



DNN-ROI Finding for ICARUS

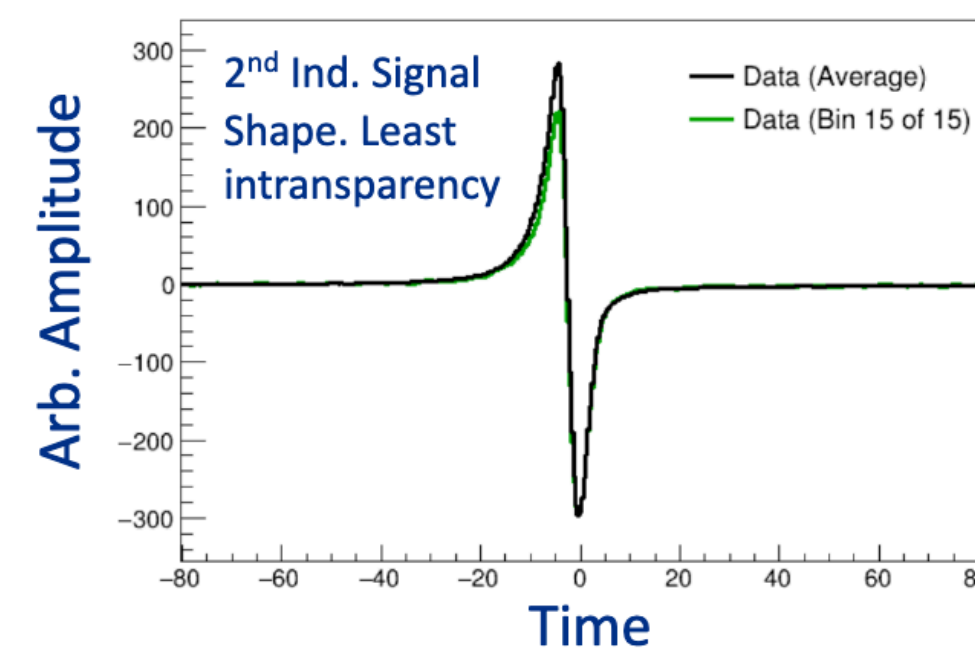
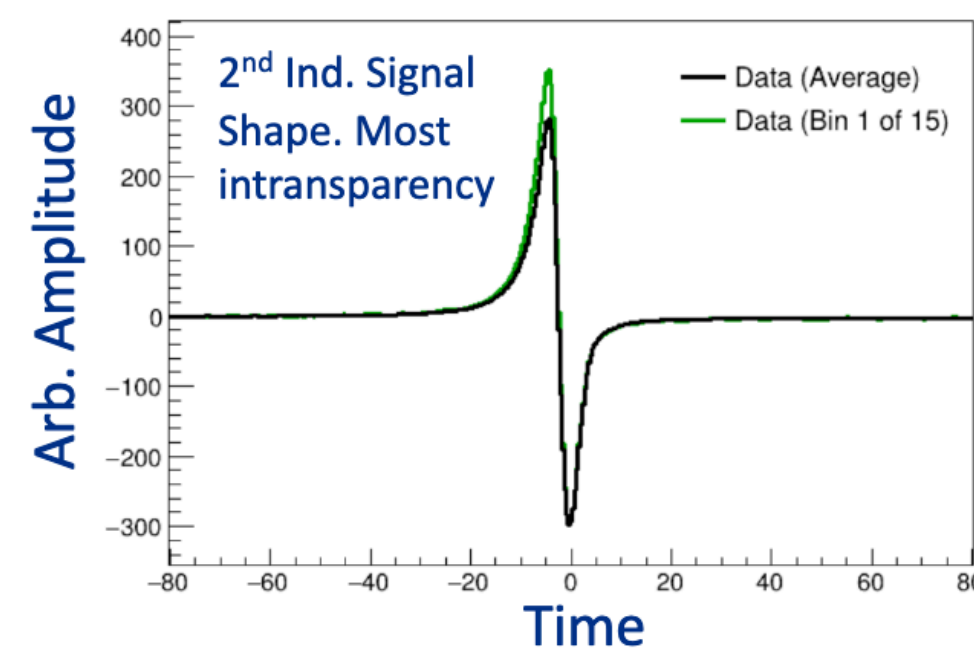
WireCell Meeting (09/19/2024)

