

Graph Neural Networks for Angle Reconstruction in the ZDC

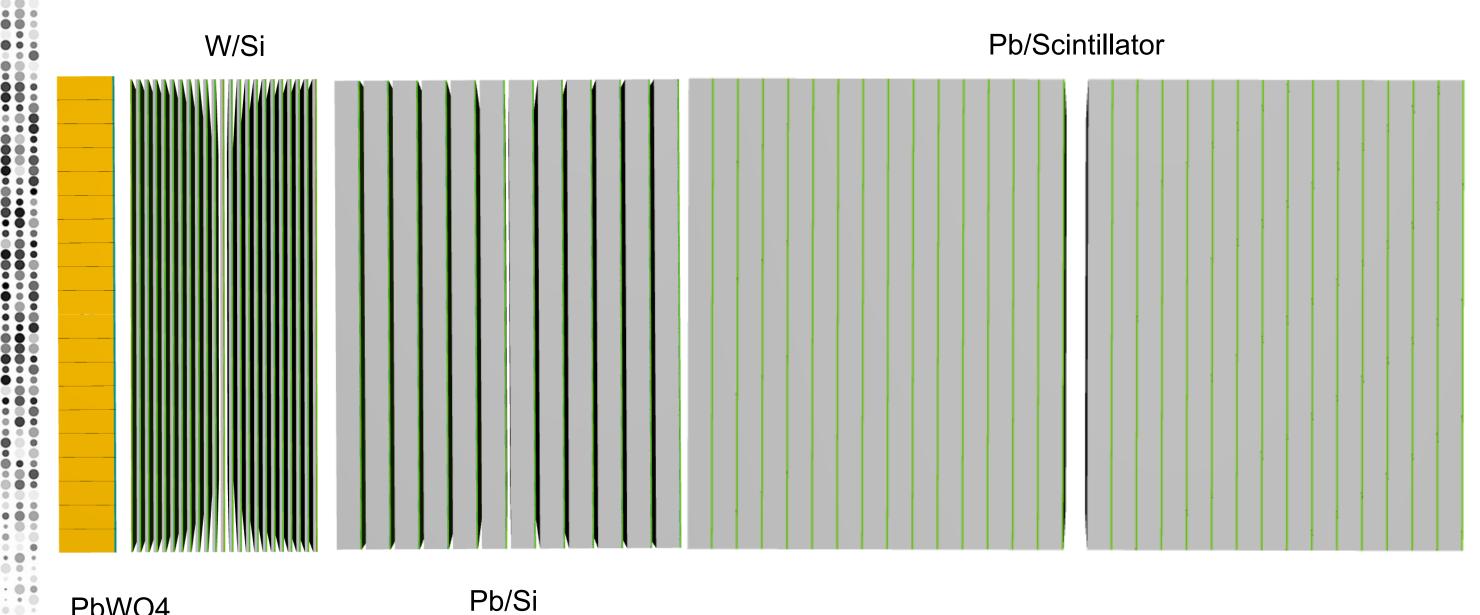
April 27, 2023





PbWO4

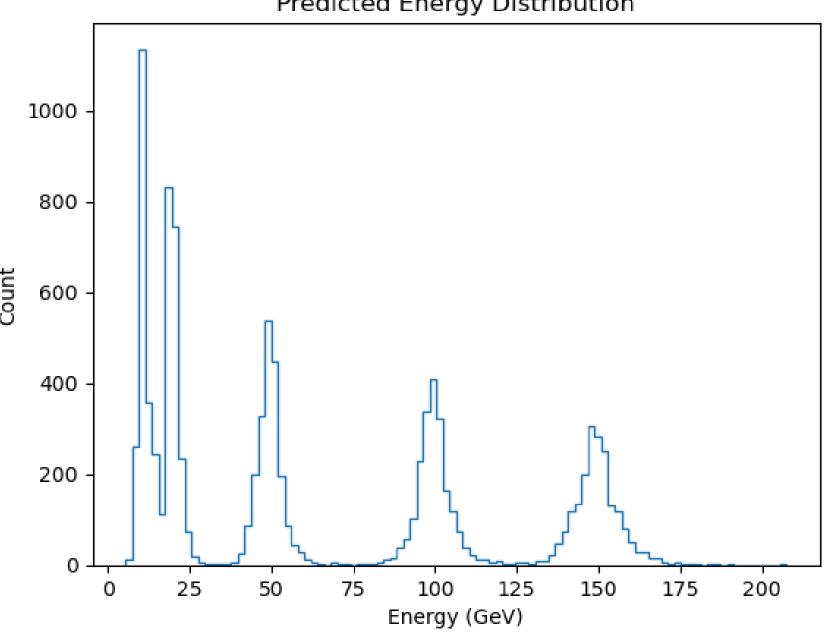
Current ZDC Design





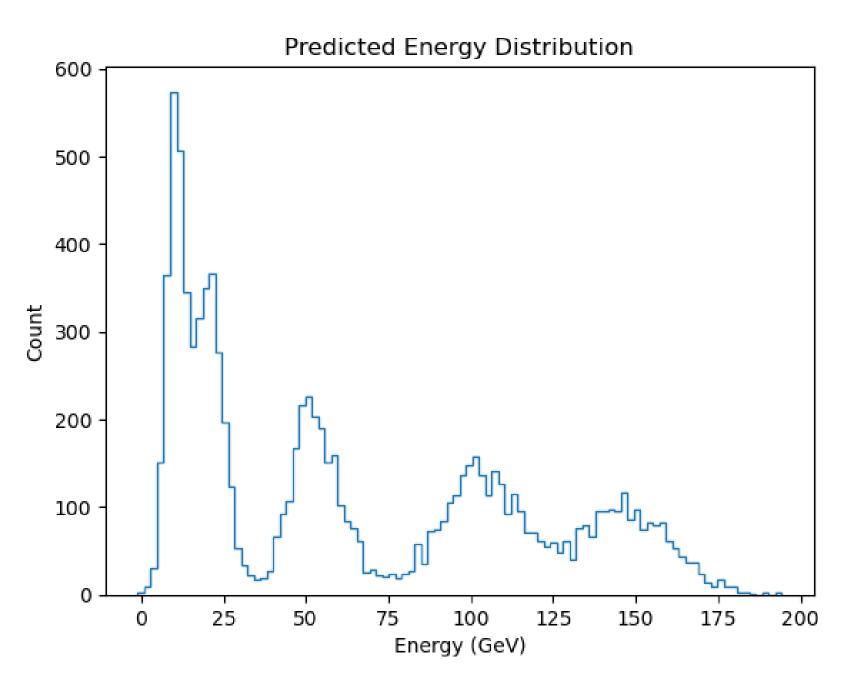
Energy Calibration (Neural Network)

