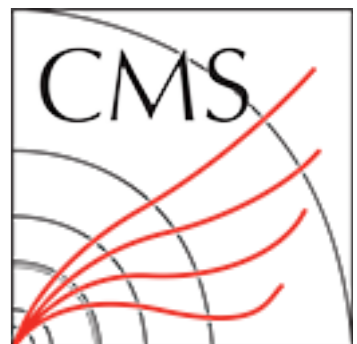


Exploring End-to-End Deep Learning for Event & Object Classification

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S2I2 HEP/CS Workshop, 2017-MAY-02



Motivation

The Current State of Affairs

ADC, $t = -125$ ns

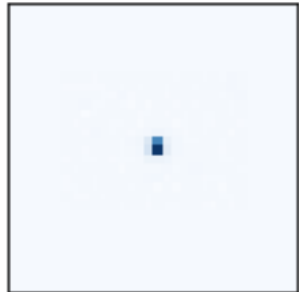


...

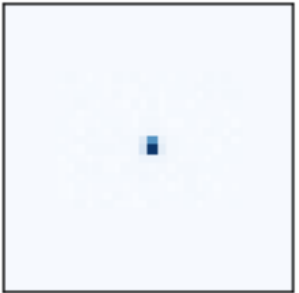
ADC, $t = -25$ ns



ADC, $t = 0$ ns

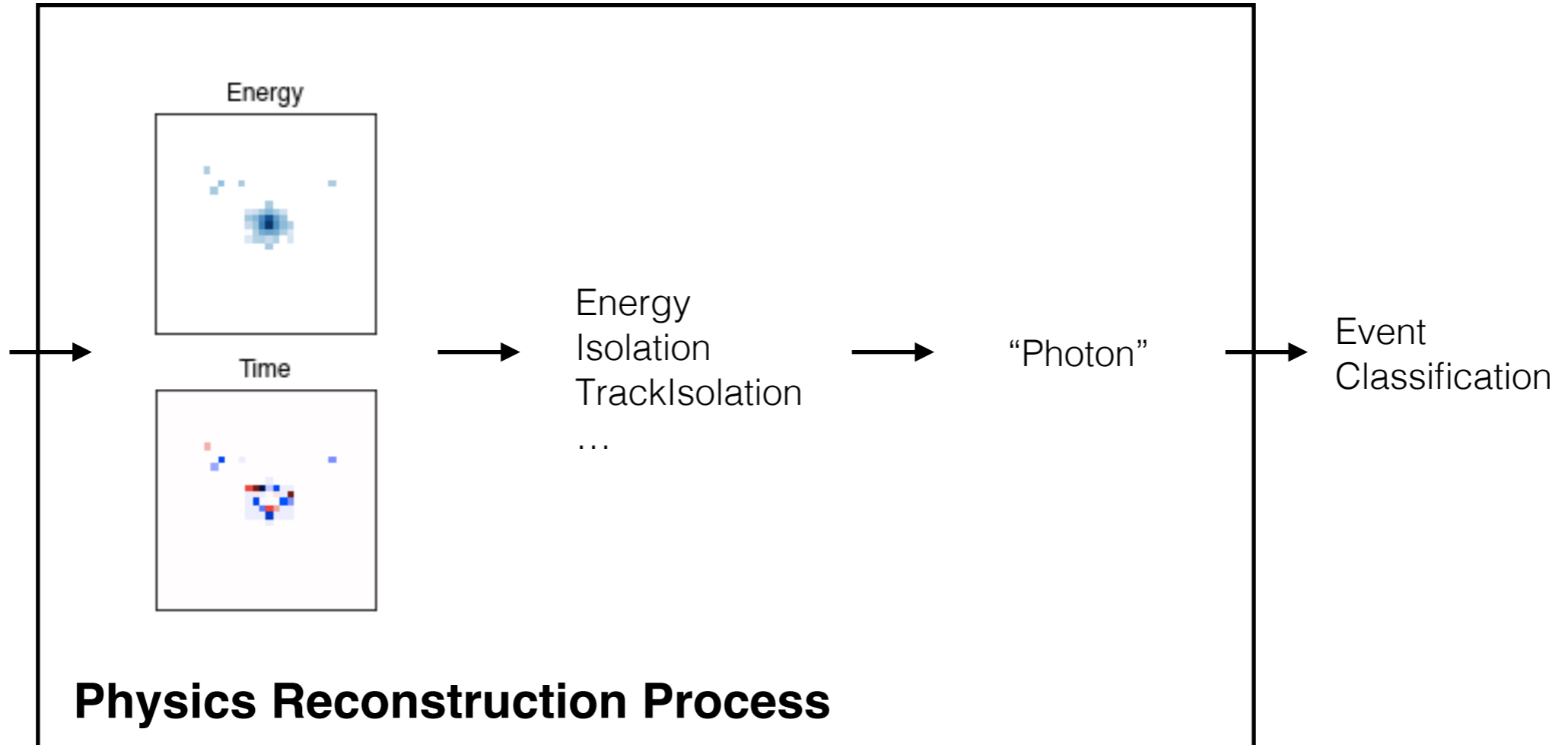
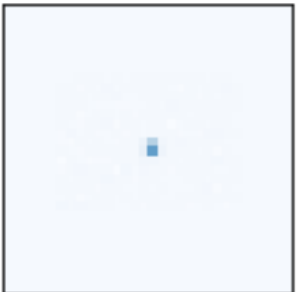


