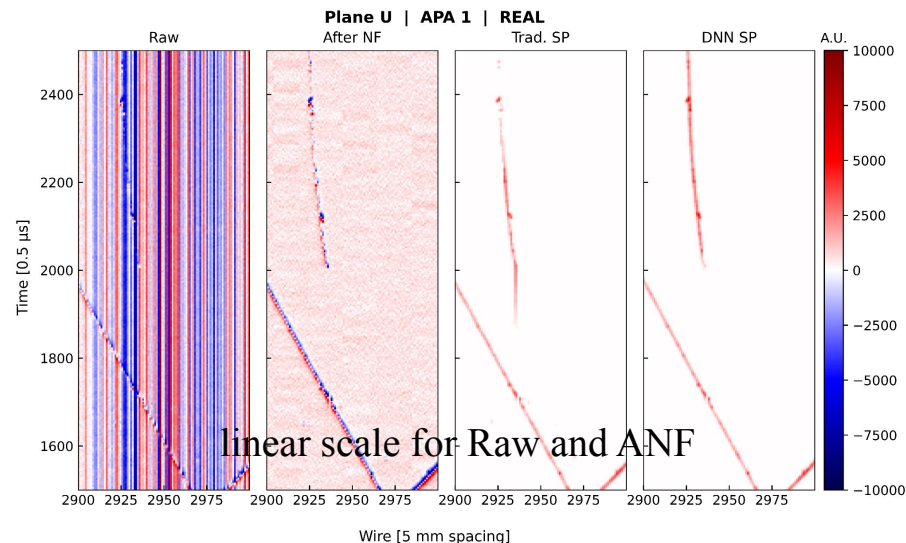
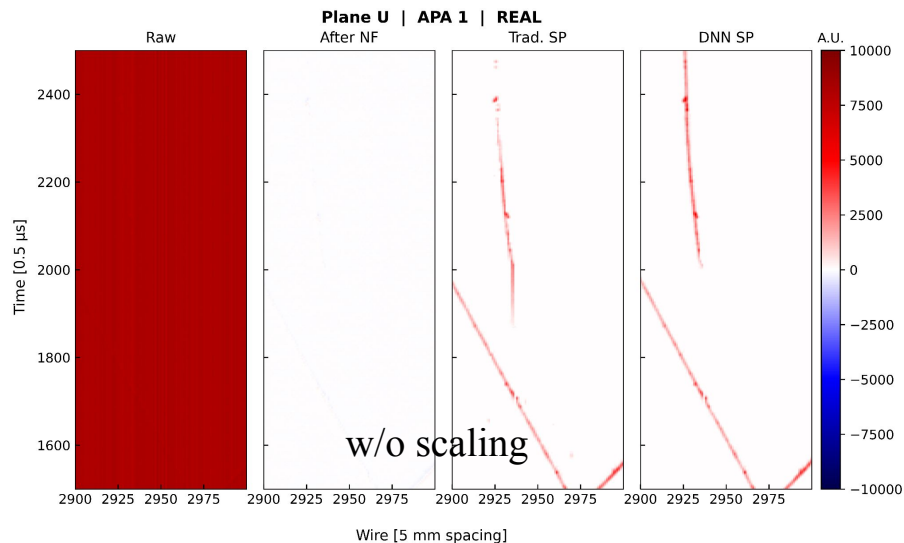




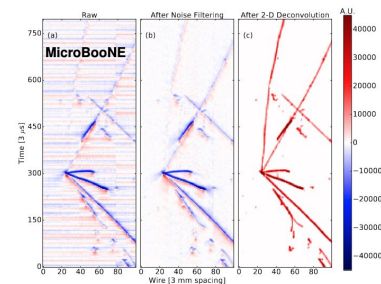
# Status report on **DNNROI sigproc**

Hokyeong Nam  
Chung-Ang University

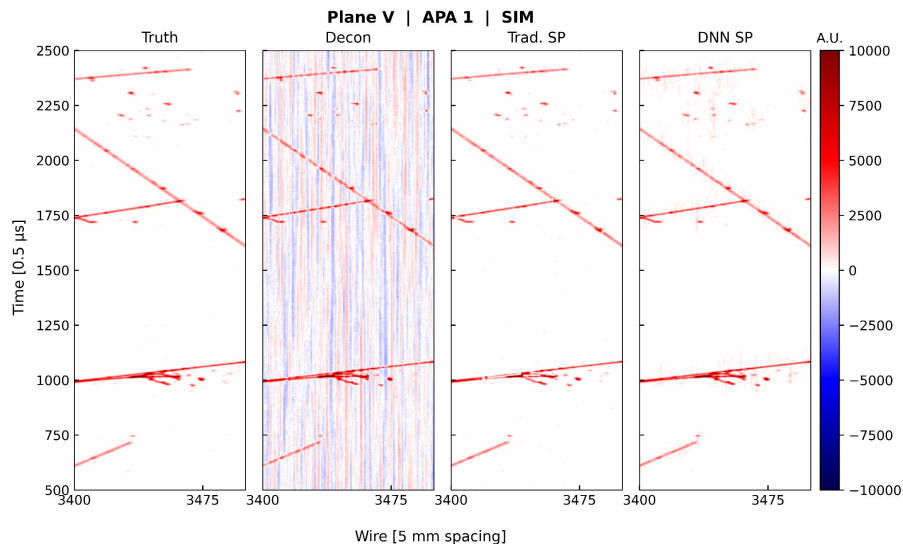
# 2D Waveform Evolution in SigProc



- PD-HD data: 28588-4562
- Wire-Cell version 0.30.2 | LAr version art 3.14.04
- TorchScript:
  - Architecture: MobileU-Net (MobileNetV3 + U-Net)
  - Rebin factor: 10
  - Truth threshold: 10

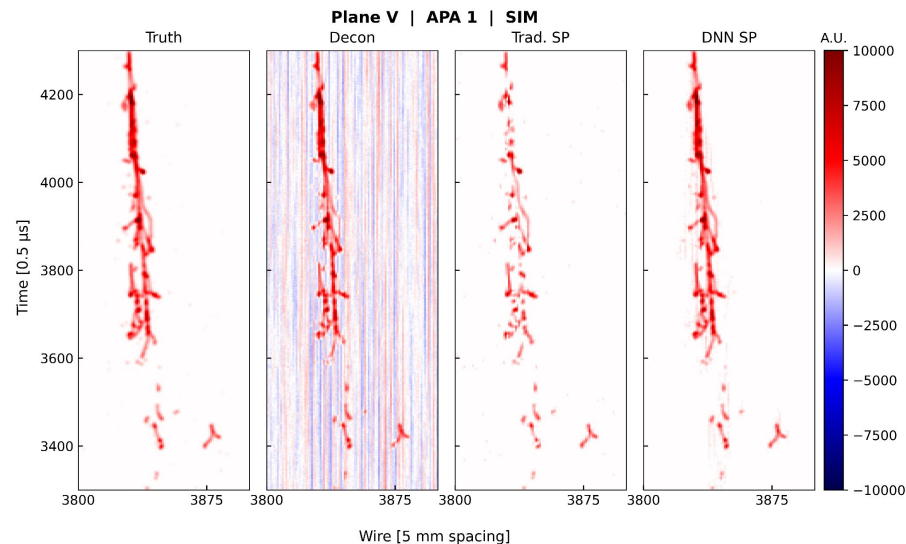


# 2D Waveform Evolution in SigProc



□ Sim: Cosmic-ray

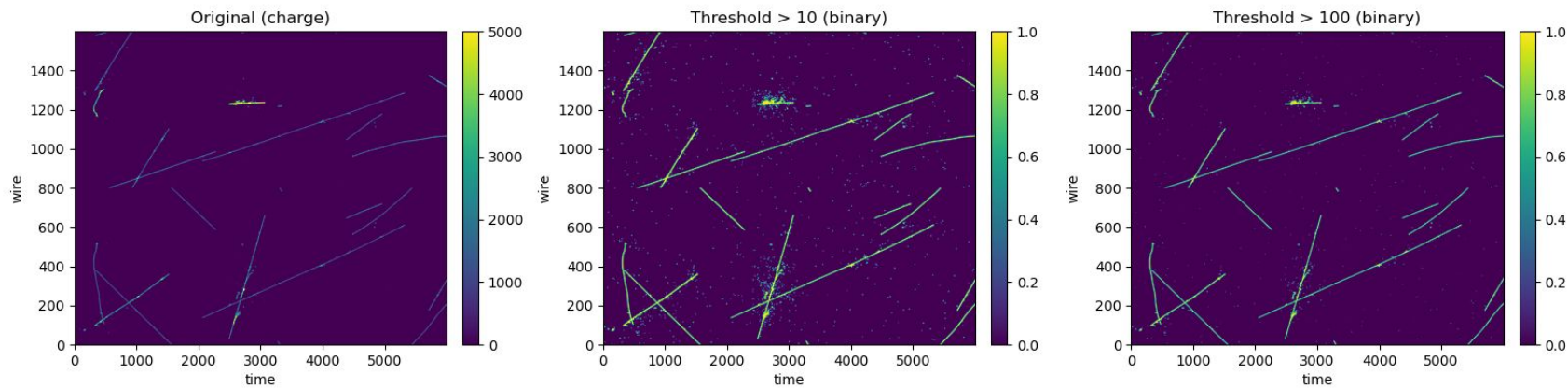
- Wire-Cell version 0.30.2 | LAr version art 3.14.04
- TorchScript:
  - Architecture: MobileU-Net (MobileNetV3 + U-Net)
  - Rebin factor: 10
  - Truth threshold: 10



□ Sim: Single shower (1 GeV)

- DNN ROI
  - Position resolution ↓
  - Charge reconstruction ↑

# True ROI definition in training script

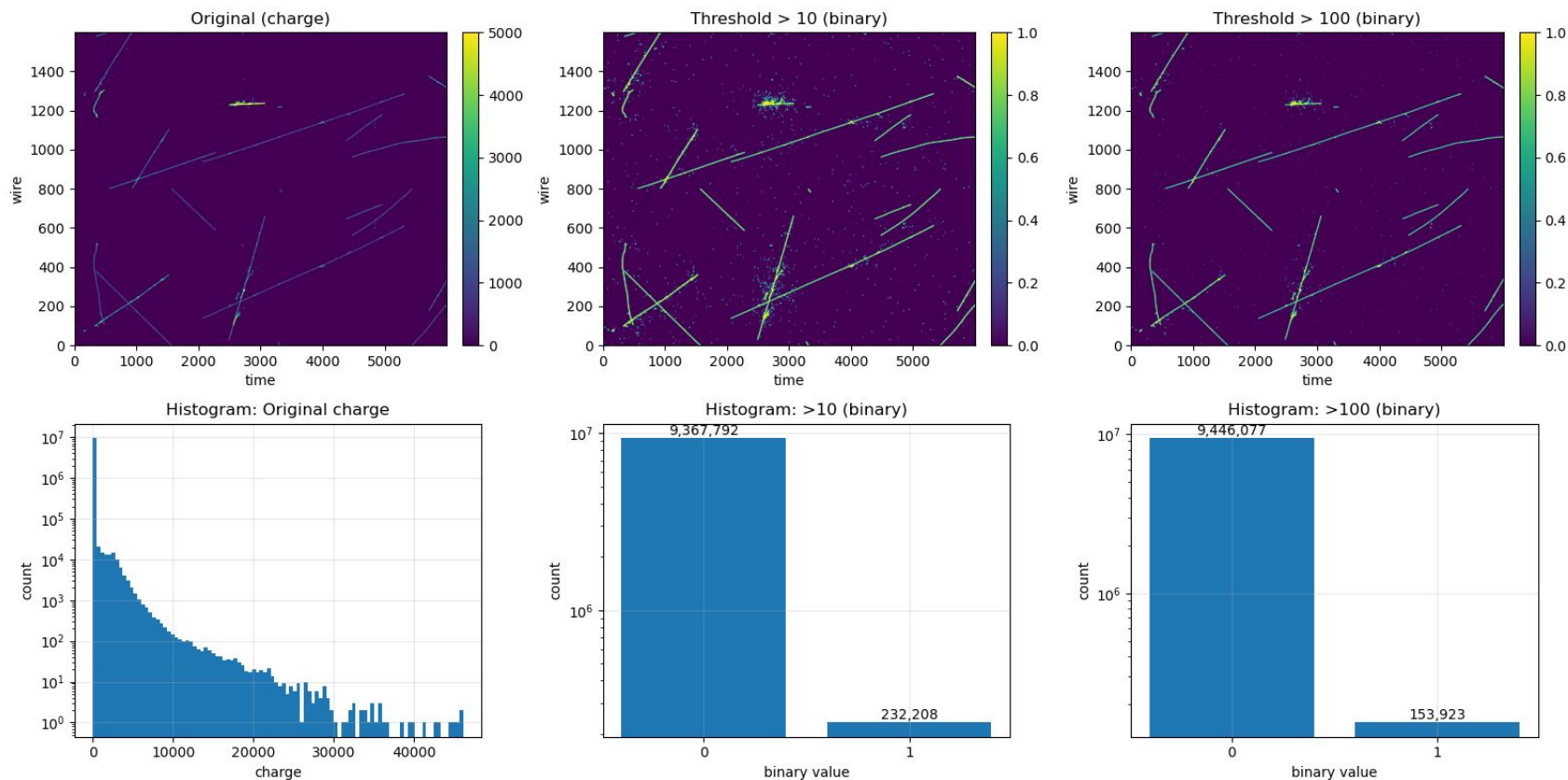


True ROI: Convert continuous charge map into binary mask for training (for BCE loss)