Predicted Energy Distribution



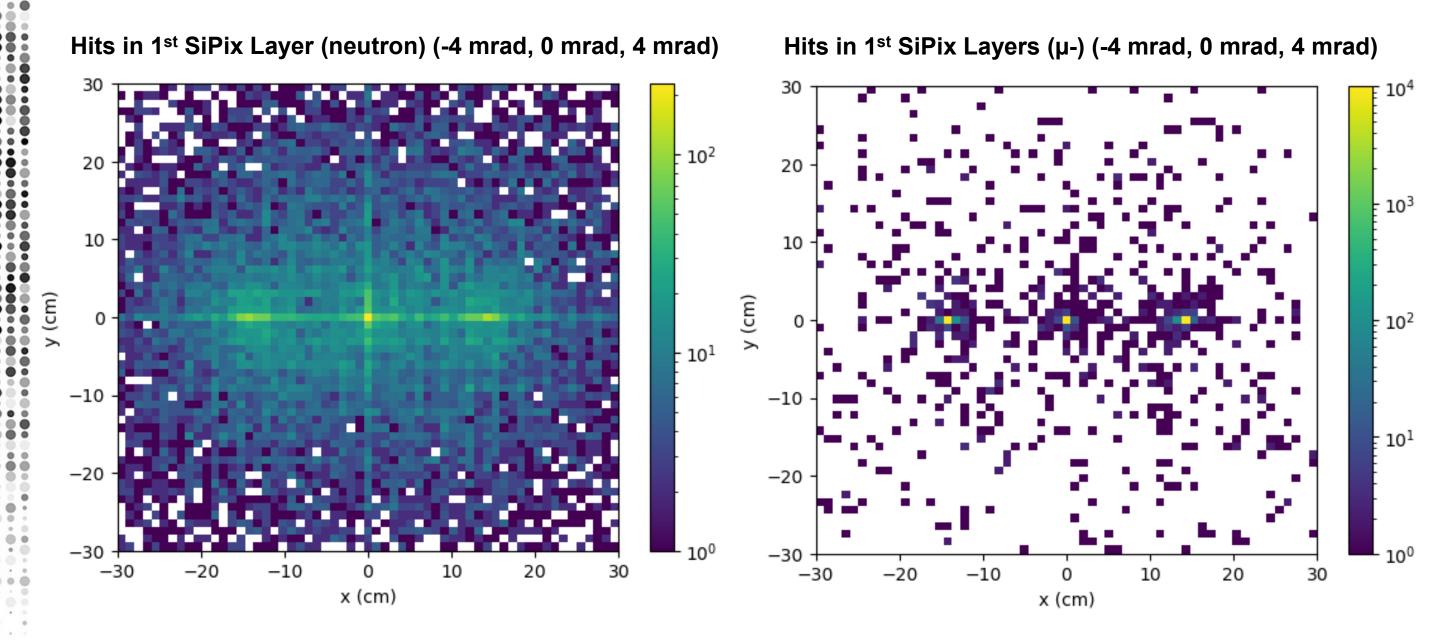


Energy Calibration (Graph Neural Network)