Avinay Bhat, Gray Putnam

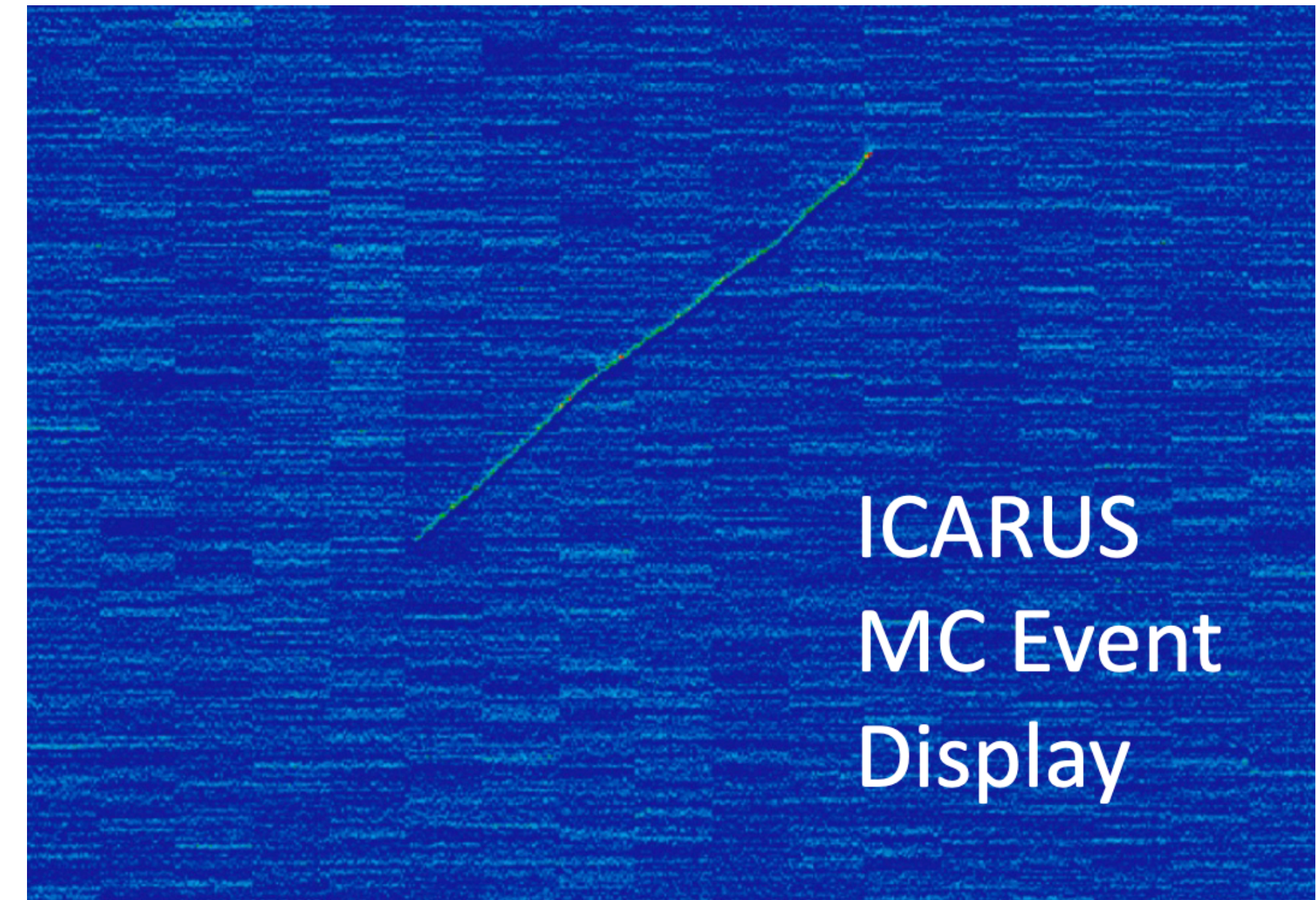
ICARUS Challenges

Signal Processing in ICARUS

- The ICARUS TPC is a challenging environment for precise charge extraction
- Challenges:
 - Warm readout electronics, long wires
 - Significant coherent noise on each readout board (64 channels)
 - Smoothly varying signal shapes across detector (middle induction intransparency)