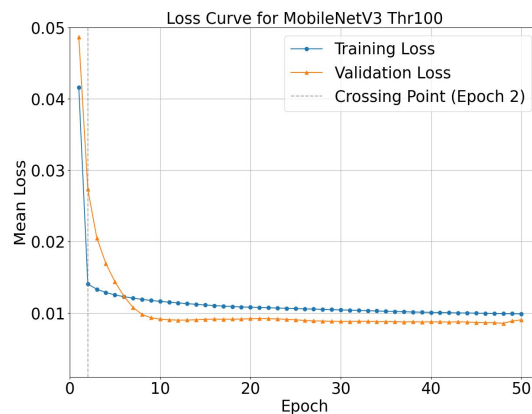
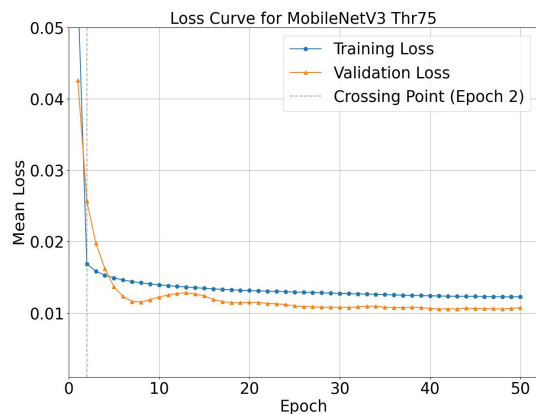
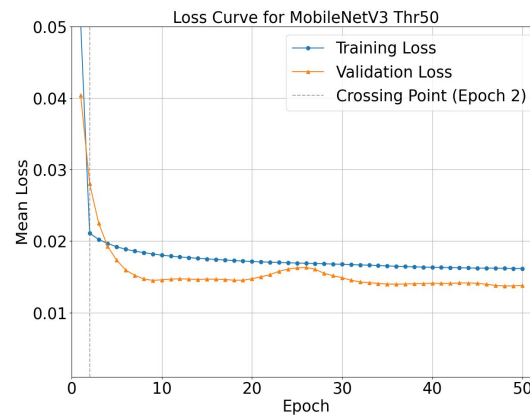
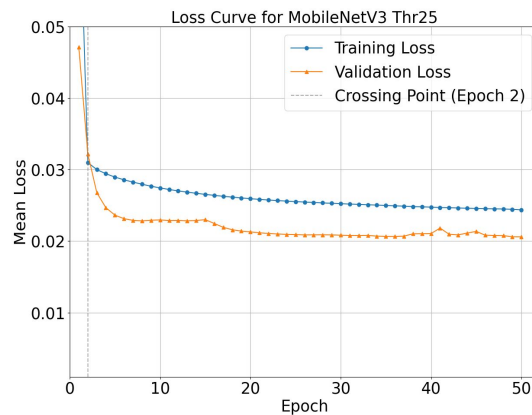
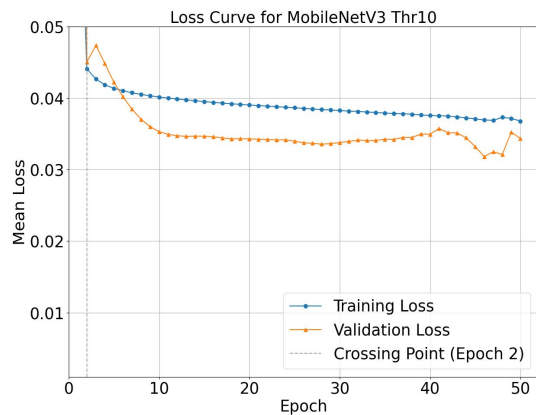
- Raw Truth Data: each pixel = deposited charge at (wire, time)
- Rebin & Crop: downsample in time using rebin factor, and select region for training
- Thresholding over bin content (truth\_th in code):
  - If charge > threshold  $\rightarrow$  mask as signal (1)
  - If charge  $\leq$  threshold  $\rightarrow$  background (0)
- Low threshold  $\rightarrow$  more pixels labeled as signal (recall  $\uparrow$ , precision  $\downarrow$ )
- High threshold  $\rightarrow$  fewer pixels labeled as signal (recall  $\downarrow$ , precision  $\uparrow$ )



# True ROI definition in training script

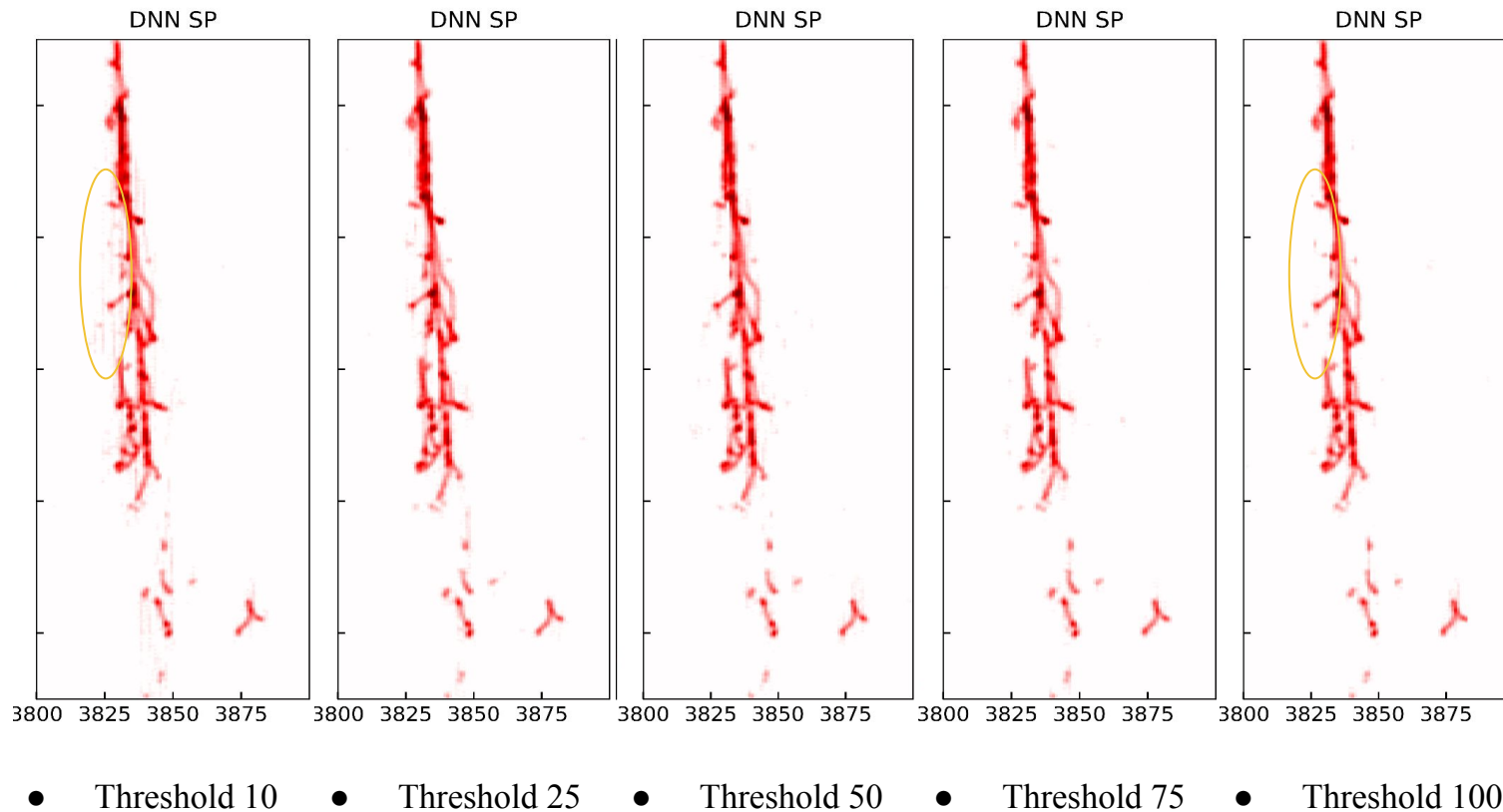


# MobileU-Net (V3) - Train vs. Val loss with different truth threshold

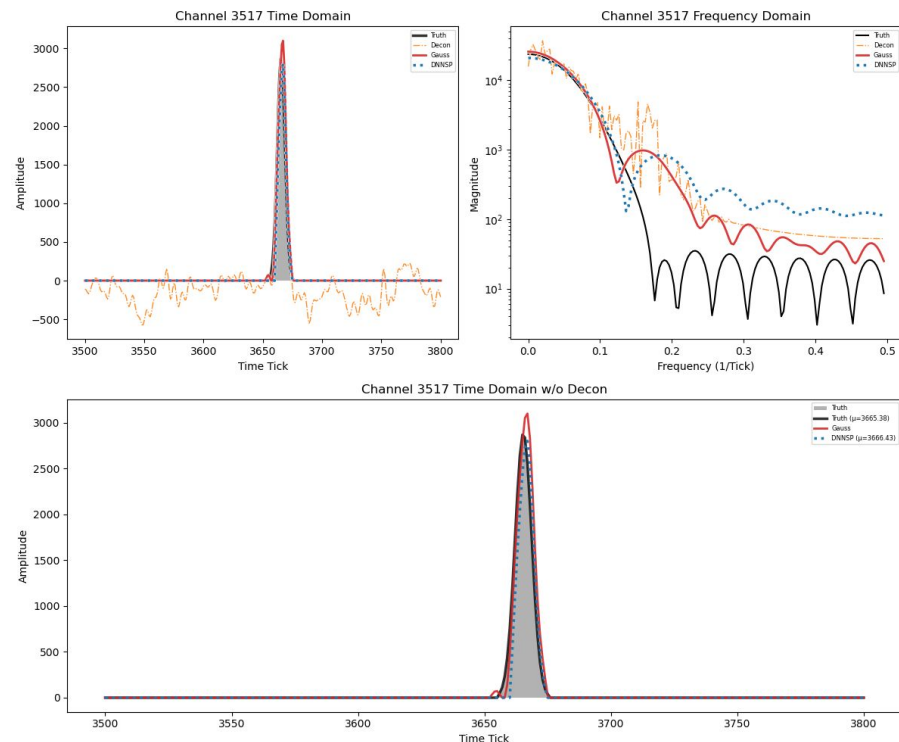
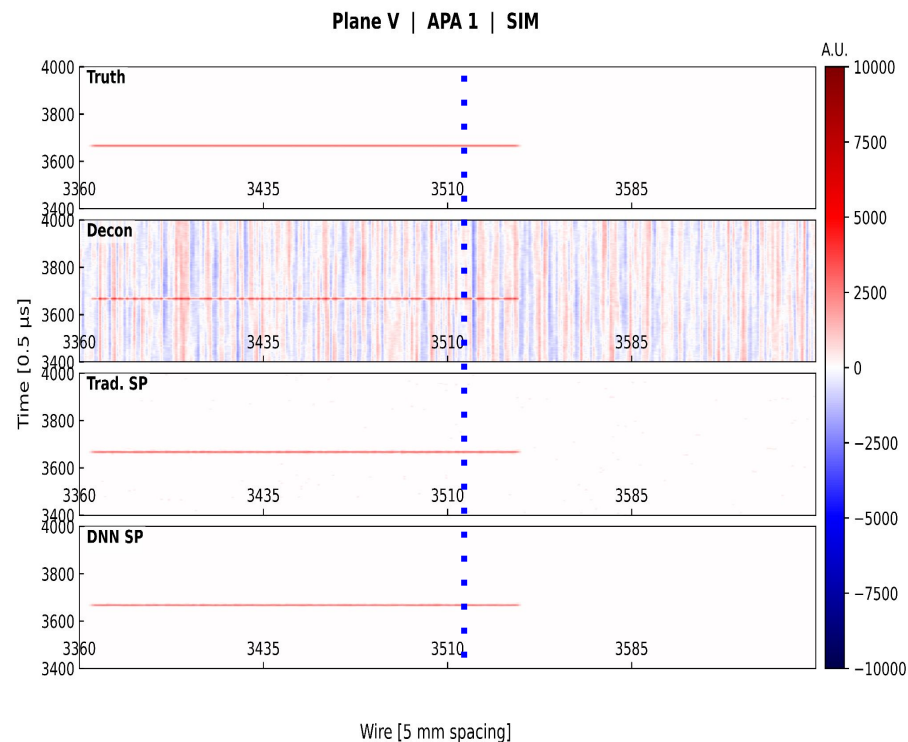


