

Machine Learning BSM Searches at the Fermilab SBN Program

Jamie Dyer
Colorado State University

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Authors:
B. Batell, J. Berger, J. Dyer, A. Ismail



The Higgs Portal – *The Model*



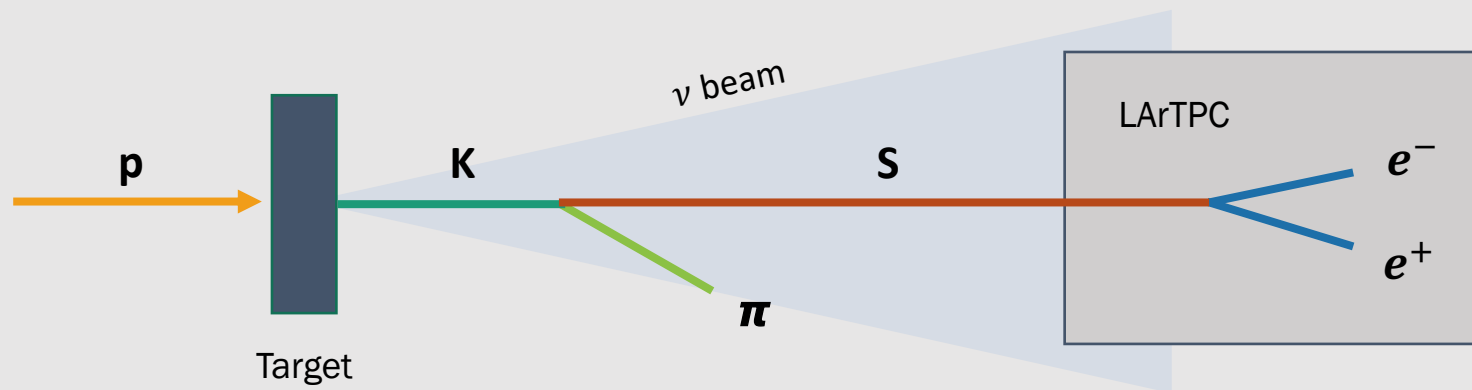
- S mixes with the Higgs boson, and provides an avenue by which Dark Matter can couple to the SM
- Two parameters: m_S, θ_S
- Production: $K \rightarrow \pi S$
- Decay: $S \rightarrow e^-e^+, \mu^-\mu^+, \pi^-\pi^+$

The Higgs Portal – *Sensitivity at Fermilab SBN Program*

Conveniently...

- Neutrino beams produce many kaons, &
- The excellent particle identification abilities of Liquid Argon Time Projection Chambers (LArTPC's) make distinguishing these rare signal events from background feasible

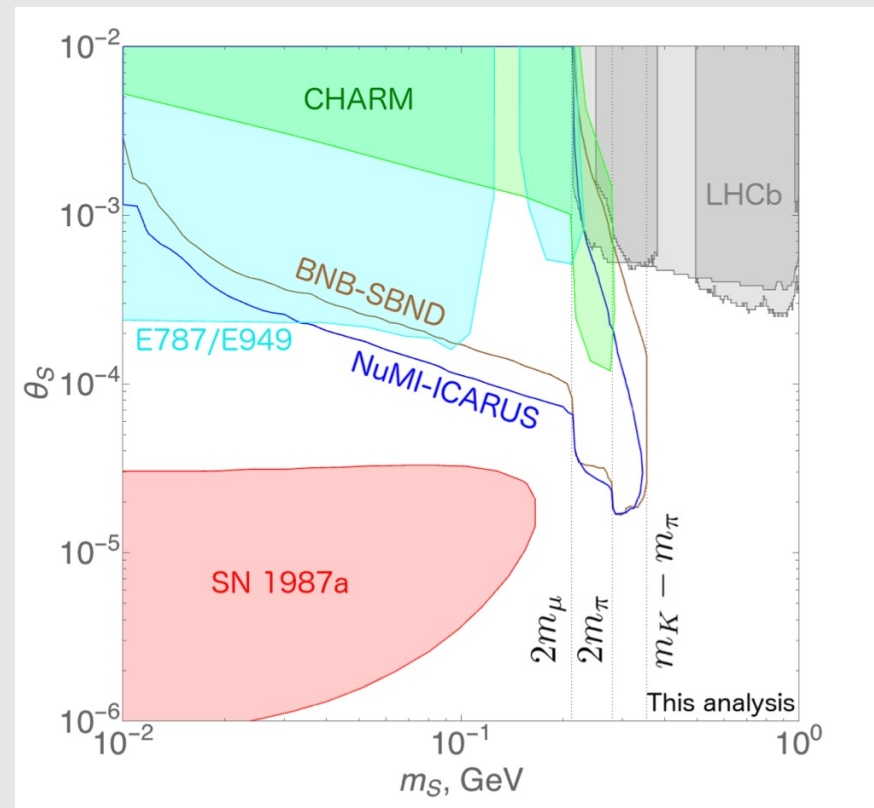
→ The Higgs Portal Model is testable at current and near-term neutrino experiments!