SBN DocDB
32136

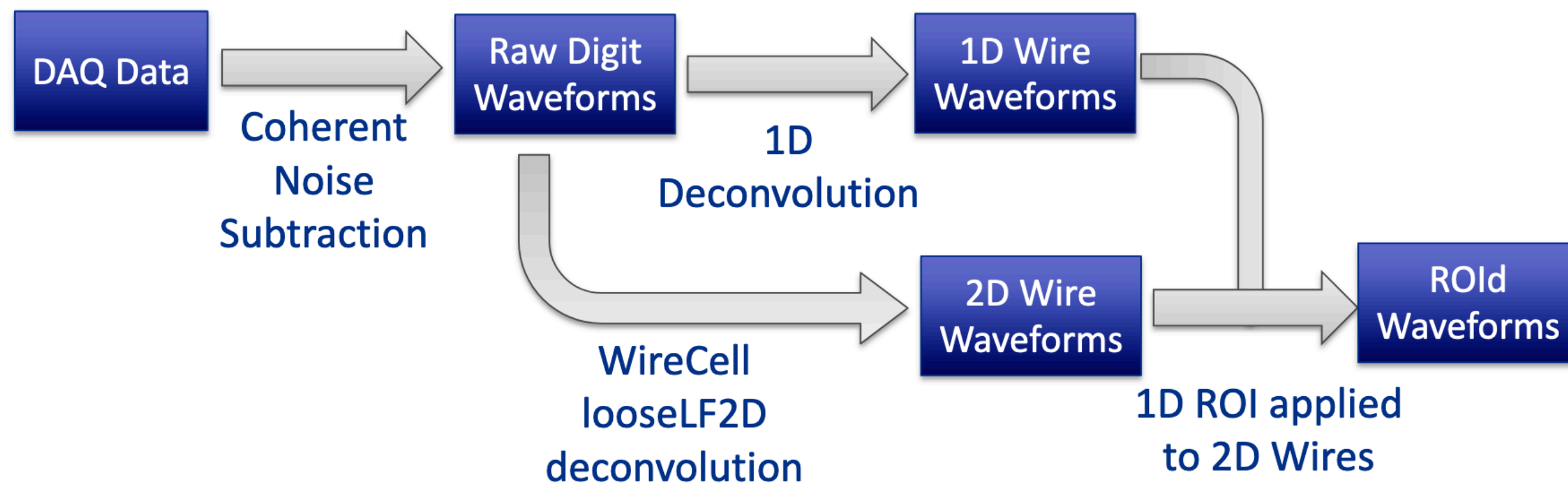


Presentation by Gray Putnam on 7/25/24

ICARUS Challenges

Alternative WireCell Implementation in ICARUS

- Sergey and Tracy also put together a partial integration of WireCell into ICARUS simulation (1D ROI + 2D charge extraction)



- Drawbacks: no 2D deconvolution for ROI finding, looseLF filter (Wiener-style filter w/ low-frequency cutoff) is not suitable for charge extraction