Threshold	Train Loss
10	0.0367
25	0.0243
50	0.0161
75	0.0122
100	0.0098

## 2D Waveform with Different Truth Threshold

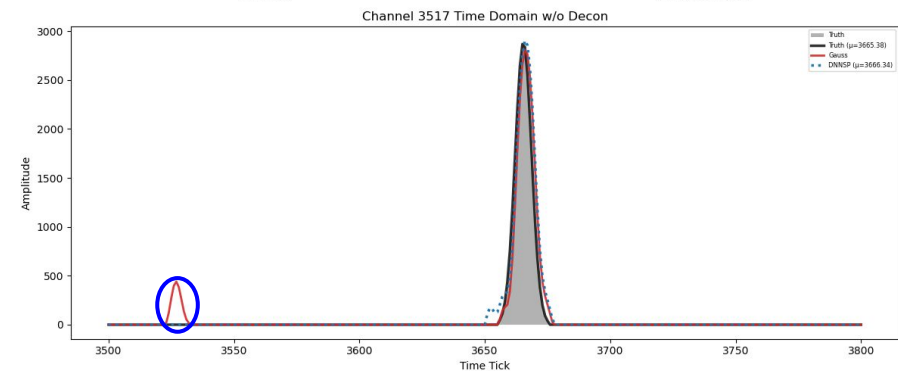
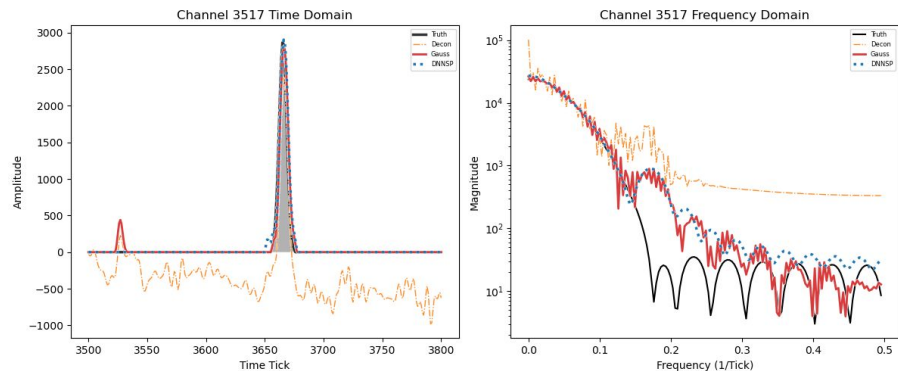


# Waveform from Isochronous Track

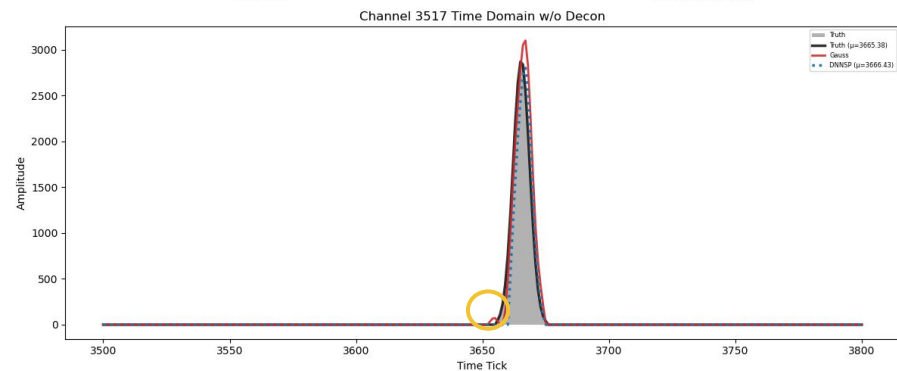
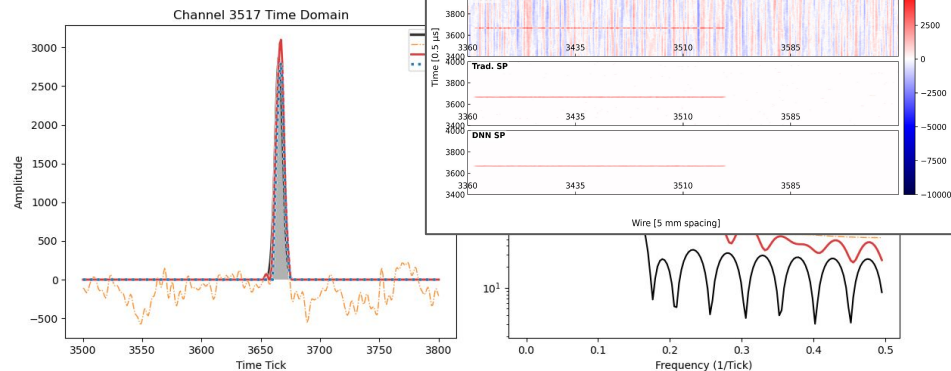


- Both methods demonstrated good charge reconstruction

# Waveform from Isochronous Track

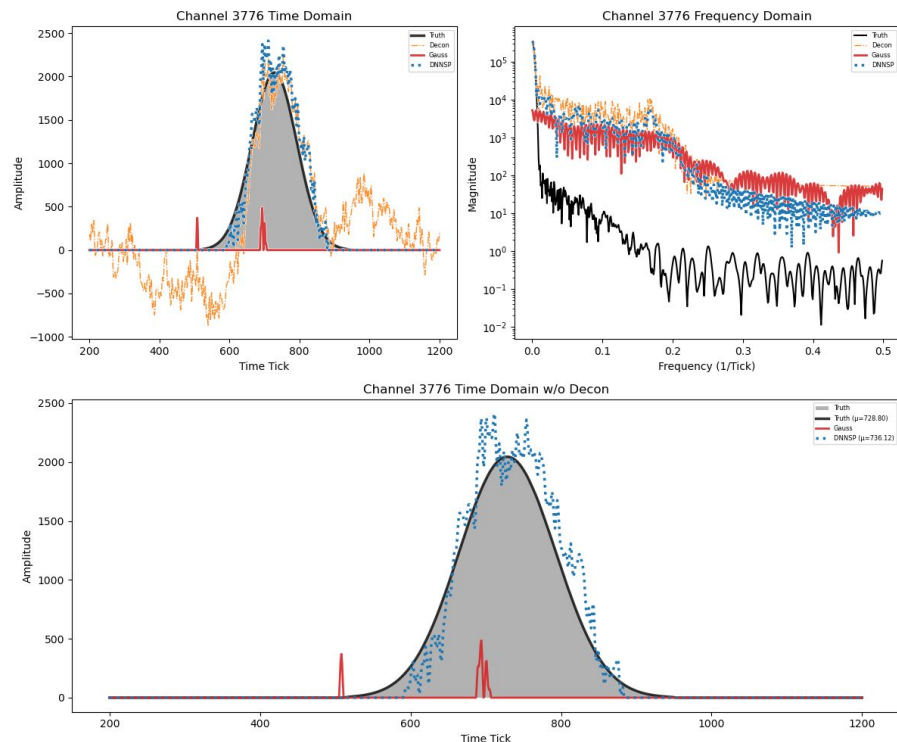
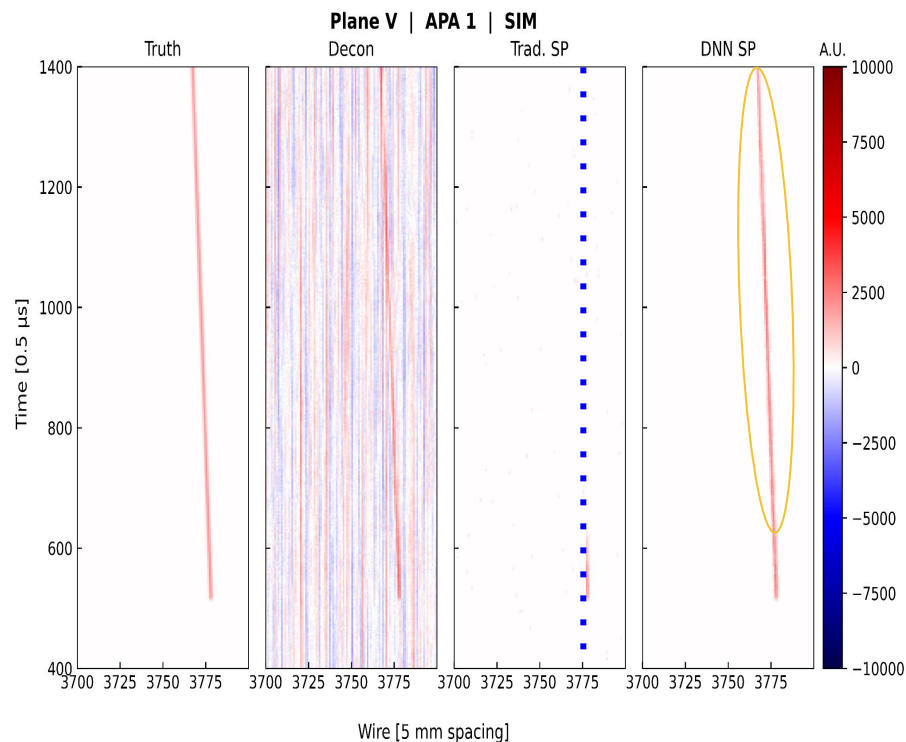


- Threshold 10
- Higher truth\_th has narrowed ROI window
- Traditional ROI more frequently picks up fake signals at time ticks that have no truth



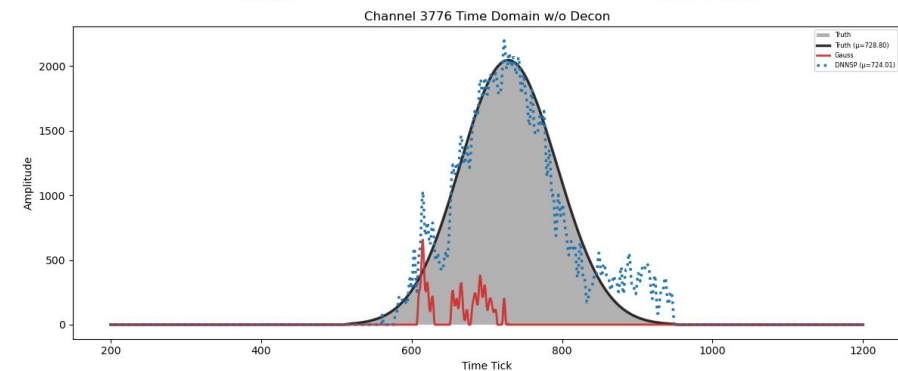
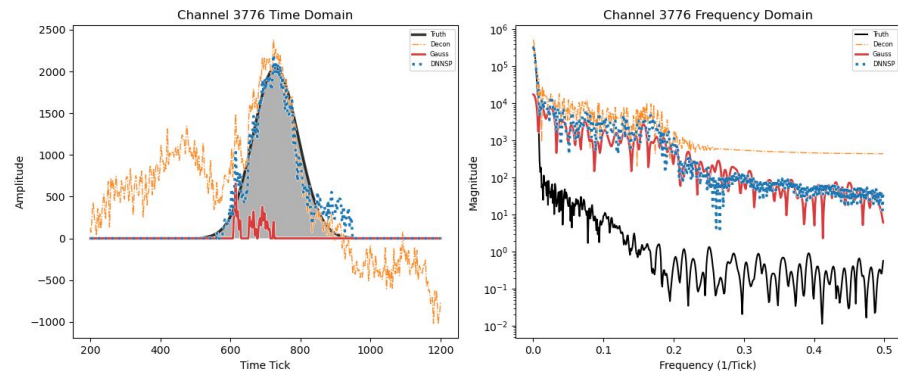
- Threshold 100