ADC, $t = +25$ ns



...

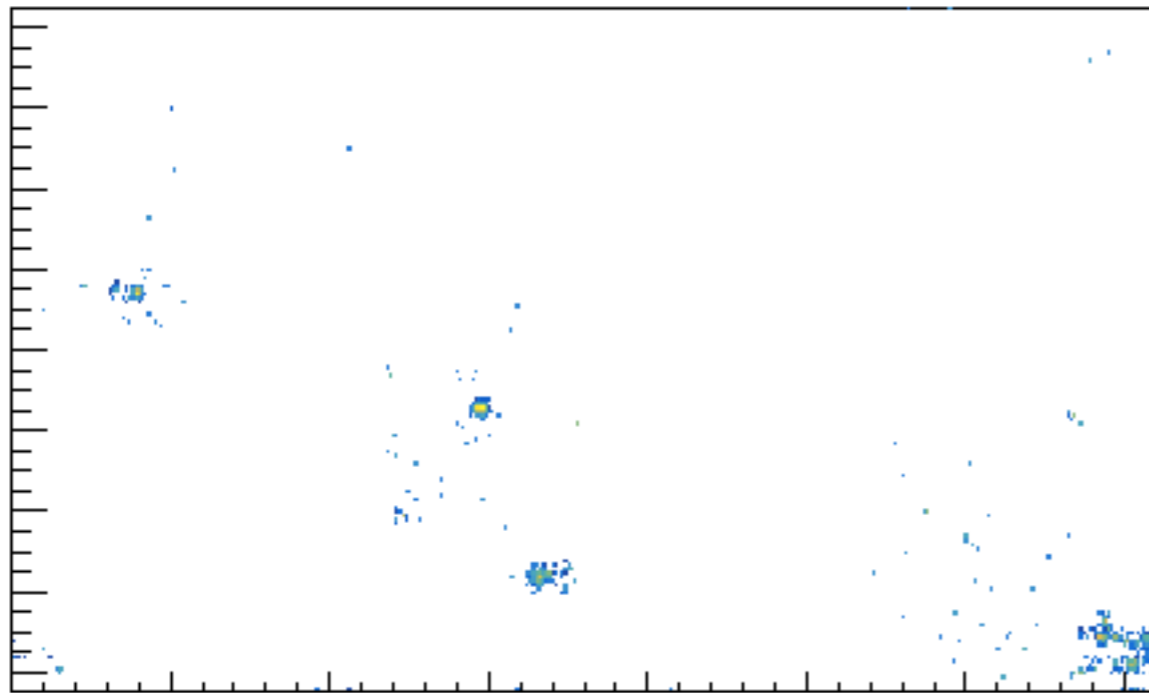
ADC, $t = +100$ ns



Q: *Are we throwing away information?*

Should this be done before or after event classification?