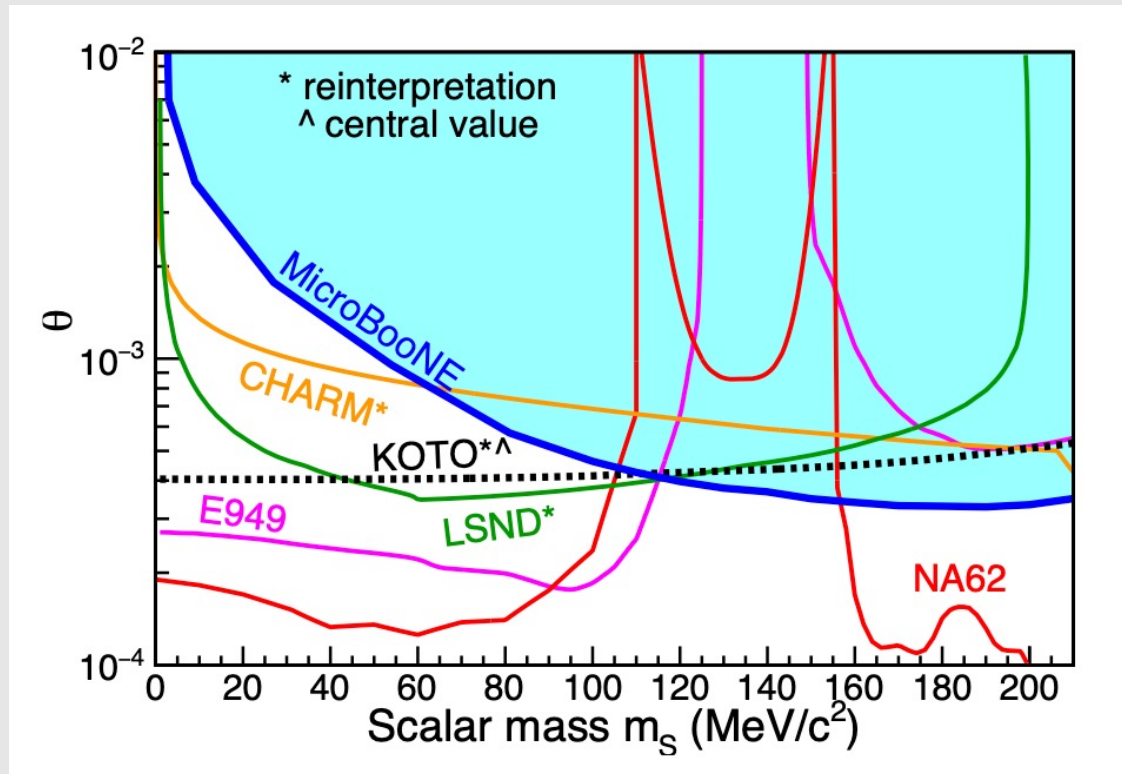
The Higgs Portal – *Sensitivity at Fermilab SBN Program*

Batell, Berger, Ismail (2019)
[arXiv:1909.11670](https://arxiv.org/abs/1909.11670)

- 2σ sensitivity to 5 signal events $m_S = \mathcal{O}(100 \text{ MeV})$ demonstrated for ICARUS detector with w/ NuMI beam