# Waveform from Prolonged Track



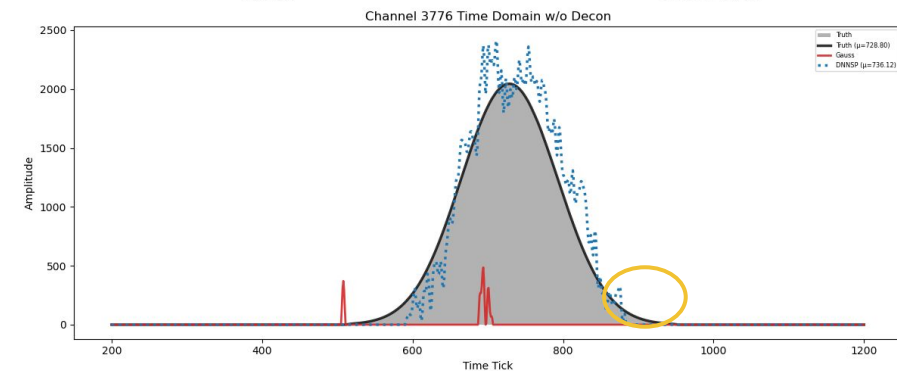
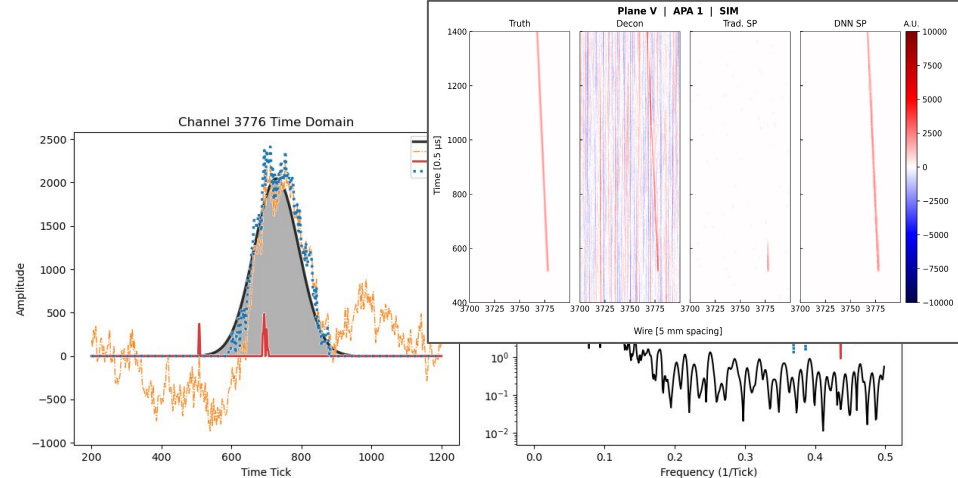
- In this case, the DNN ROI reconstructed the charge much better

# Waveform from Prolonged Track



- Threshold 10

- For prolonged track, traditional ROI barely reconstruct charge

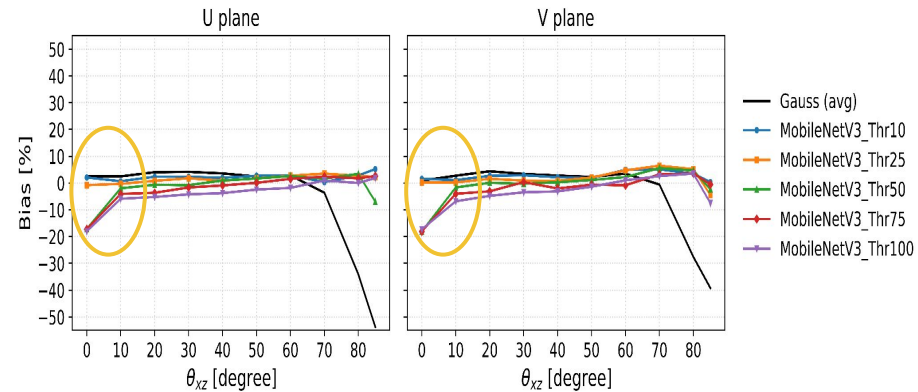


- Threshold 100

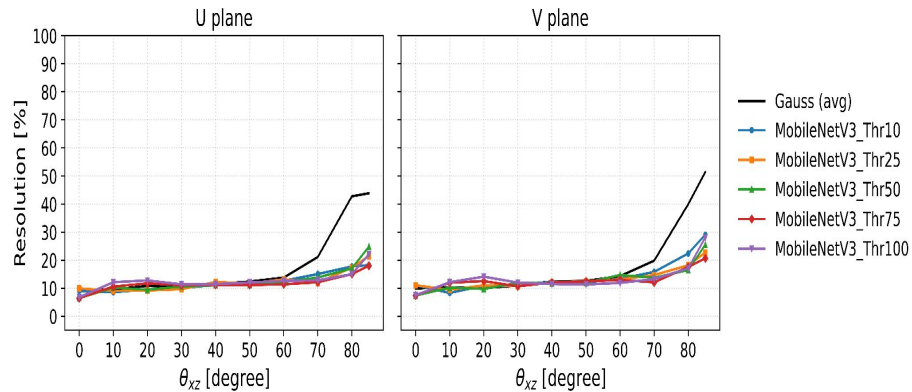


# Single Track Evaluation

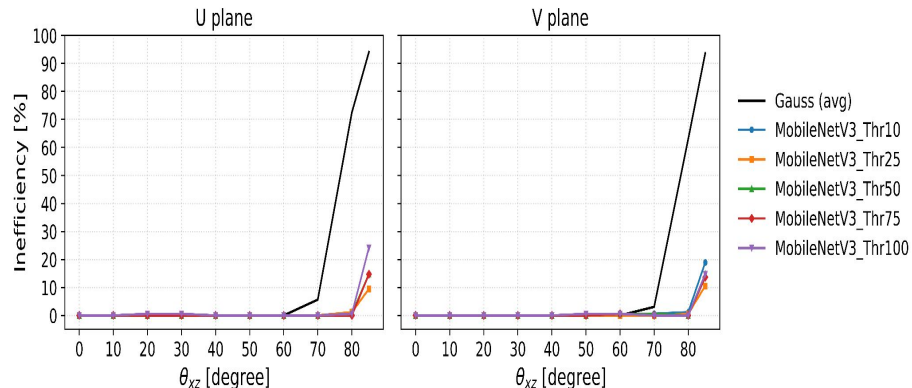
Bias vs  $\theta_{xz}$  | n500 | DNN vs Gauss(avg)



Resolution vs  $\theta_{xz}$  | n500 | DNN vs Gauss(avg)



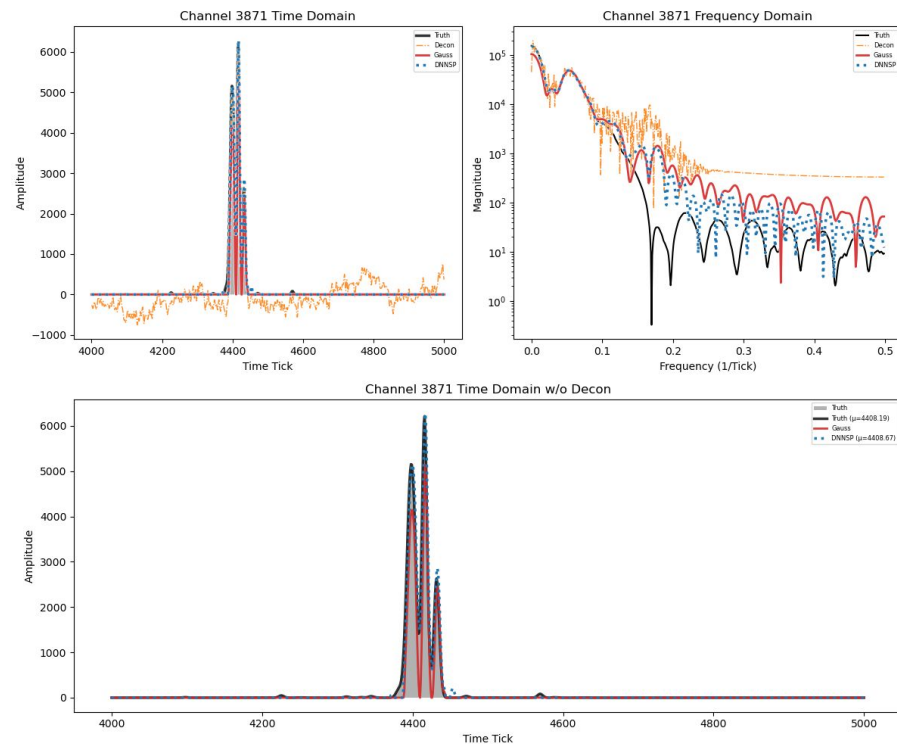
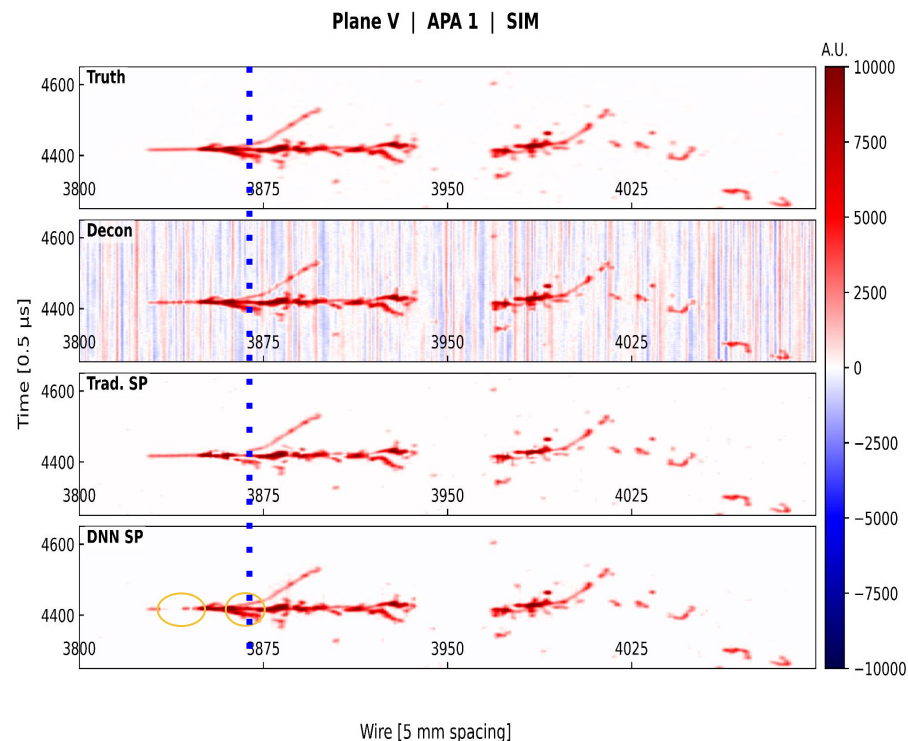
Inefficiency vs  $\theta_{xz}$  | n500 | DNN vs Gauss(avg)



- Models with `thruth_th > 50` showed about -20% bias in the low angle region
- Possible causes (from visual inspection of waveforms)
  - ROI window is slightly smaller** than the truth
  - Peak charge amplitude is lower** than the truth