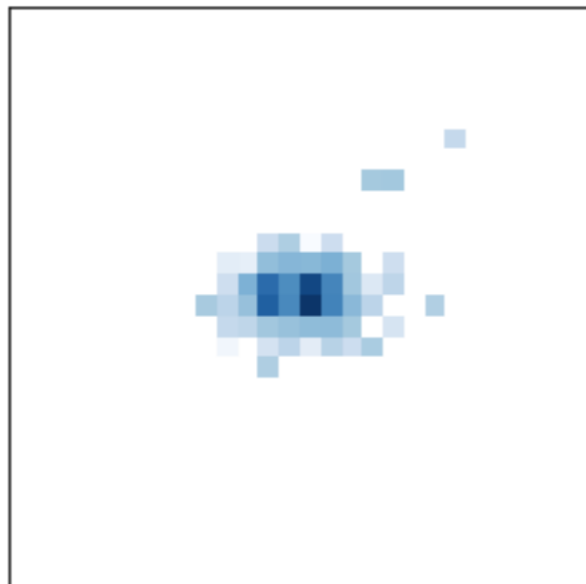
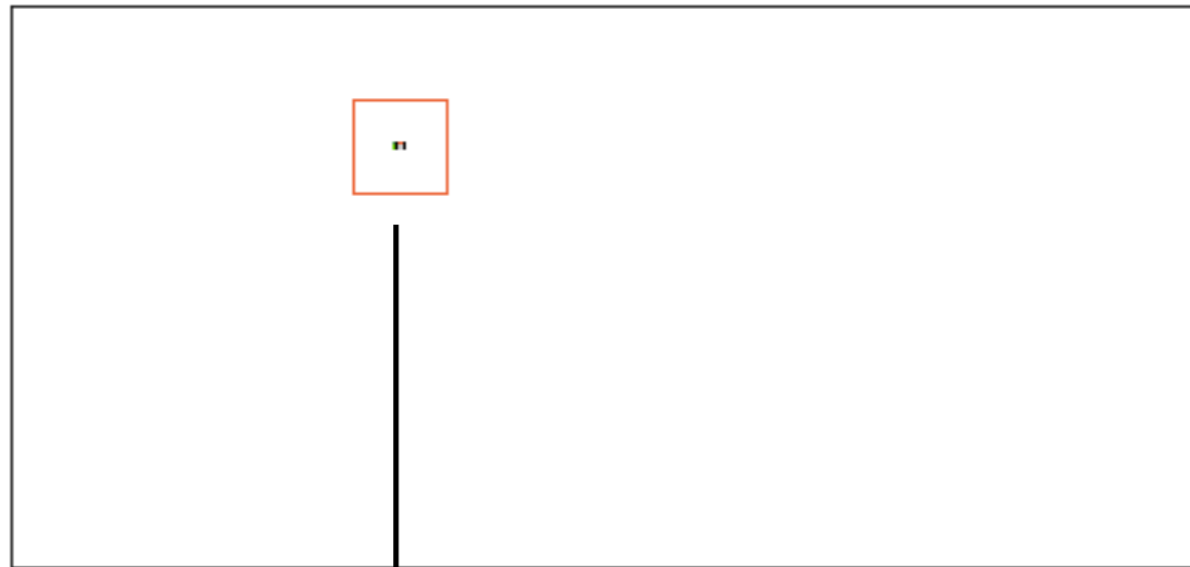
End-to-End Approach



Event Class
Prediction

Take detector-level inputs and directly output event class without bothering with reconstruction