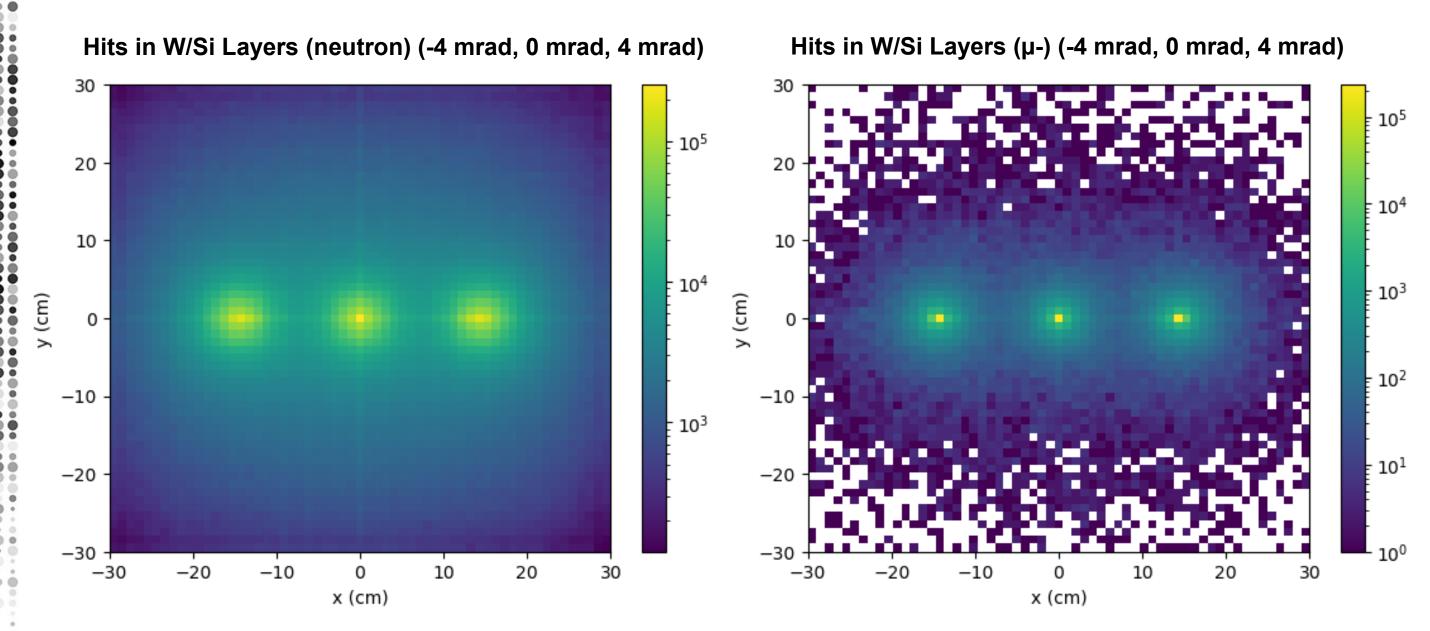


Transverse Distributions





Transverse Distributions





Angle Reconstruction: Methods

- Linear fit between "center of gravity" of hits in first and last layers
 - Cannot reconstruct small angles due to low granularity in Pb/Scint layers
- Linear fit from IP using "center of gravity" in first layer
 - Extremely sensitive to noise (i.e. due to multiple scattering in upstream material)
 - Many particles may pass through the first layer without interacting
- Convolutional Neural Network
 - Spatial neighborhoods do not scale to varying transverse segmentations
 - Inefficient for sparse data
- Graph Neural Network
 - KNN clustering naturally scales for different segmentations
 - Aggregation only utilizes cells with non-zero energy deposition



Linear Fit (Front to Back)

Suppose a particle originates from the interaction point at an angle θ relative to the detector z axis, where θ is the rotation about the y axis. The ZDC has a total length of 162 cm, thus the maximum difference in the transverse position of a particle from the first layer to the last is