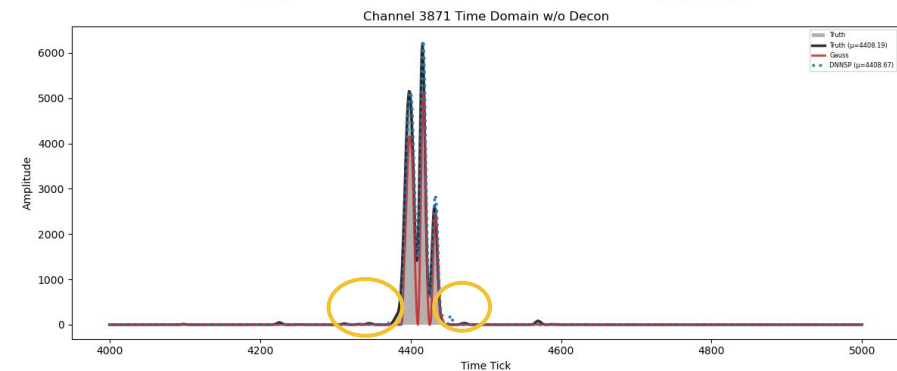
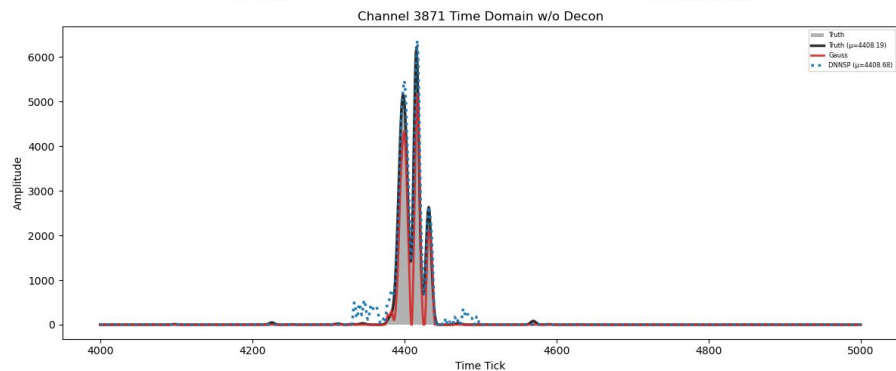
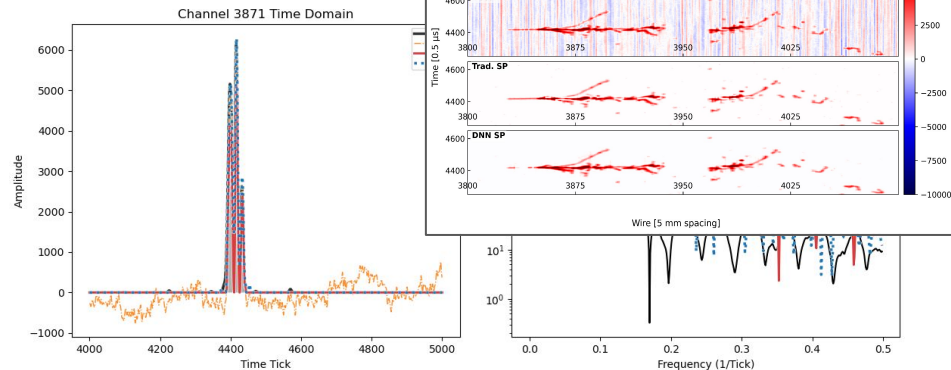
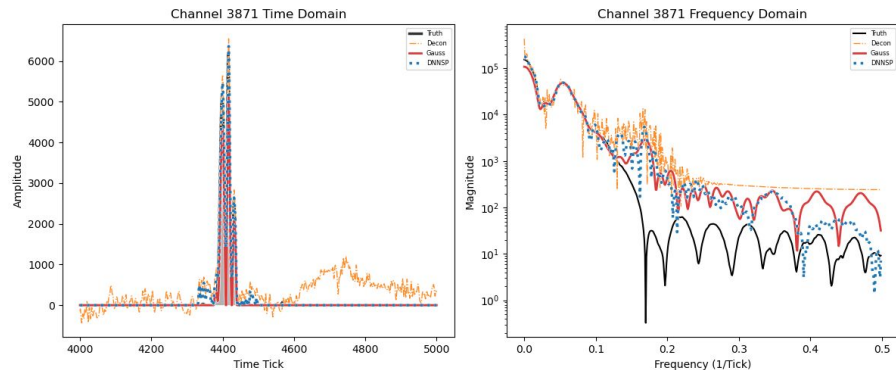
(See slide 19 for supporting plots)

# Waveform from Isochronous Shower



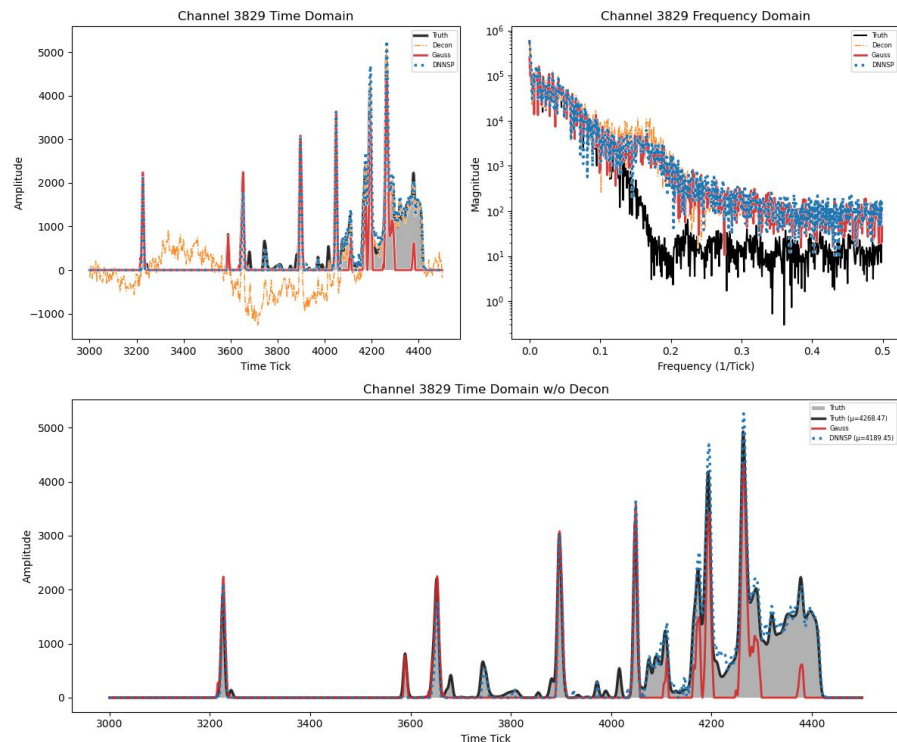
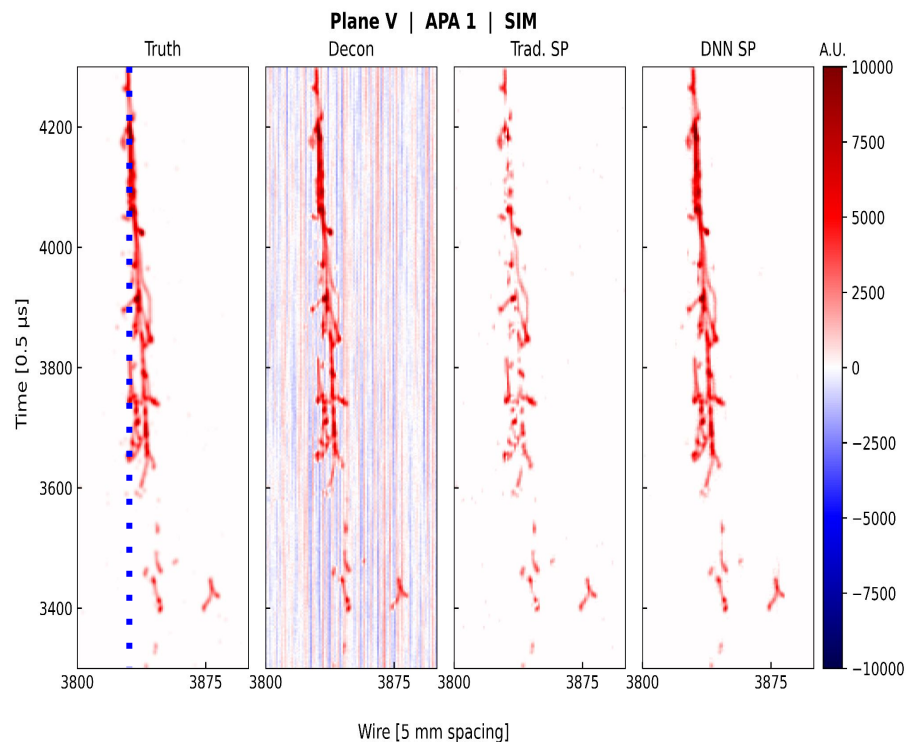
- While the DNN ROI reconstructed the charge effectively in regions with complex waveforms, certain portions were not reconstructed as the truth threshold became higher

# Waveform from Isochronous Shower



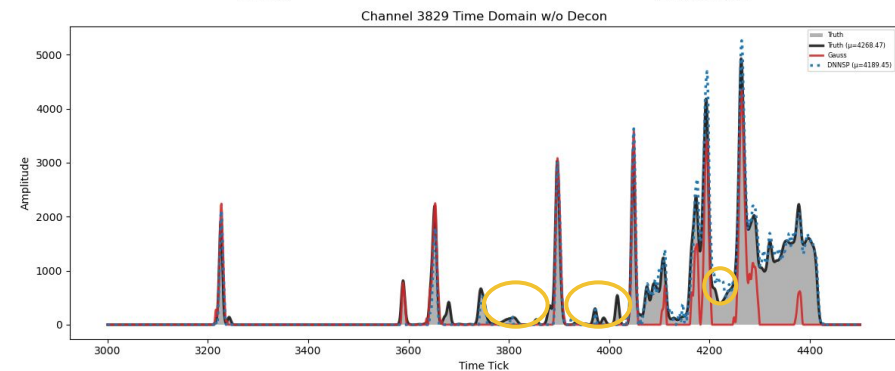
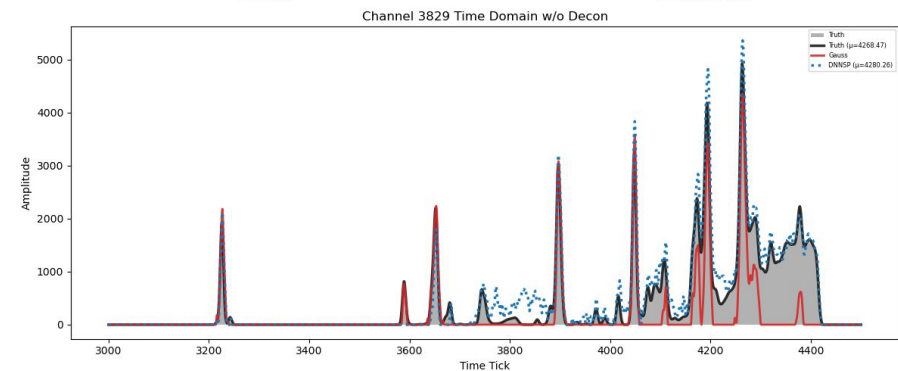
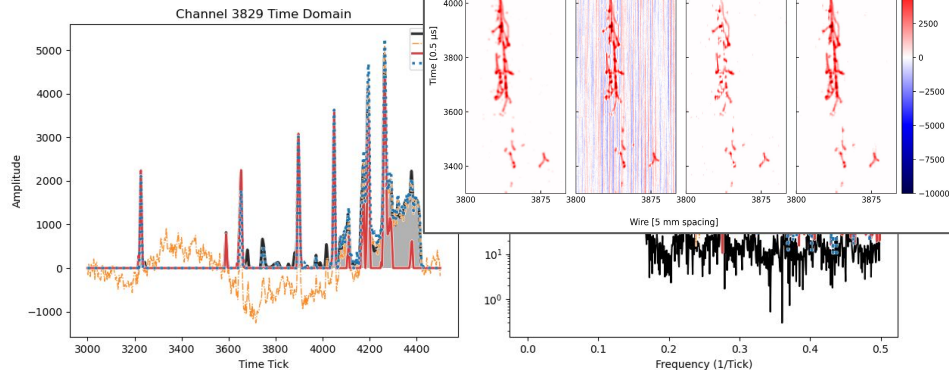
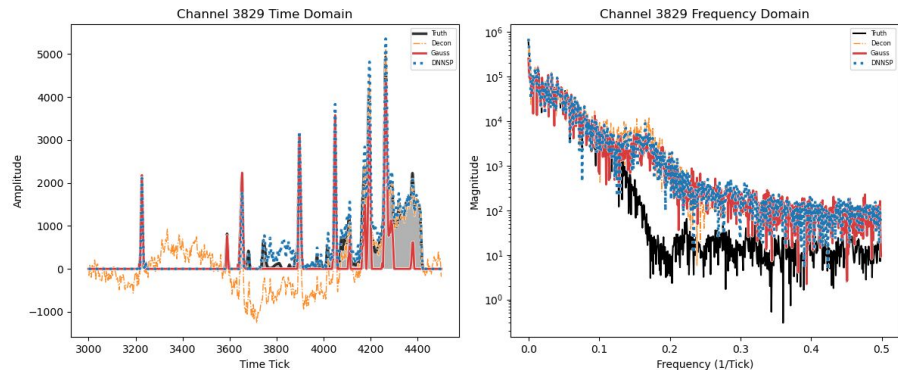
- Threshold 10
- Threshold 100
- For the same g4.root file, the decon result differs (How to have same decon shape?)