Proof Of Concept



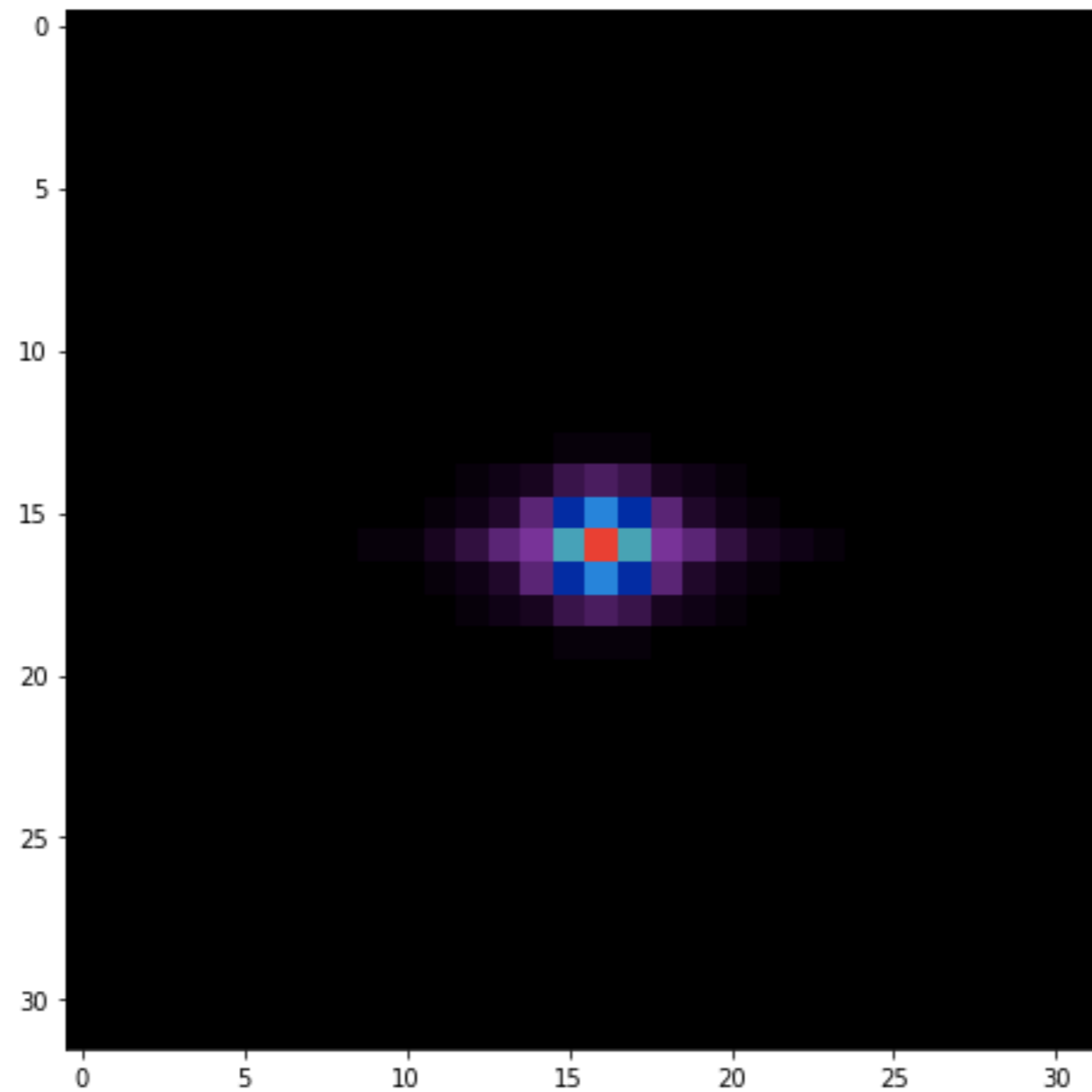
Particle ID
Prediction

As a proof-of-concept,
classify particle showers in ECAL

Shower Visualization

Photon-Induced EM Shower

