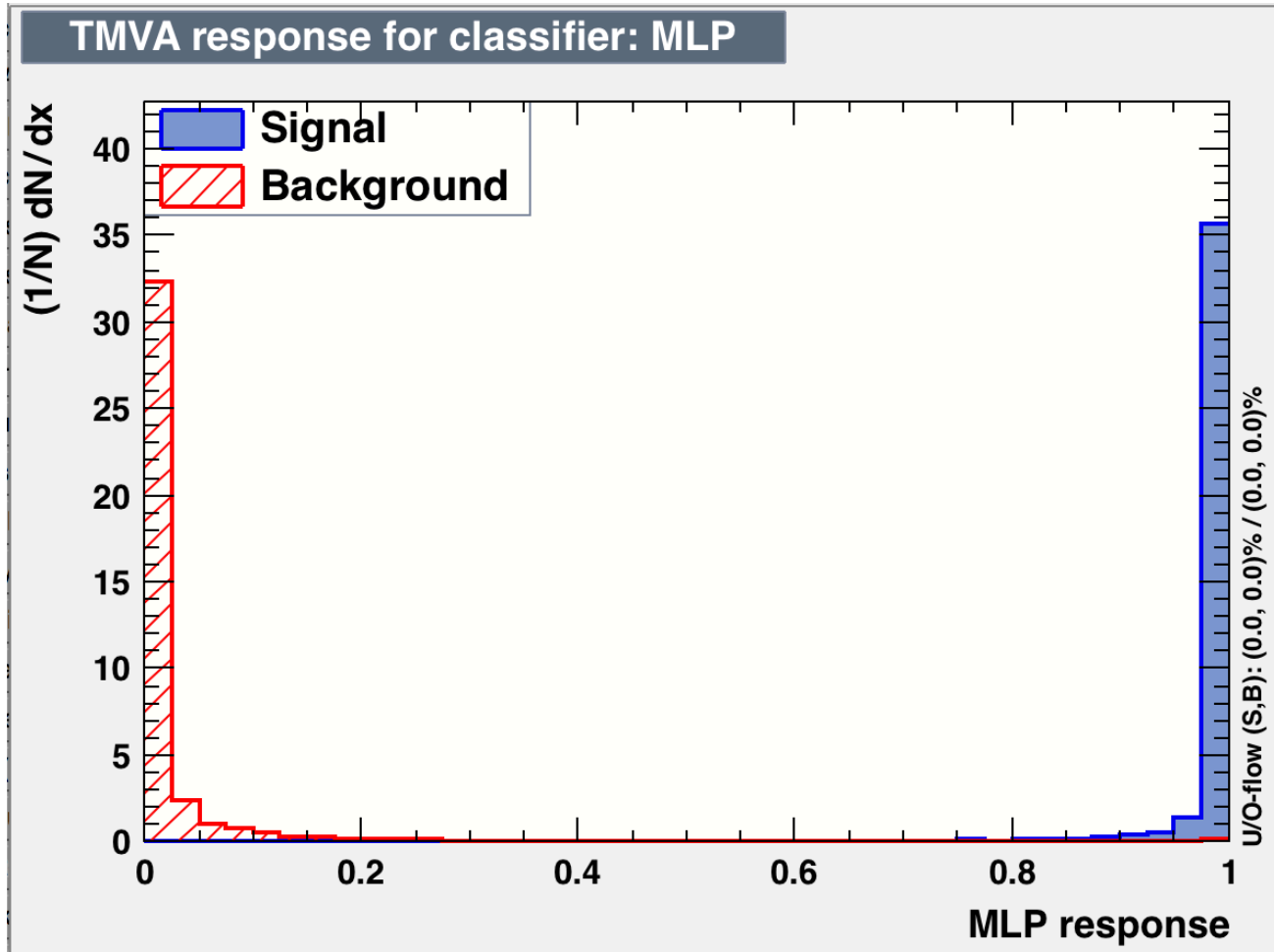




# Algorithms

- **Performance:** Multi-layer perceptron (MLP)
- **Implementation Difficulty:** MLP
- **Runtime:** PDE Foam Boost
  - .5 ms per event as compared to 1 ms per event

# Classifier Output Distribution



# Summary

- Most important thing is good signal efficiency to avoid discarding good photons
- The Multi-Layer Perceptron algorithm offers a promising option with good signal efficiency and background rejection
- Lot of useful input variables and parameters that can also be tweaked to serve individual purposes
- Currently working to implement the MLP algorithm separate from TMVA