# Waveform from Prolonged Shower



- In this case, the DNN ROI reconstructed the charge much better

# Waveform from Prolonged Shower

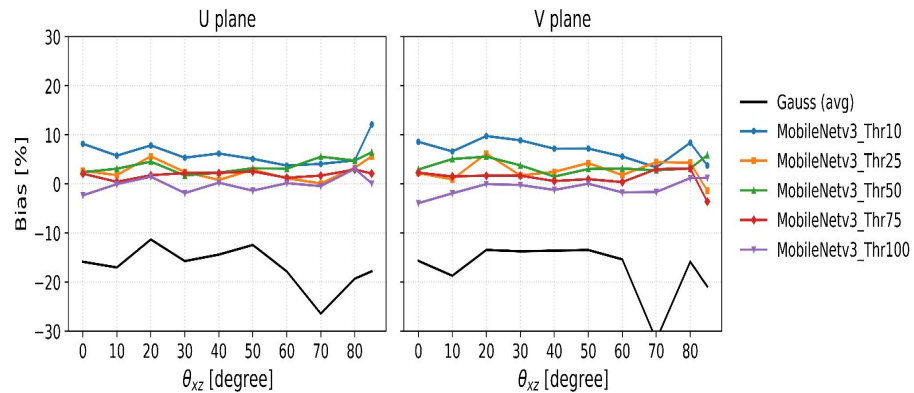


- Threshold 10
- Threshold 100
- The decon is almost identical in this case, higher truth\_th succeeded to decrease the noise

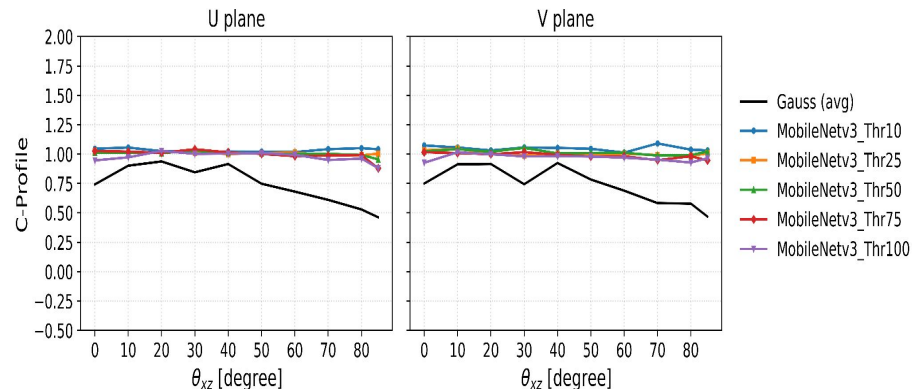


# Single Shower Evaluation (1 GeV)

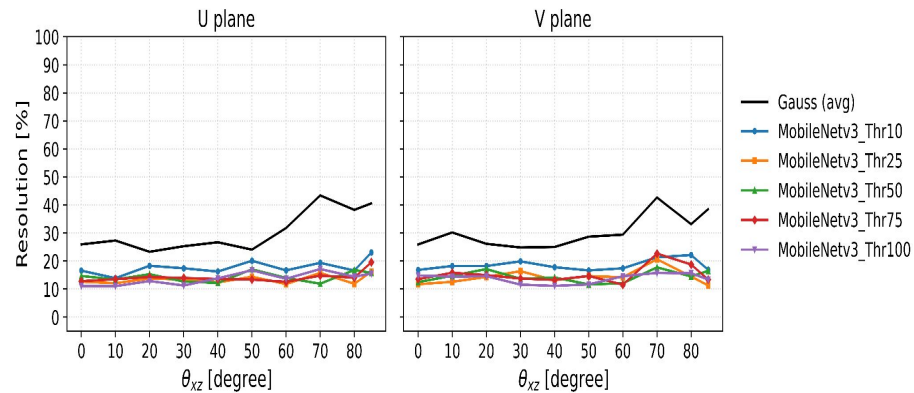
Bias vs  $\theta_{xz}$  | 1GeV | DNN vs Gauss(avg)



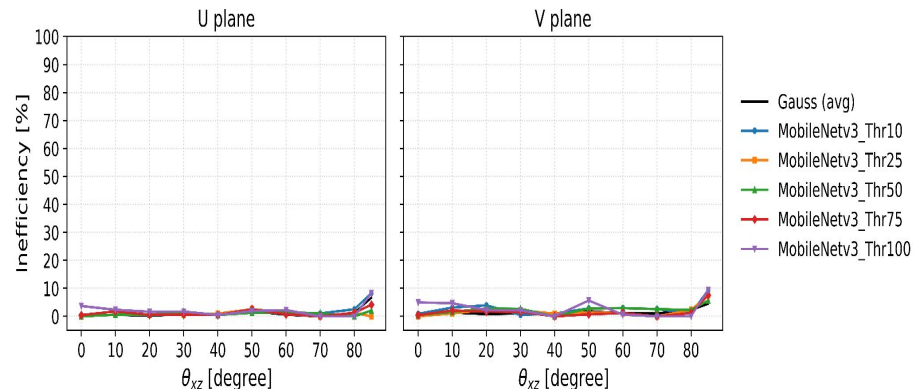
Cprofile vs  $\theta_{xz}$  | 1GeV | DNN vs Gauss(avg)



Resolution vs  $\theta_{xz}$  | 1GeV | DNN vs Gauss(avg)



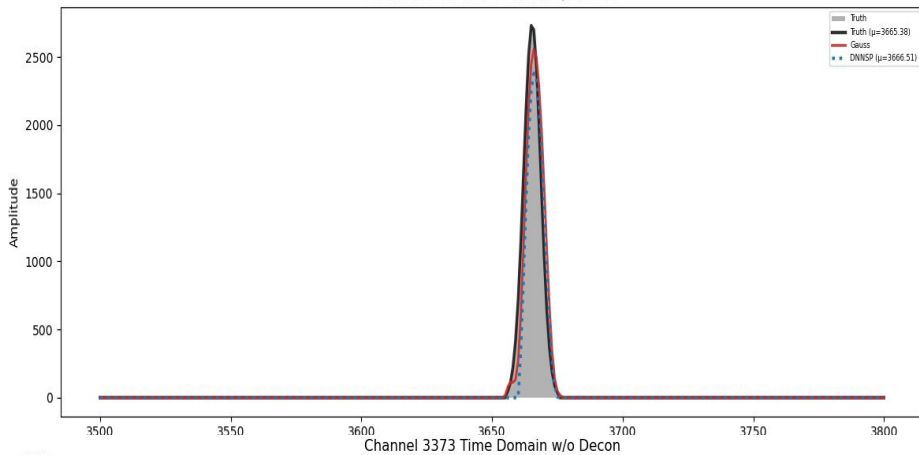
Inefficiency vs  $\theta_{xz}$  | 1GeV | DNN vs Gauss(avg)