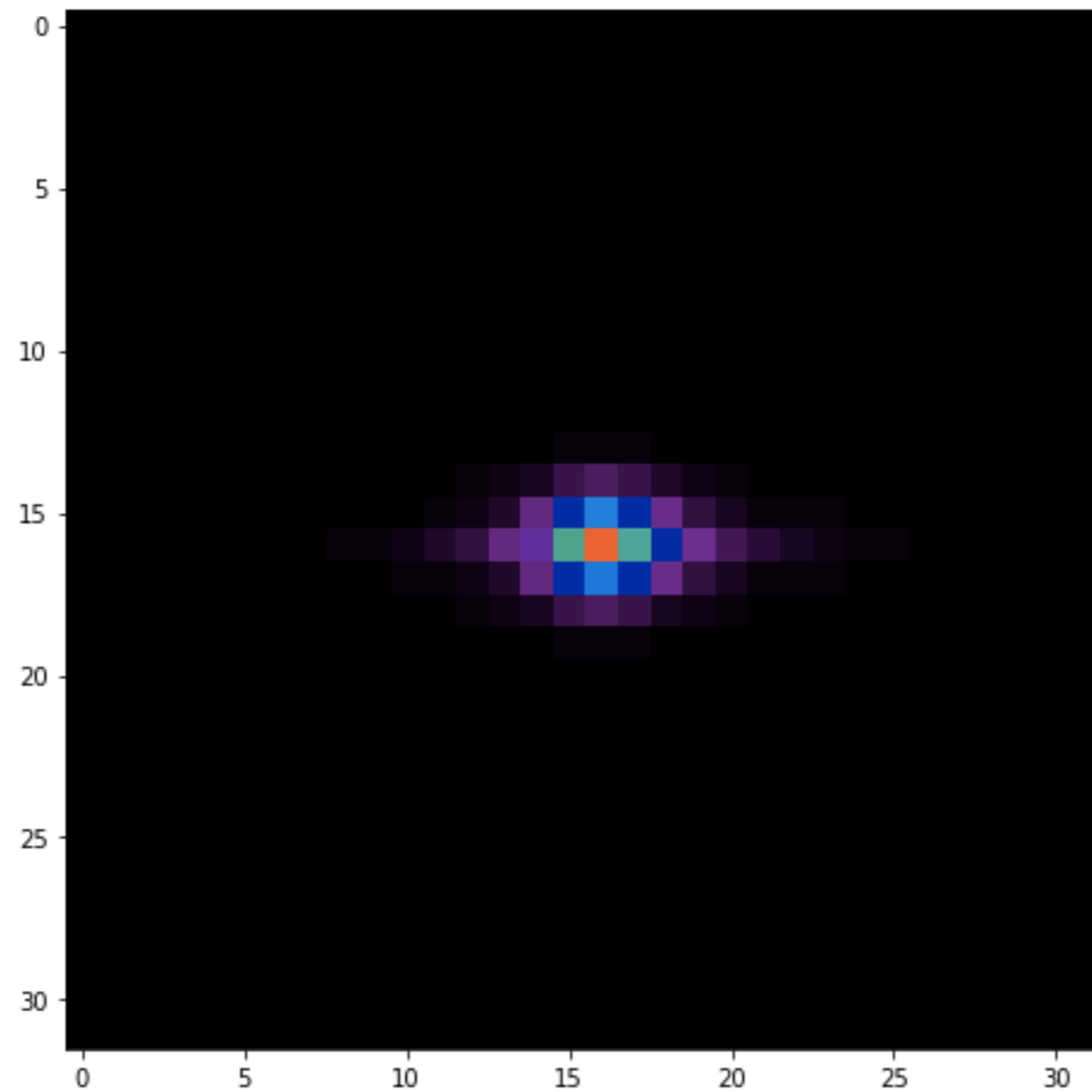
mean energy distribution over 10k events



Shower Visualization

Electron-Induced EM Shower

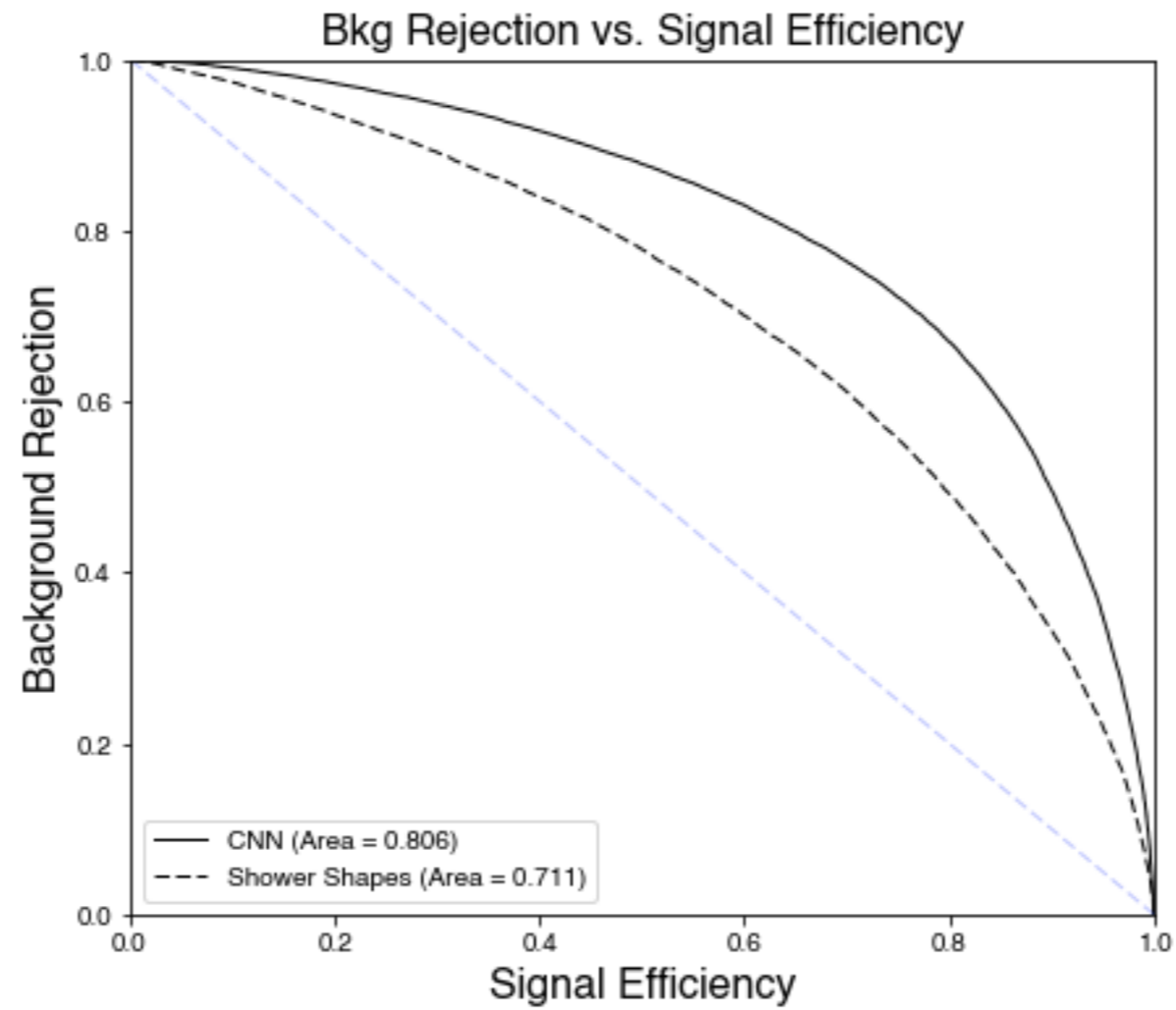
mean energy distribution over 10k events