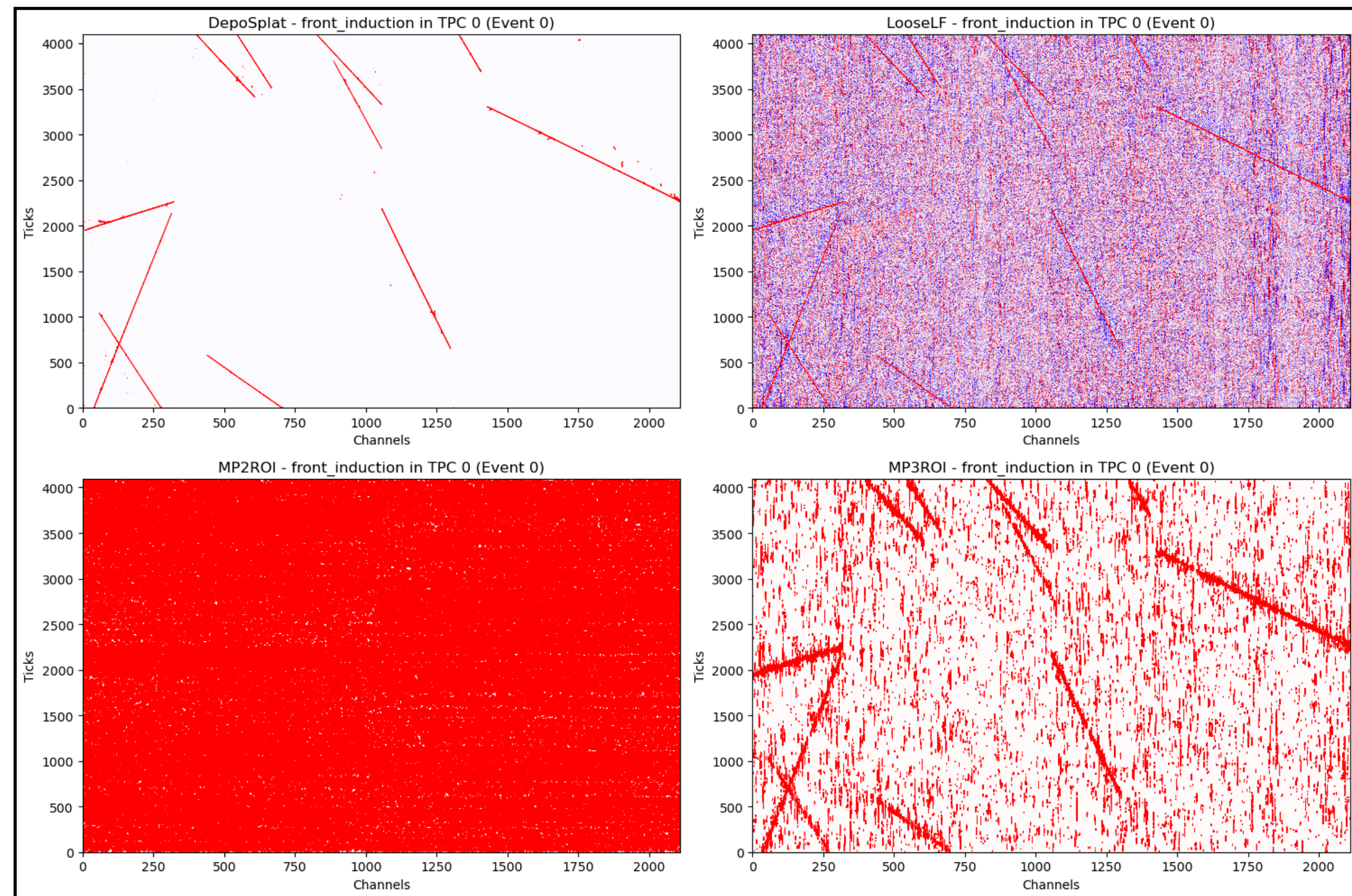
What I want to do

- I want to get 2D deconvolution working in ICARUS to improve charge extraction performance
- To do this, I plan to:
 - Optimize the filters in the standard 2D deconvolution ICARUS
 - Implement DNN-ROIs for ICARUS (with Avinay+Moon) ✓