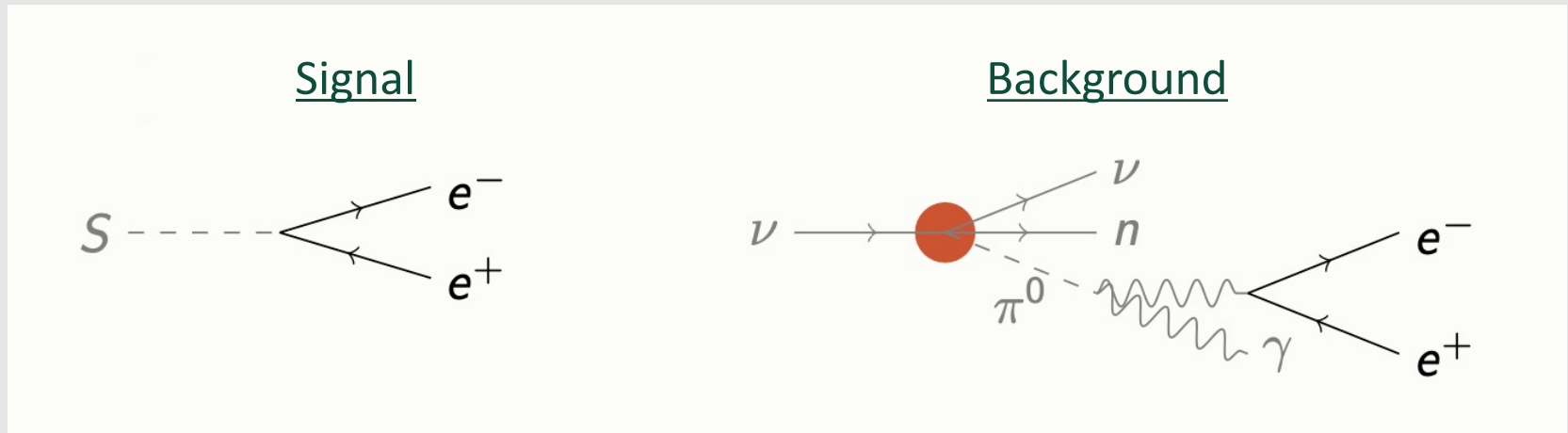


The Higgs Portal – *The Search is Happening Now!*



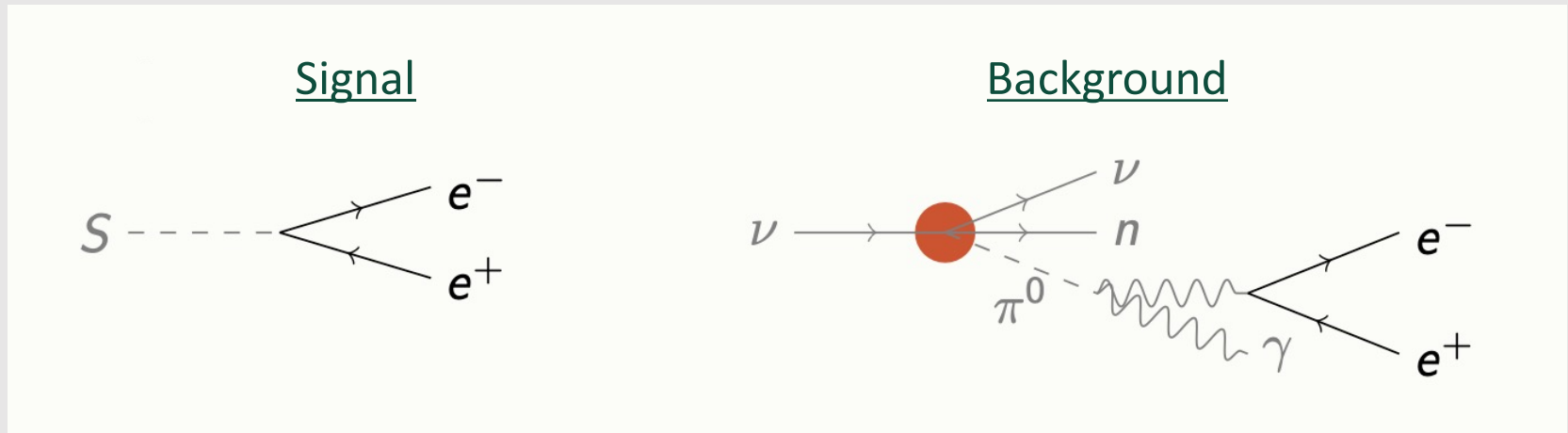
Microboone Collaboration (2021)
[arXiv:2106.00568](https://arxiv.org/abs/2106.00568)

The Higgs Portal – *Background*



- The lighter the scalar, the more columnated the shower pair and the more it resembles shower conversion.