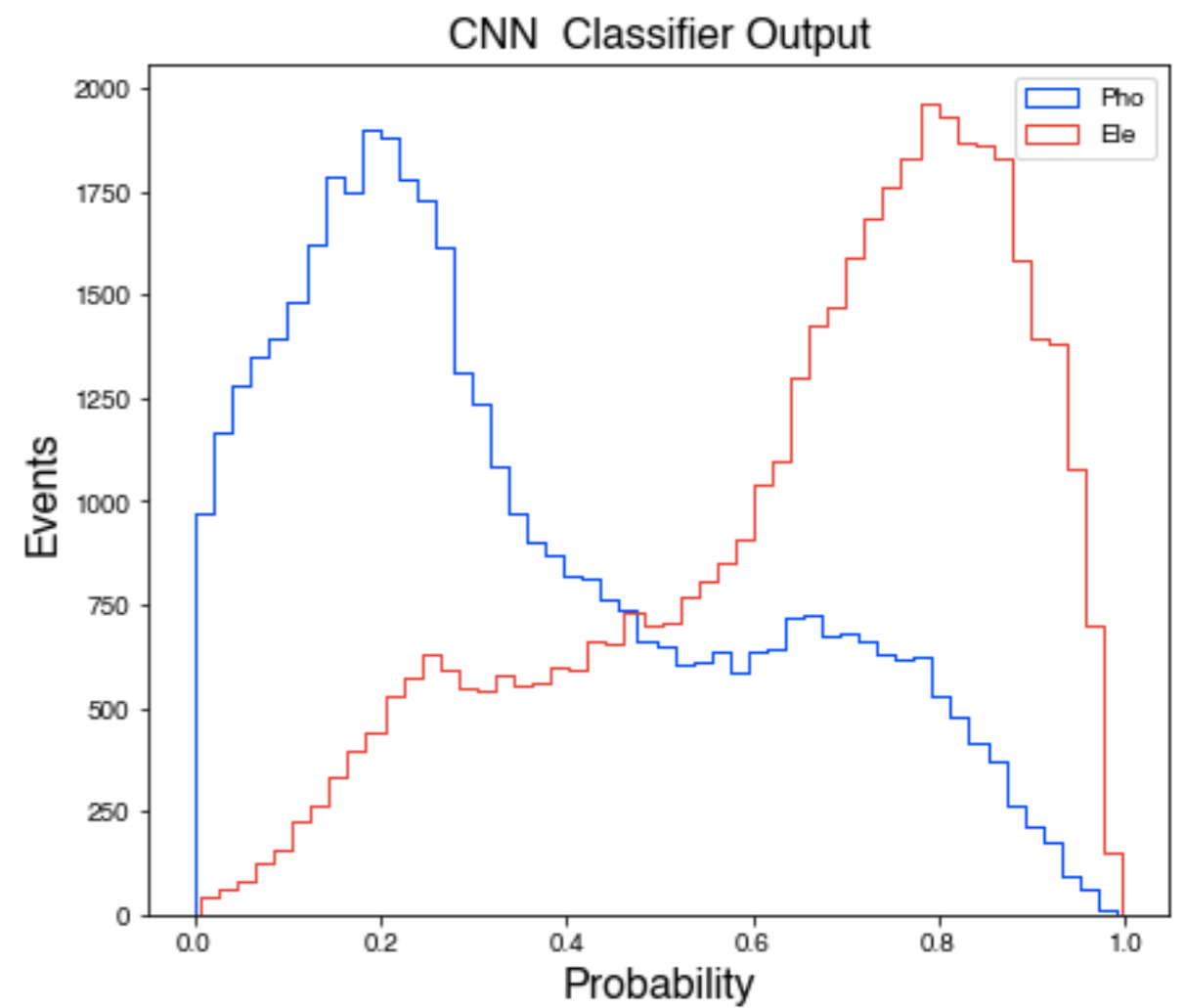