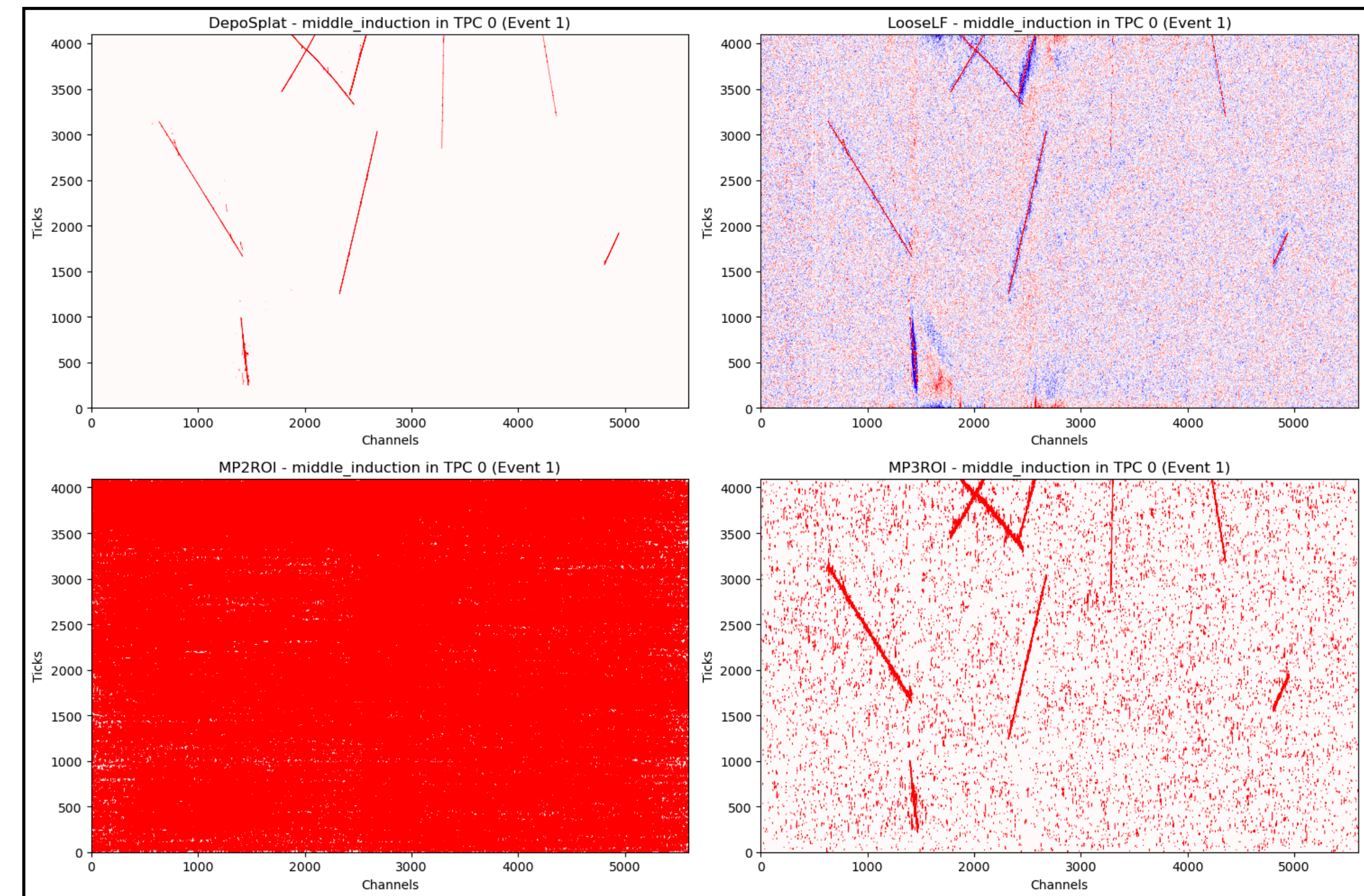
DNN ROI Training Samples

- Gray produced a sample of ~1000 BNB + Cosmics Events for ICARUS.
- ICARUS has 4 TPC readouts with 2 induction planes each.
- The induction planes themselves have different number of channels
 - 2112 for Front Induction Plane (say U Plane)
 - 5600 for Middle Induction Plane (say V Plane)
- Started with a strategy of trying to train different models, one for each induction plane.
- Used a rebin factor of 8. (Can increase them number in later iterations)