The Higgs Portal – *Background*



- The lighter the scalar, the more collimated the shower pair and the more it resembles shower conversion.

Can machine learning be leveraged to discern signal from background?

This Work

Compare sensitivity to Higgs Portal Scalar using traditional cut-and-count analysis vs. convolutional neural network (CNN).

Simulation

1) Generate 4-momenta

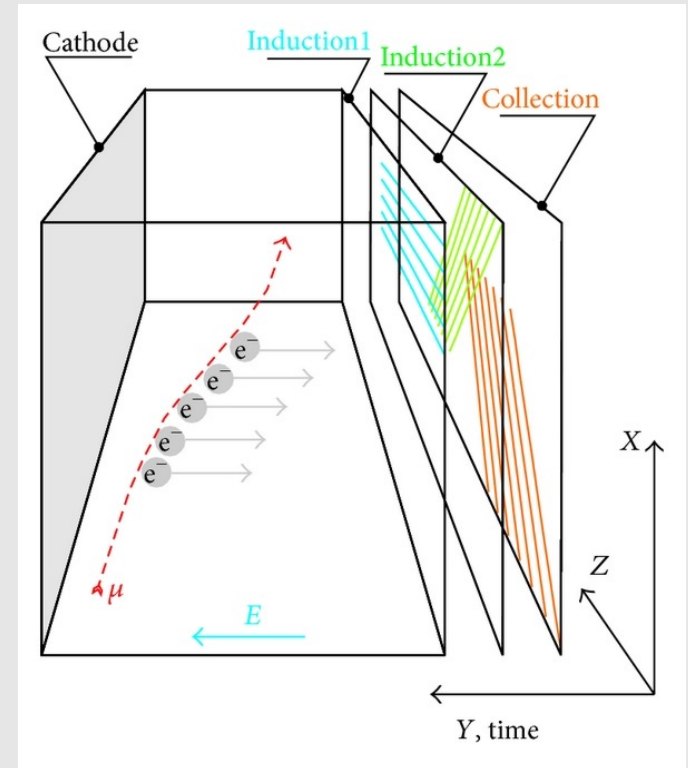
- Signal: find kaon parents from g4numi ν events, decay to S , propagate to detector, decay to e^+/e^- .
- Background: 3-5 GeV ν_e events w/ GENIE w/ similar topology to signal

2) Run kinematic variables through Geant4 configured for ^{40}Ar . Save non-zero energy deposits

3) Convert to charge & project onto LArTPC wire planes:

$$Q_{dep} \approx E_{dep}/W_{ion}$$

$$Q_{det} \approx Q_{dep} \frac{A}{1 + k(dE/dx)/|E_{drift}|}$$



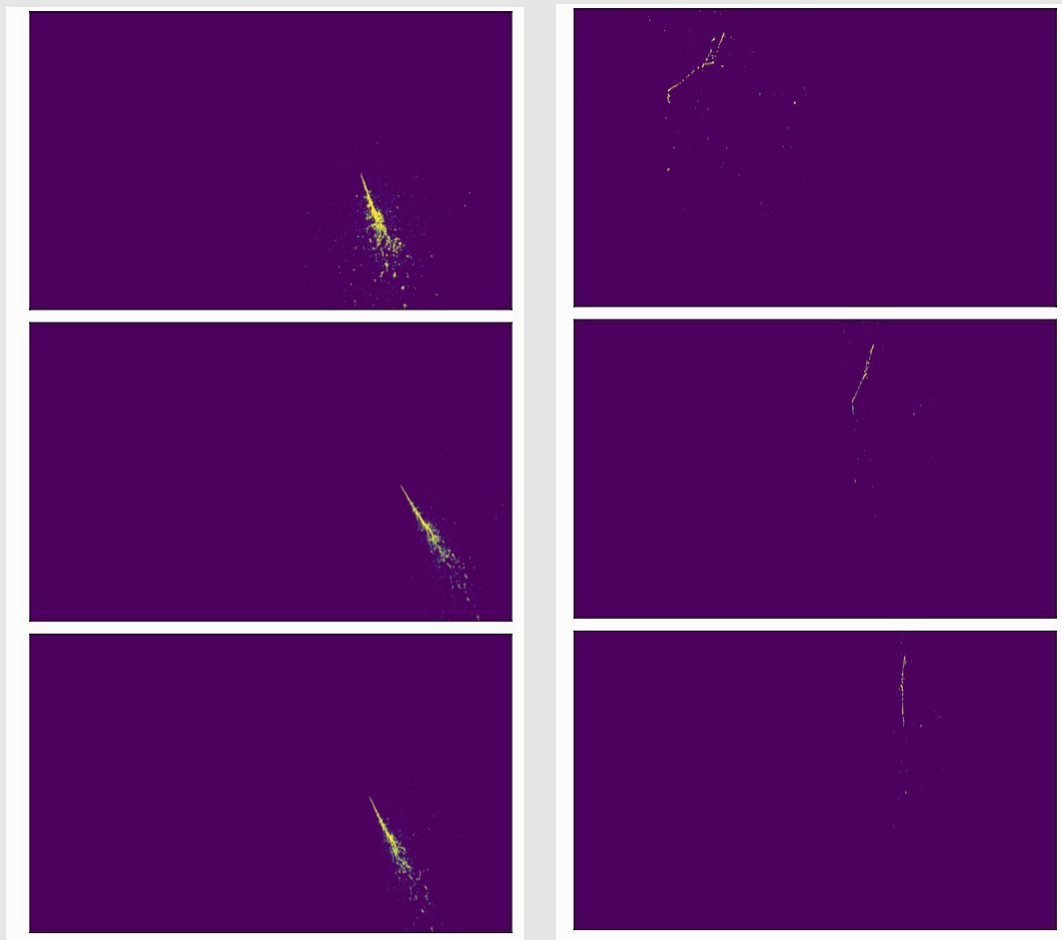
ICARUS LArTPC Schematic.
Antonello et. al. (2013)

<https://doi.org/10.1155/2013/260820>

Example Events

Signal

Background



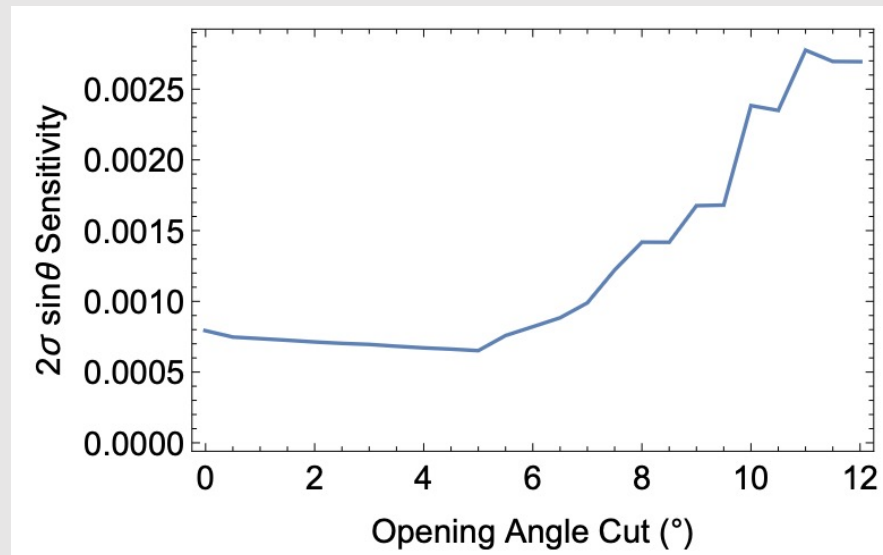
Advantages of this simulation pipeline:

- Much faster than a full detector simulation
- More detailed than GENIE or BSM generator four-vectors

Cut and Count Analysis

- Determine the efficiency and background rejection of opening angle cut as function of choice of angle
- Minimize $\sin\theta$ for 2σ using

$$Z \sim \frac{\text{signal}}{\sqrt{\text{background}}} \sim (\text{efficiency}) * (\sin^4 \theta) * \sqrt{\text{rejection}}$$