$$(\Delta x)_{\text{max}} = 162 \text{ cm} * \tan(\theta)$$

The Pb/Scint layers have a transverse segmentation of $10 \text{ cm} \times 10 \text{ cm}$, thus we have a lower bound on the angular resolution of

$$\arctan\left(\frac{10 \text{ cm}/\sqrt{12}}{162 \text{ cm}}\right) \approx 18 \text{ mrad}$$

(In reality, this value is actually even higher, since the first layer also has a finite spatial resolution of $3 \text{ mm} \times 3 \text{ mm}$)



Linear Fit From IP

Using only the first layer, define the position of the first hit as

$$r_{SiPix} = (\mathbf{x}_1, y_1).$$

Note that for a particle originating at the IP,

$$x_1 = r_{ZDC} * \tan(\theta)$$
,

where $r_{ZDC} = 3550 \text{ cm}$ is the distance from the interaction point to the face of the detector. Then

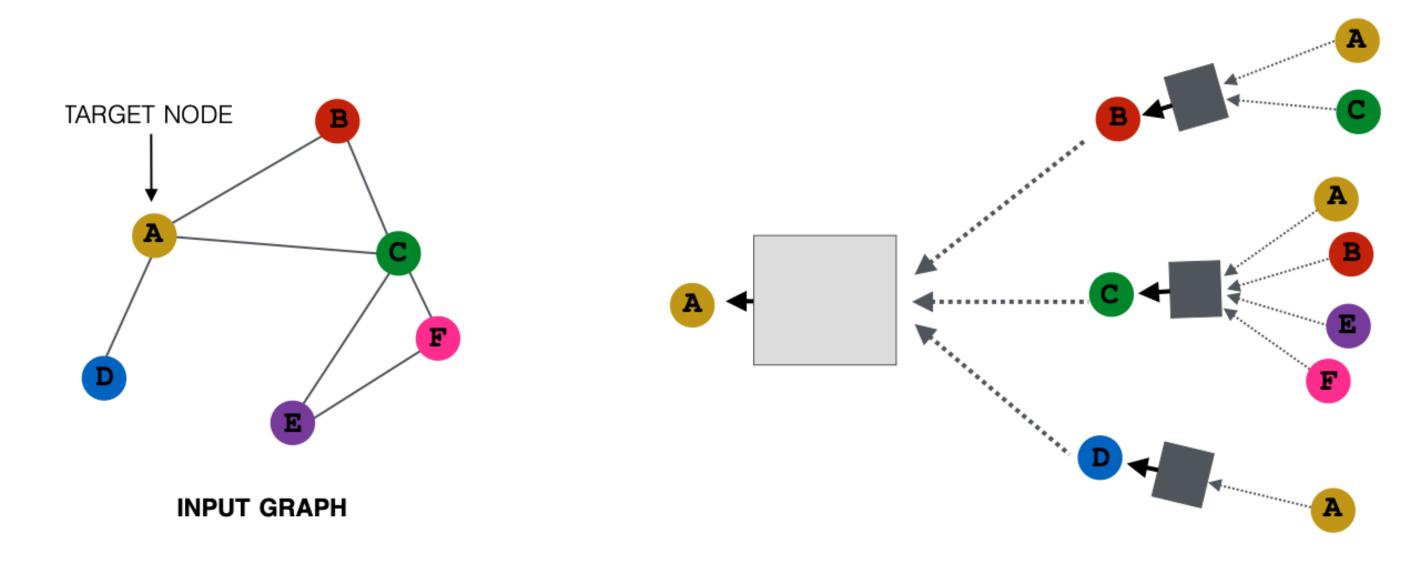
$$x_1 = 3550 \text{ cm } * \tan(\theta).$$