DNN ROI Training Samples

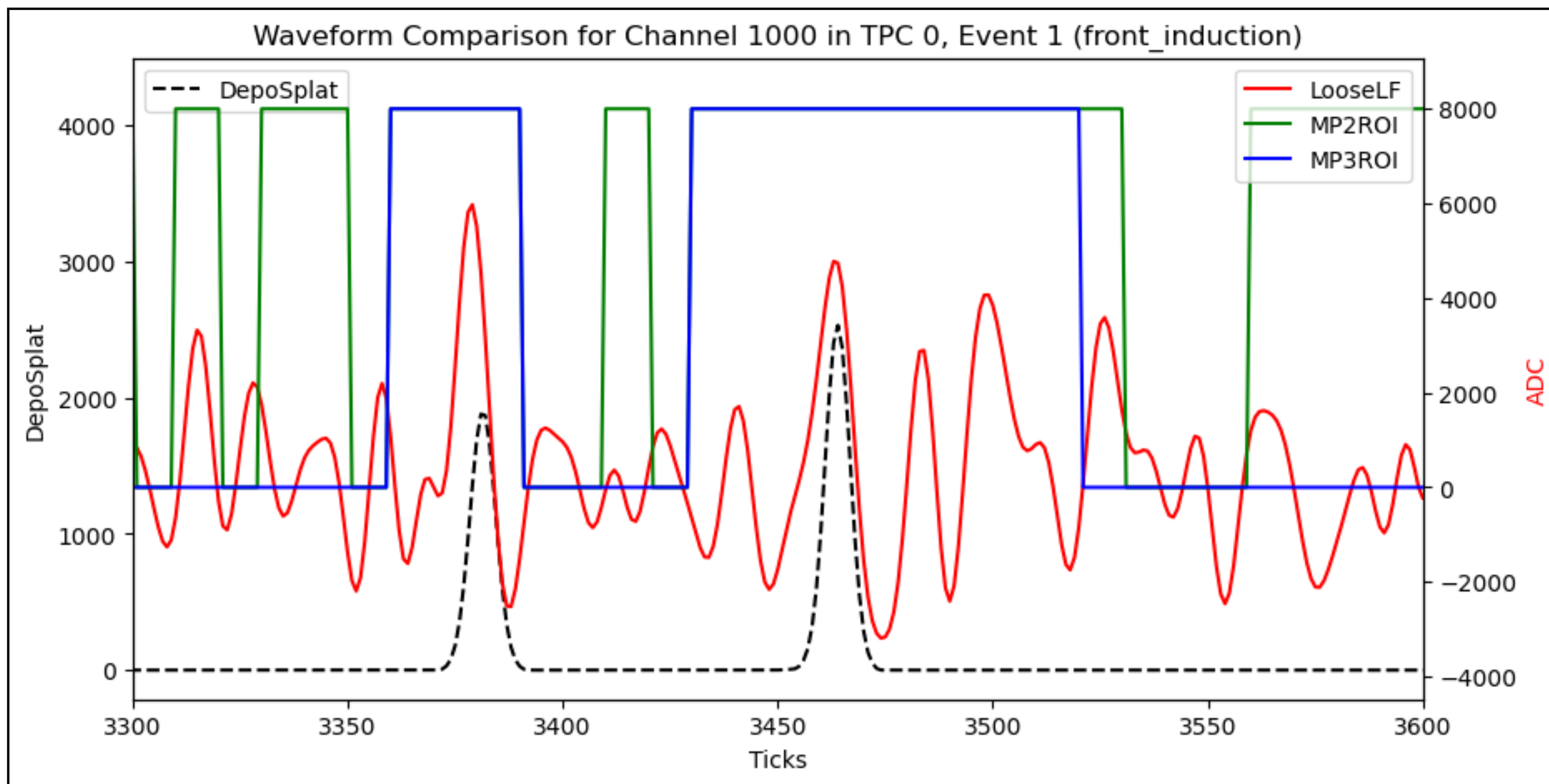


U Plane