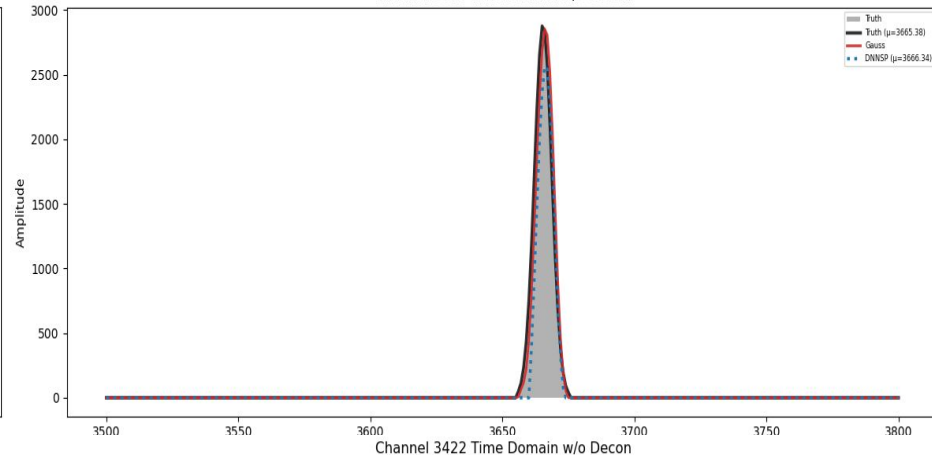
# Back Up



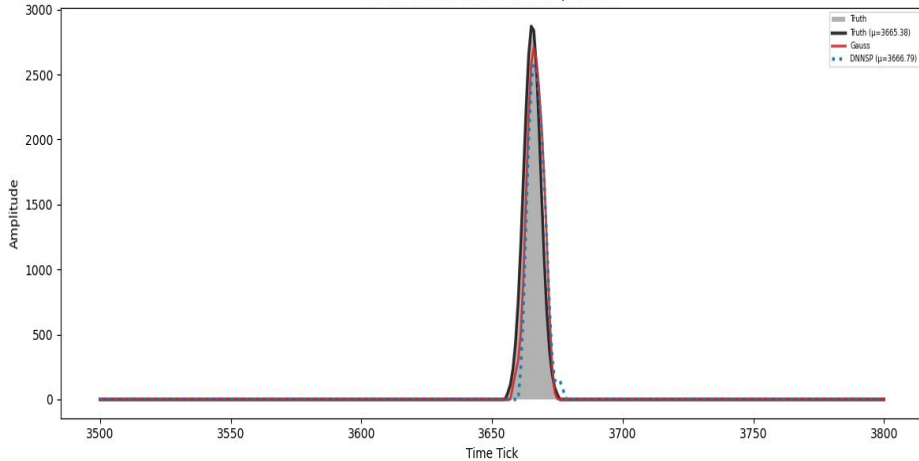
Channel 3367 Time Domain w/o Decon



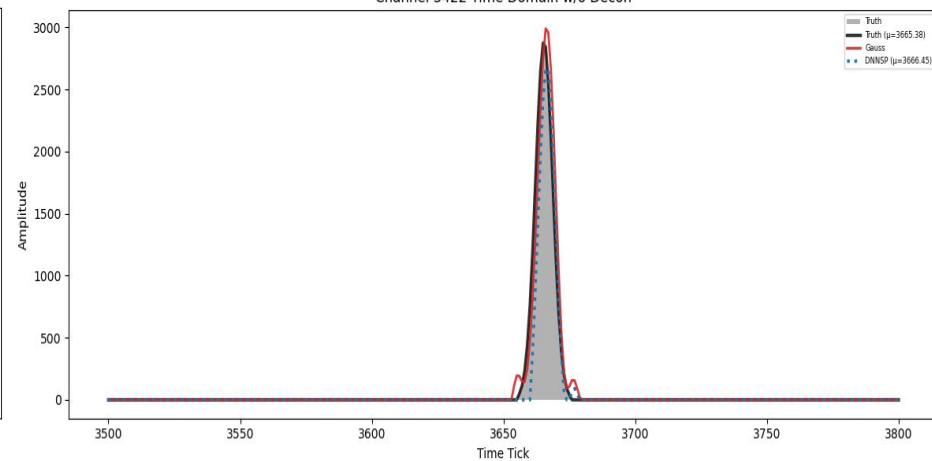
Channel 3378 Time Domain w/o Decon



Channel 3373 Time Domain w/o Decon



Channel 3422 Time Domain w/o Decon



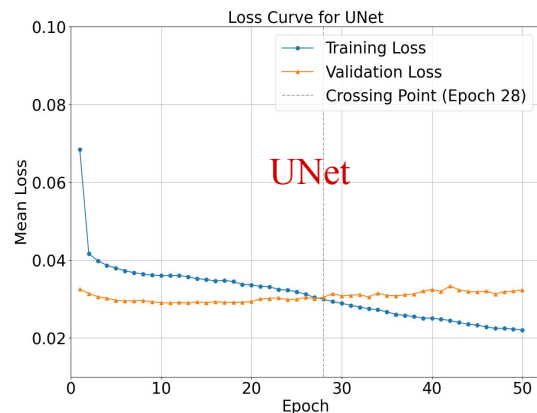
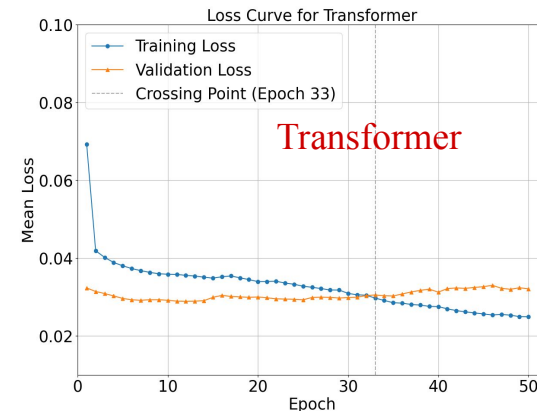
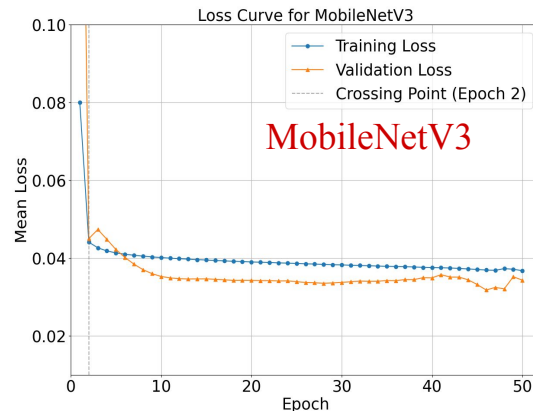
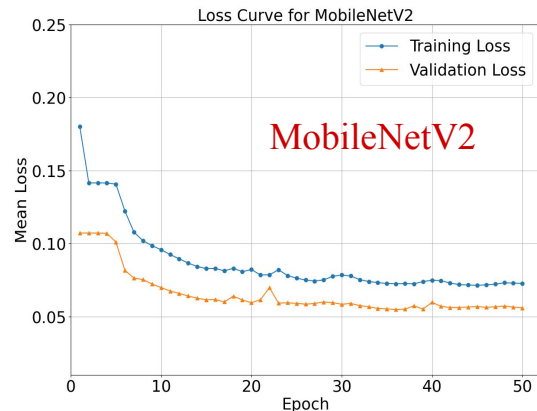
# Model Comparison - Network Architectures

Model	Encoder	Decoder	Skip Connection	Downsampling Depth	Activations	
					Encoder	Decoder
UNet	Convs	4 Convs	Yes	1/16	ReLU	ReLU
MobileNetV2-UNet	Depthwise separable	2 Convs	None	1/32	ReLU6	ReLU
MobileNetV3-UNet	Depthwise separable + SE	4 Convs	Yes	1/32	h-swish + ReLU	ReLU
Transformer-UNet	Convs + Transformer bottleneck	4 Convs	Yes	1/16	ReLU	ReLU
					GELU (transformer)	

- Training dataset: 590 cosmic-ray events
- Optimizer: SGD (Stochastic Gradient Descent)
- Learning rate: 0.1
- Early stopping: Enabled
- Train/Val split: 0.9/0.1
- Loss: BCELoss (Binary Cross-Entropy Loss)
- Number of epochs : 50
- Output activation function: Sigmoid

**Training was carried out on the WC Cluster using an NVIDIA GeForce RTX 4090 GPU (24 GB)**

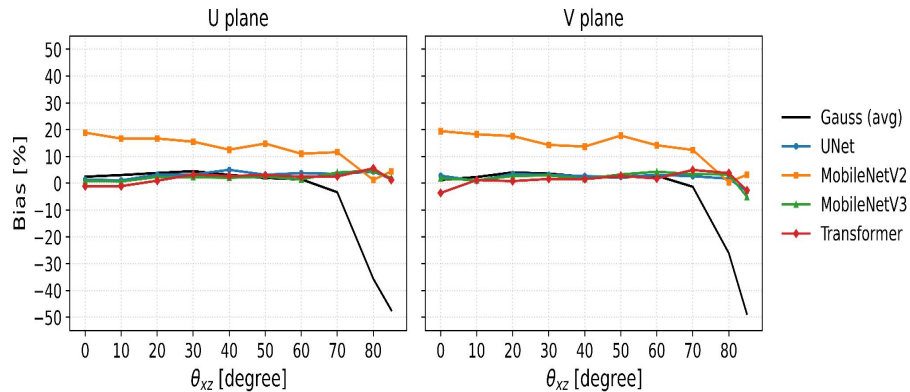
# Model Comparison - Train vs Val loss



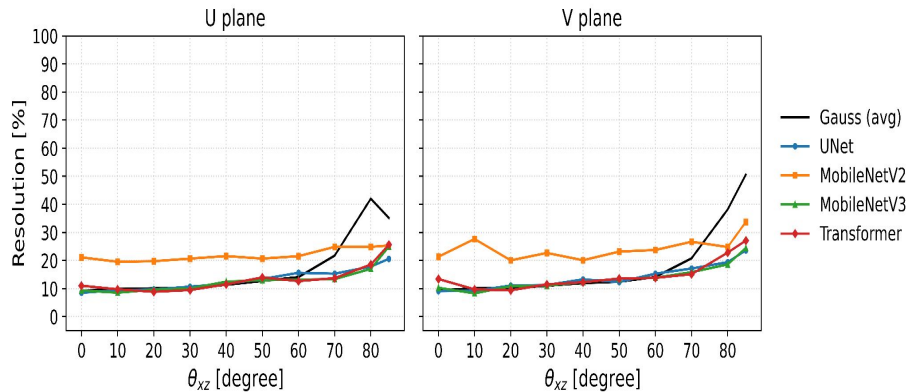
- Rebin factor was set to 10 during training
- Training losses at the selected checkpoints:
  - UNet: 0.029 (Epoch 28)
  - MobileNetV2: 0.072 (Epoch 50)
  - MobileNetV3: 0.044 (Epoch 50)
  - Transformer: 0.029 (Epoch 32)
- Among the models, UNet and Transformer reached the most stable convergence with the lowest final losses

# Single Track Evaluation

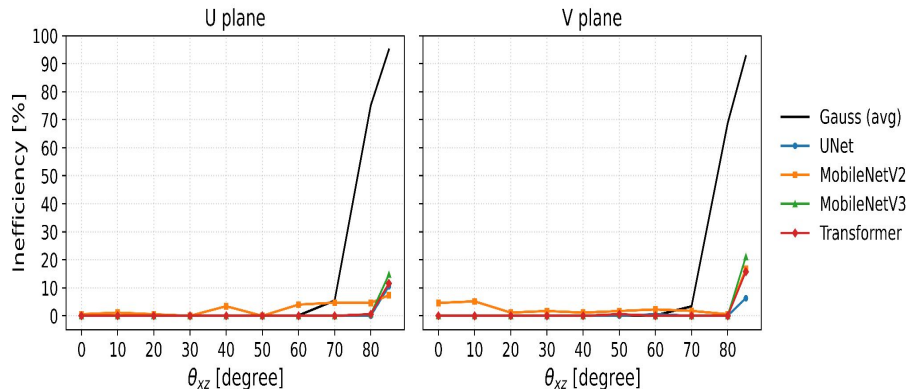
Bias vs  $\theta_{xz}$  | n500 | DNN vs Gauss(avg)



Resolution vs  $\theta_{xz}$  | n500 | DNN vs Gauss(avg)



Inefficiency vs  $\theta_{xz}$  | n500 | DNN vs Gauss(avg)



- Simulations on the normal APA (2nd)
- MobileNetV2 exhibits ~20% bias at low angles
- The lower performance is likely due to:
  - Fewer convolution blocks in the decoder
  - Absence of skip connections
- Other models demonstrate comparable performance

# DNN-ROI Performance Evaluation

- For track events, three metrics are used: Bias, Resolution, and Inefficiency

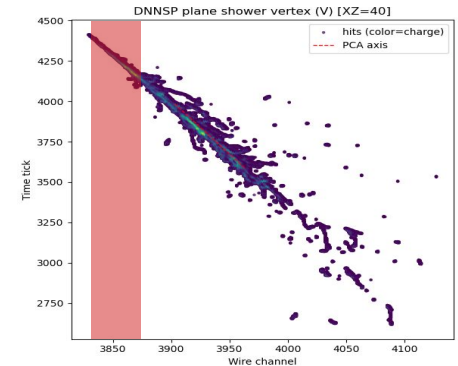
$$Bias = 100 \times \left( \left\langle \frac{Q_{reco}}{Q_{truth}} \right\rangle - 1 \right) \quad Resolution = 100 \times \frac{RMS\left(\frac{Q_{reco}}{Q_{truth}}\right)}{\left\langle \frac{Q_{reco}}{Q_{truth}} \right\rangle} \quad Inefficiency = 100 \times \frac{Number\ of\ bad\ channels}{Number\ of\ valid\ truth\ channels}$$

- For shower events, a charge profile based on vertex information was added as the fourth metric:
  - Sum the charge along the shower direction up to 42 wire channels ( $\approx 1-2$  radiation lengths)
  - Compare the reconstructed-to-truth ratio charge ratio

$$Q_{method} = \sum_{w \in W} Q_{method}(w) \quad R_{cprofile} = \frac{Q_{reco}}{Q_{truth}}$$

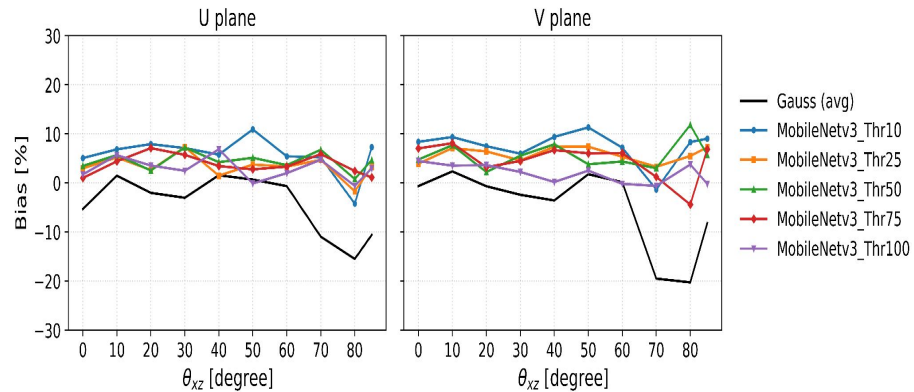
- Samples were generated with

- Detector configuration: ProtoDUNE - Horizontal Drift (PD-HD)
- XZ angle:  $0^\circ$ ,  $10^\circ$ ,  $20^\circ$ ,  $30^\circ$ ,  $40^\circ$ ,  $50^\circ$ ,  $60^\circ$ ,  $70^\circ$ ,  $80^\circ$ ,  $85^\circ$
- Shower energies: 100 MeV, 500 MeV, 1 GeV, 2 GeV, 3 GeV, 5 GeV
- Software: WCT standalone (Tracks), LAr-WCT (Showers)