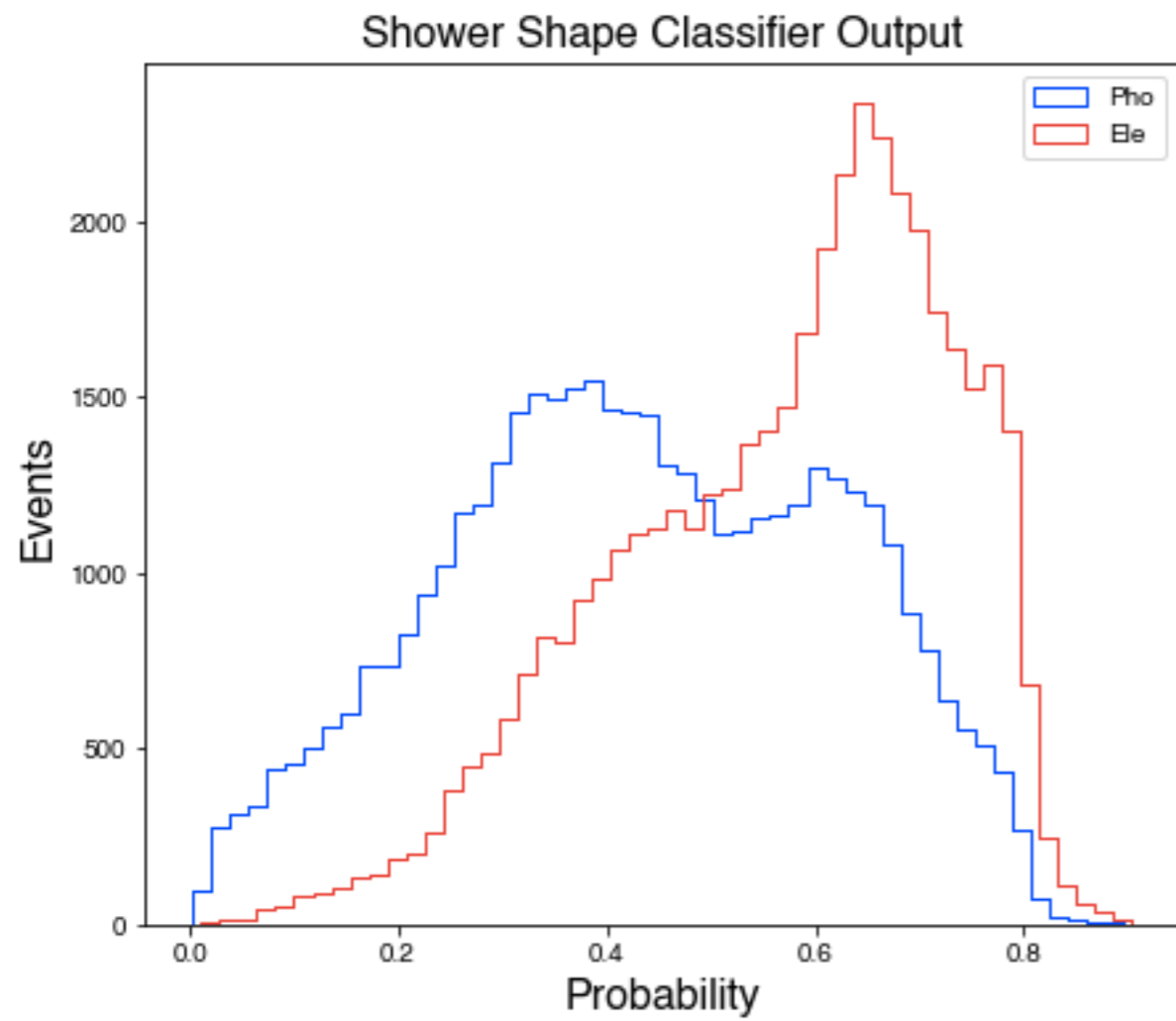
Preliminary Results