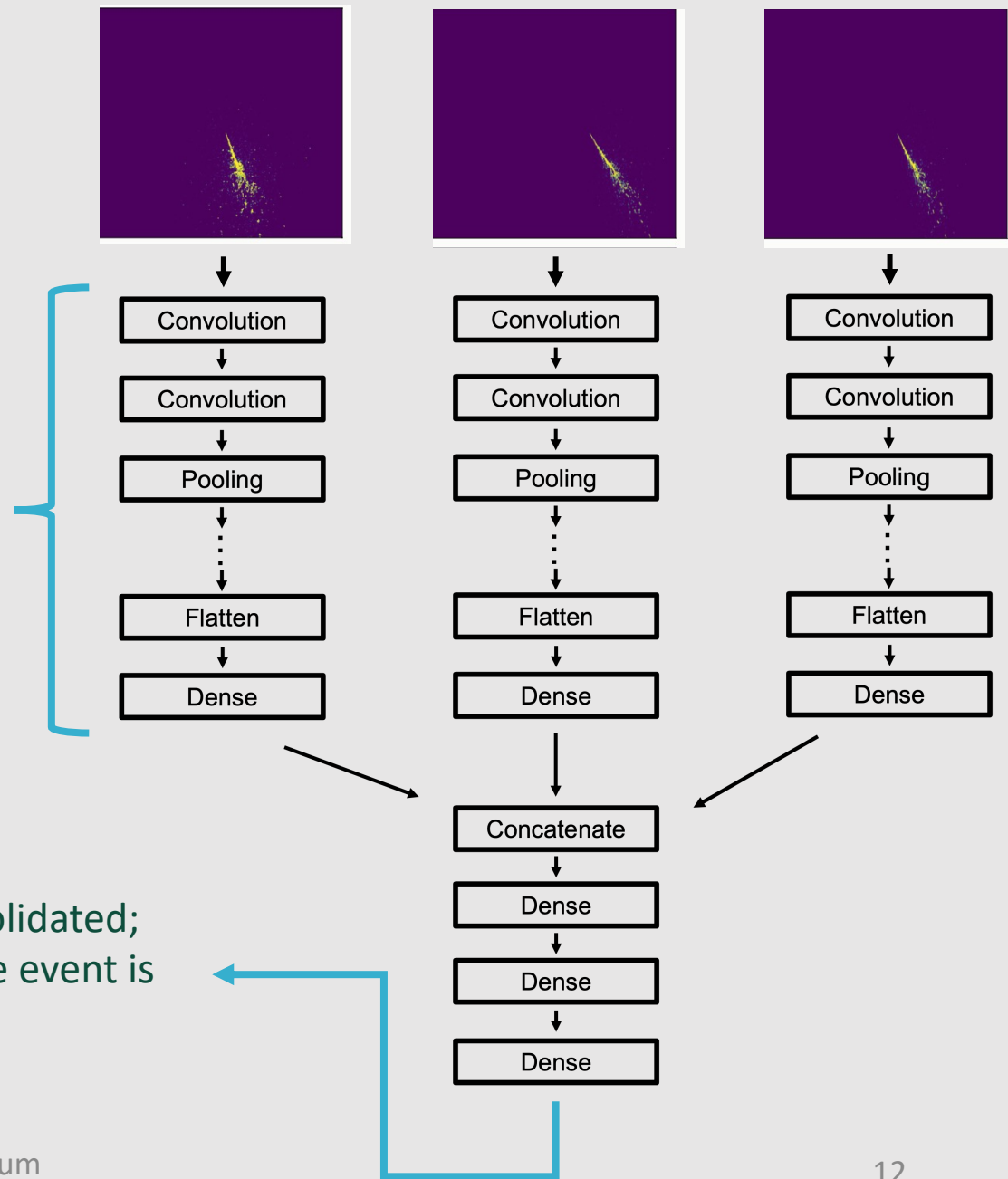


The CNN

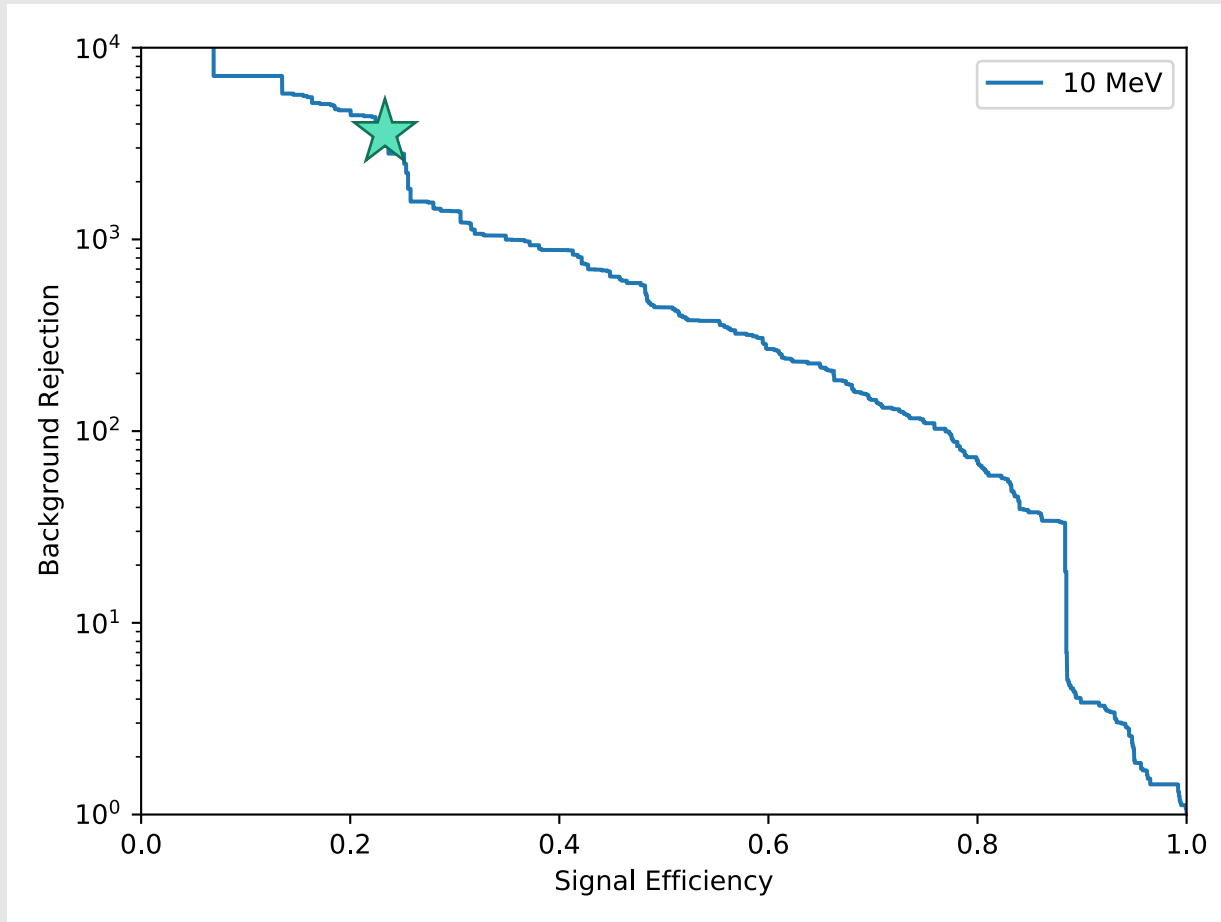
Wire plane images first evaluated with separate branches of CNN

Network finds correlations between spacially-distant pixels via convolution and pooling steps

Info from wire planes is consolidated; score for how “signal-like” the event is generated.



CNN Background Rejection vs. Efficiency



Results

Analysis	Limit on $\sin\theta$ for 10 MeV S
Cut-based	$8.1 * 10^{-4}$
CNN	$5.48 * 10^{-4}$

- Modest improvement for $\sin\theta$
- Event rate $\propto \sin^4\theta$
- Significant improvement in the rate

Takeaways & Outlook

- Argon-configured Geant4 simulation provides efficient and detailed simulation pipeline accessible to theorists.
- Described simulation + machine learning techniques can be used to improve sensitivity to BSM models within LArTPCs at current and near-term experiments.