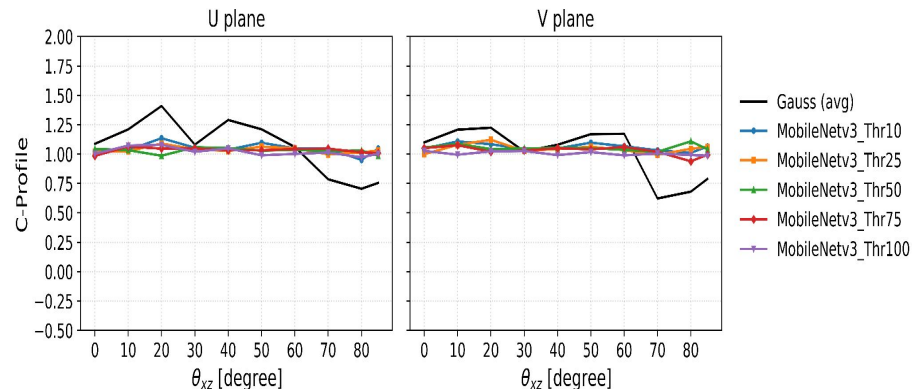


# Single Shower Evaluation (Thr 100, 100 MeV)

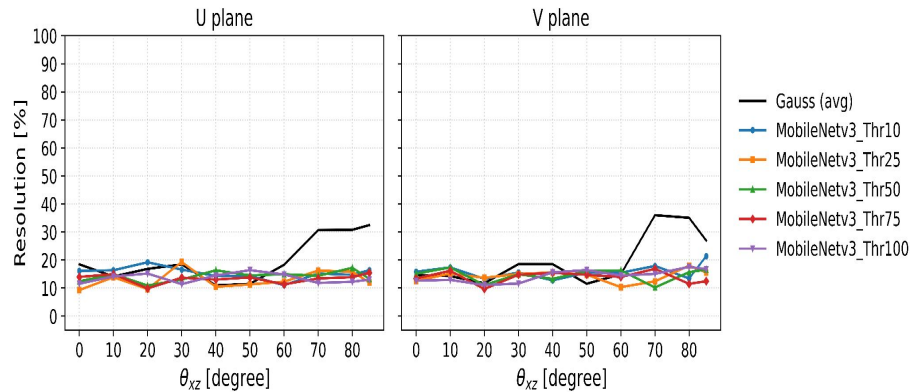
Bias vs  $\theta_{xz}$  | 100MeV | DNN vs Gauss(avg)



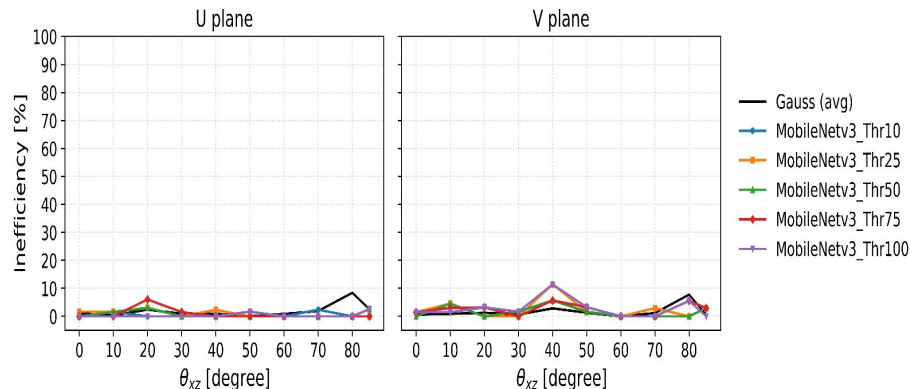
Cprofile vs  $\theta_{xz}$  | 100MeV | DNN vs Gauss(avg)



Resolution vs  $\theta_{xz}$  | 100MeV | DNN vs Gauss(avg)

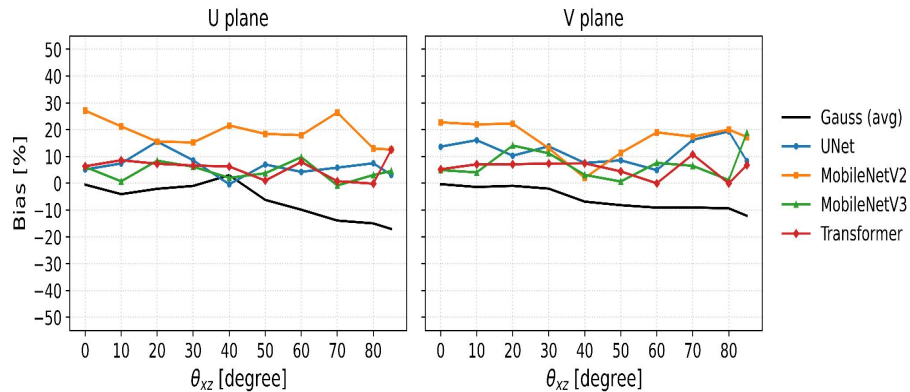


Inefficiency vs  $\theta_{xz}$  | 100MeV | DNN vs Gauss(avg)

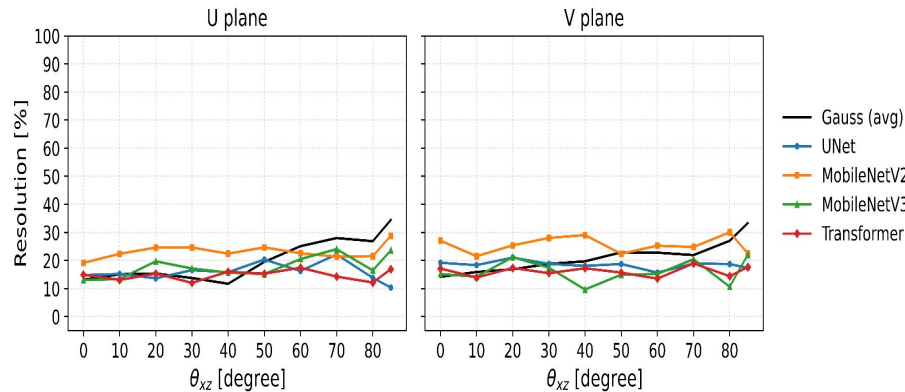


# Single Shower Evaluation (Thr 10, 100 MeV)

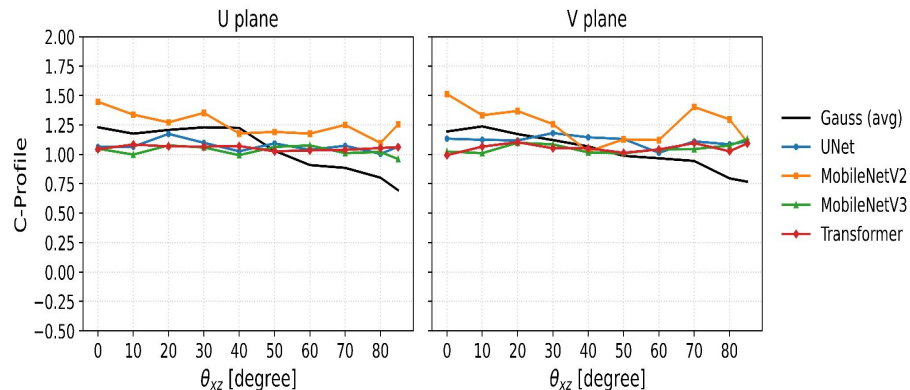
Bias vs  $\theta_{xz}$  | 100MeV | DNN vs Gauss(avg)



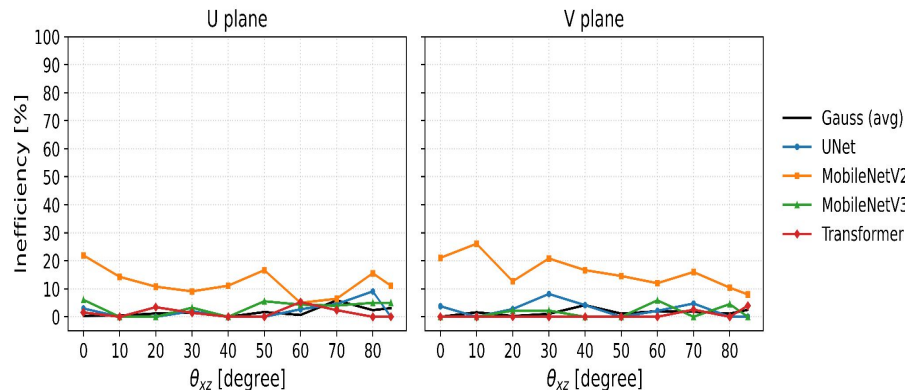
Resolution vs  $\theta_{xz}$  | 100MeV | DNN vs Gauss(avg)



Cprofile vs  $\theta_{xz}$  | 100MeV | DNN vs Gauss(avg)



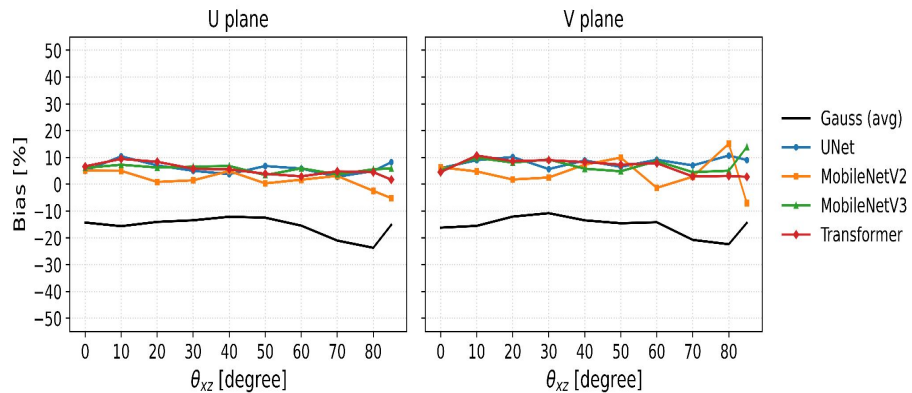
Inefficiency vs  $\theta_{xz}$  | 100MeV | DNN vs Gauss(avg)



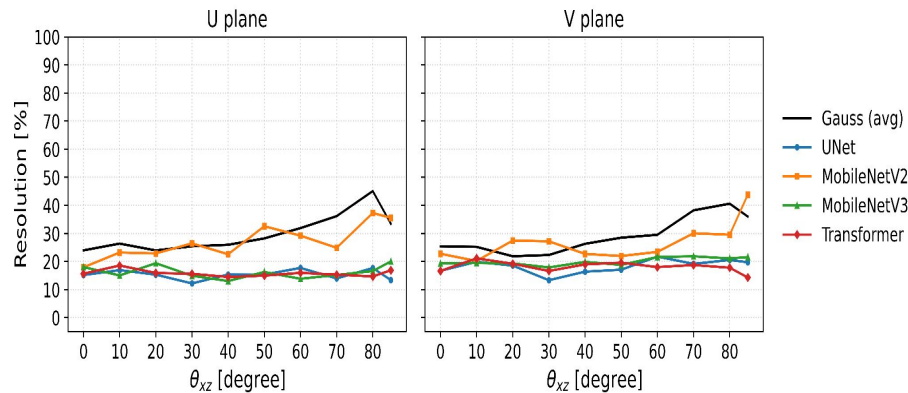


# Single Shower Evaluation (Thr 100, 1 GeV)

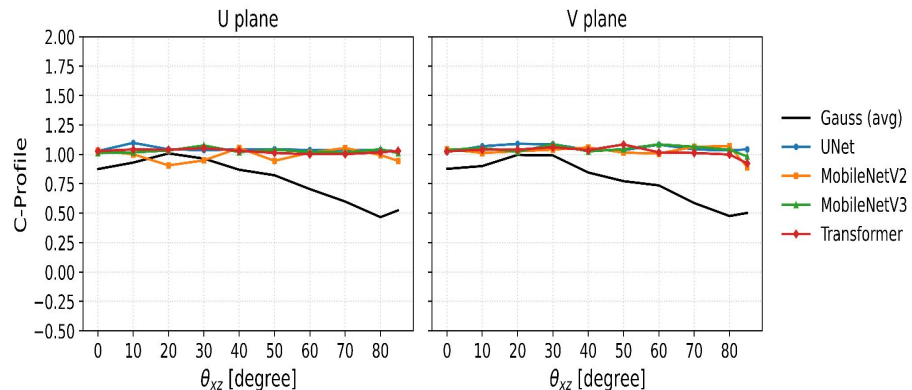
Bias vs  $\theta_{xz}$  | 1GeV | DNN vs Gauss(avg)



Resolution vs  $\theta_{xz}$  | 1GeV | DNN vs Gauss(avg)



Cprofile vs  $\theta_{xz}$  | 1GeV | DNN vs Gauss(avg)



Inefficiency vs  $\theta_{xz}$  | 1GeV | DNN vs Gauss(avg)