Network Type Inputs	Fully-Connected Shower Shape Variables	Fully-Connected Flattened Image	CNN Stacked Images	CNN+LSTM Image Sequence
ROC AUC	0.708	0.770	0.806	0.799

Preliminary Results

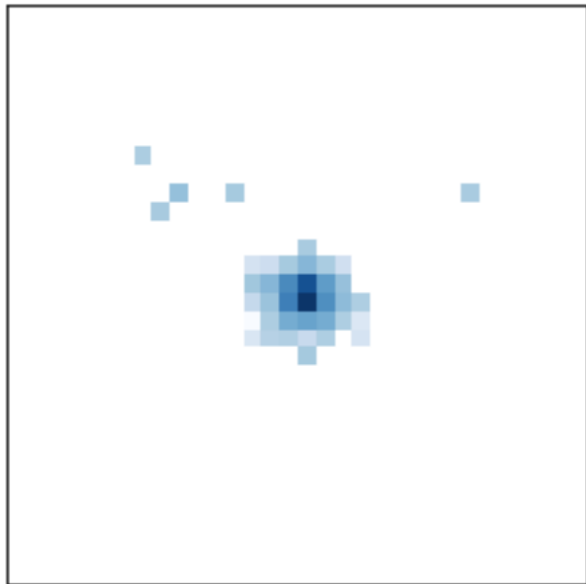


Preliminary Results



Visualizing CNN Layers

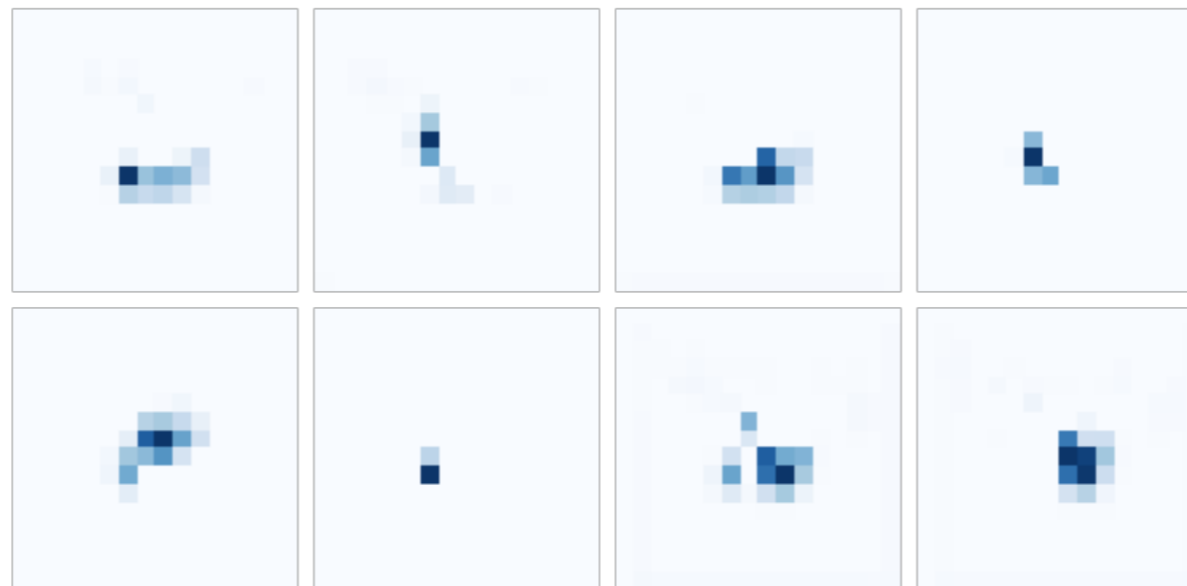
Input



Convolutional Layer 1



Convolutional Layer 3



Output

88%

Photon

(correct!)