The first silicon layer has a segmentation of $3 \text{ mm} \times 3 \text{ mm}$, so we can estimate the (idealized) angular resolution as

$$\arctan\left(\frac{3 \text{ mm}/\sqrt{12}}{3550 \text{ cm}}\right) = 0.0244 \text{ mrad}$$



Graph Neural Networks





GNN Architecture (GravNet-based)

Better performance by training separate convolutional filters for each subdetector

 Convolutional layer outputs are concatenated and passed to a single linear hidden layer to reconstruct total shower energy

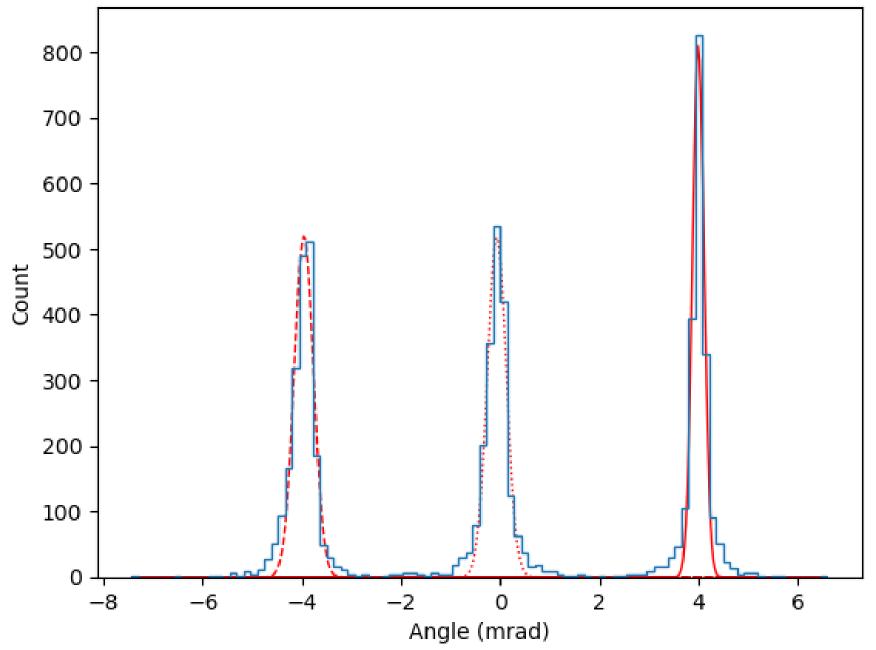
F_{IN} = [xID, yID, layerID, hitEnergy]

- Learned space for KNN clustering has dimension 3 (S = 3)
- Learned space for propagated node features has dimension 1 ($F_{LR} = 1$)



Muon Angle Reconstruction (ZDC Only)

Predicted Angle Distribution



Peak Positions:

 $\mu_1 = -3.96 \text{ mrad}$

 μ_2 = -0.07 mrad

 $\mu_3 = 3.99 \text{ mrad}$

Resolution:

 $\sigma_1 = 0.19 \text{ mrad}$

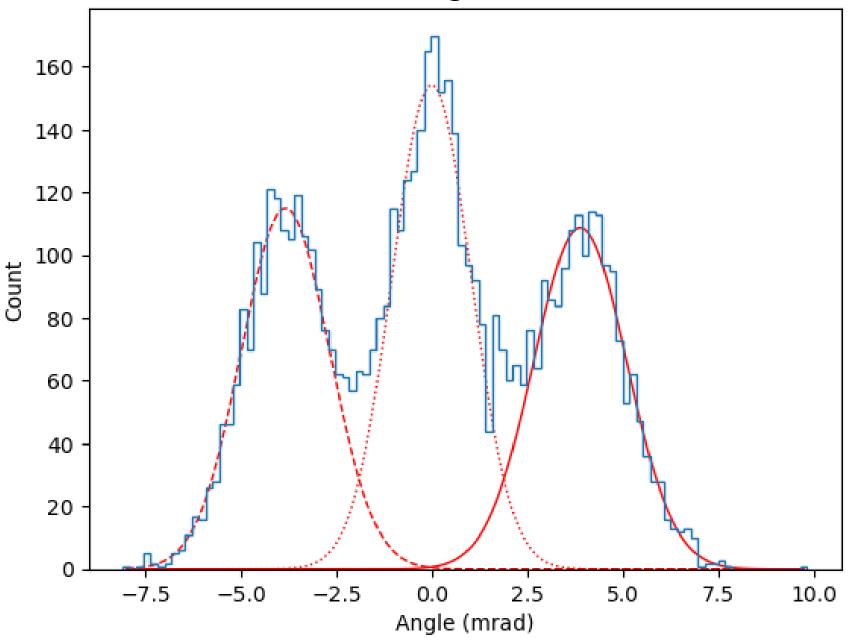
 $\sigma_2 = 0.17 \text{ mrad}$

 $\sigma_3 = 0.12 \text{ mrad}$



Neutron Angle Reconstruction (ZDC Only)

Predicted Angle Distribution



Peak Positions:

 $\mu_1 = -3.84 \text{ mrad}$

 μ_2 = -0.01 mrad

 $\mu_3 = 3.89 \text{ mrad}$

Resolution:

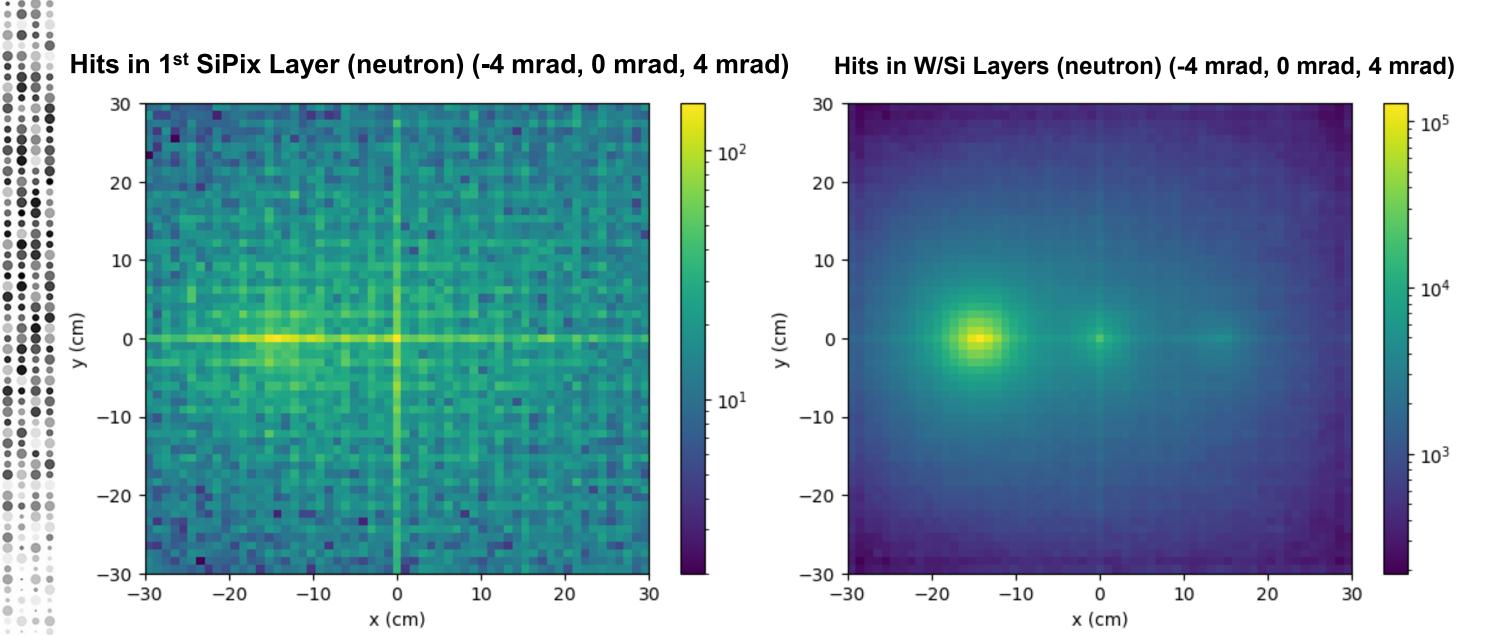
 $\sigma_1 = 1.16 \text{ mrad}$

 $\sigma_2 = 1.03 \text{ mrad}$

 $\sigma_3 = 1.22 \text{ mrad}$



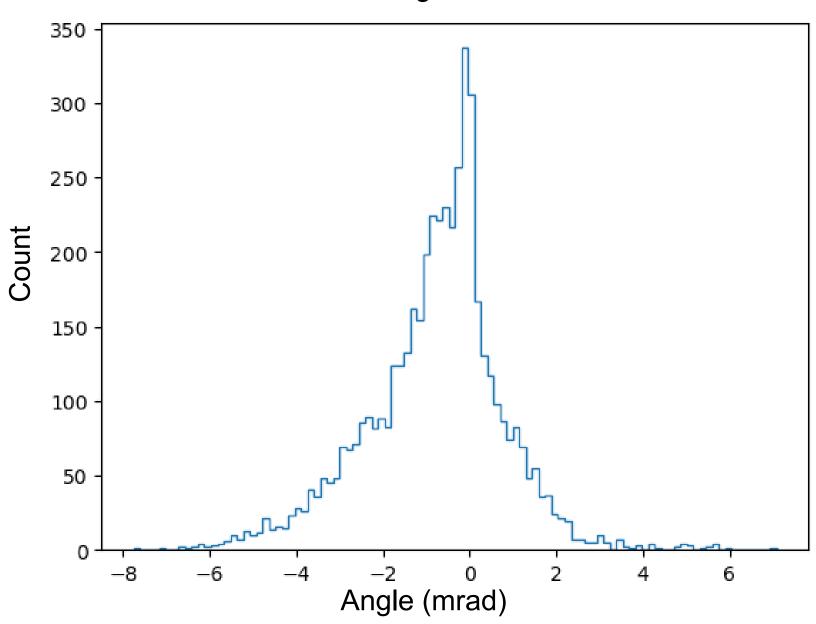
Effects of Upstream Material





Effects of Upstream Material

Predicted Angle Distribution





Summary & Conclusions

- First implementation of a graph neural network for the reconstruction of small angles (≤4 mrad) in the ZDC
 - Can be adapted to a variety of configurations and geometries due to independent treatment of each component
- Investigation of the effects of upstream material on hadronic showers in the ZDC.
 - Emphasis on how interactions with this material impacts our ability to reconstruct the angles of neutrons from the IP



Future Work

- Fine-tune hyperparameters for better optimization
 - Accounting for differences in material properties between components
- Include hit timing information in node feature embedding
 - Important for shower separation with multiple primary particles
- Further quantify effects of upstream material
 - Informs the design & layout of the far-forward region



Thank you