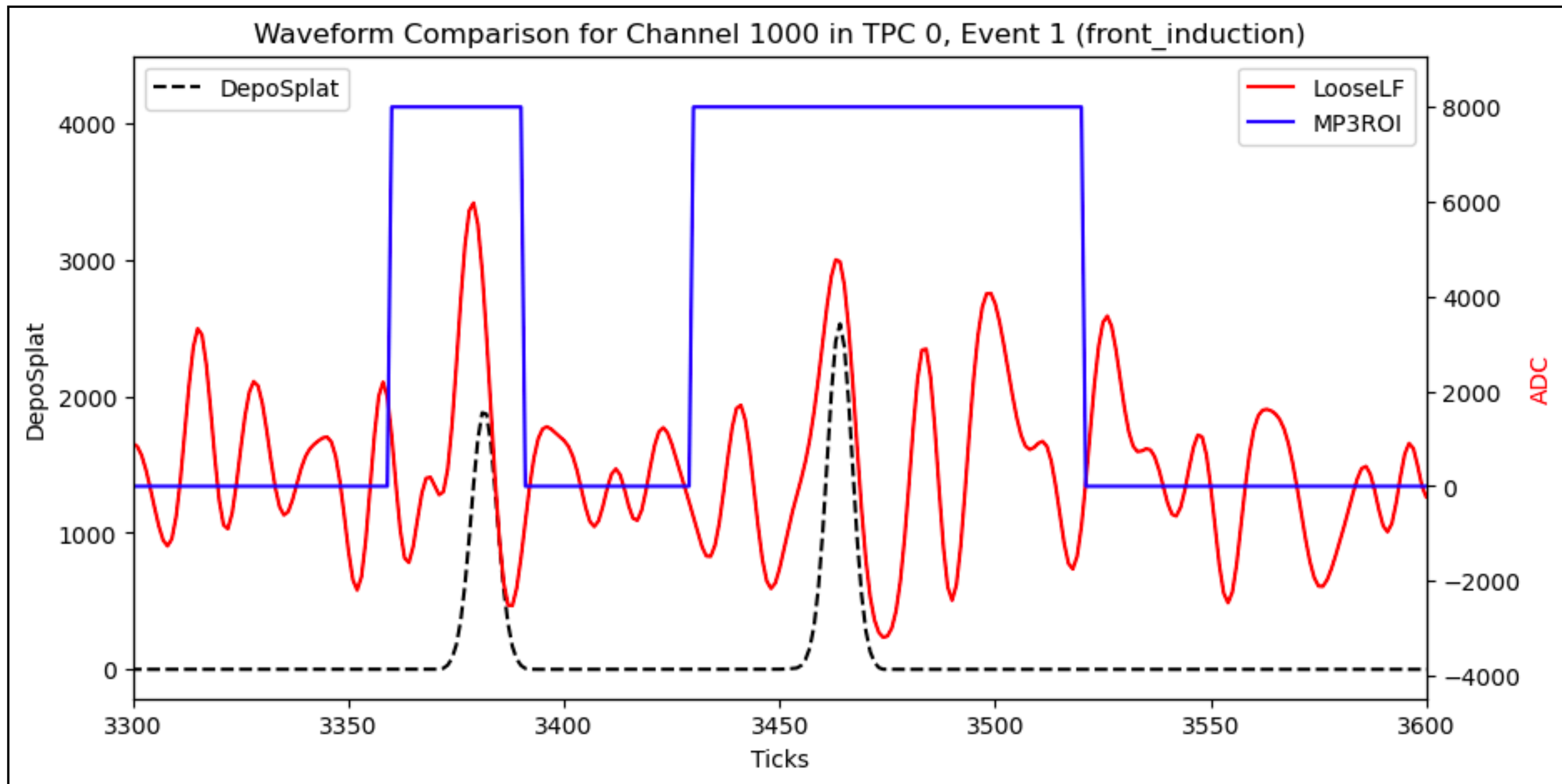


V Plane

DNN ROI Training Samples

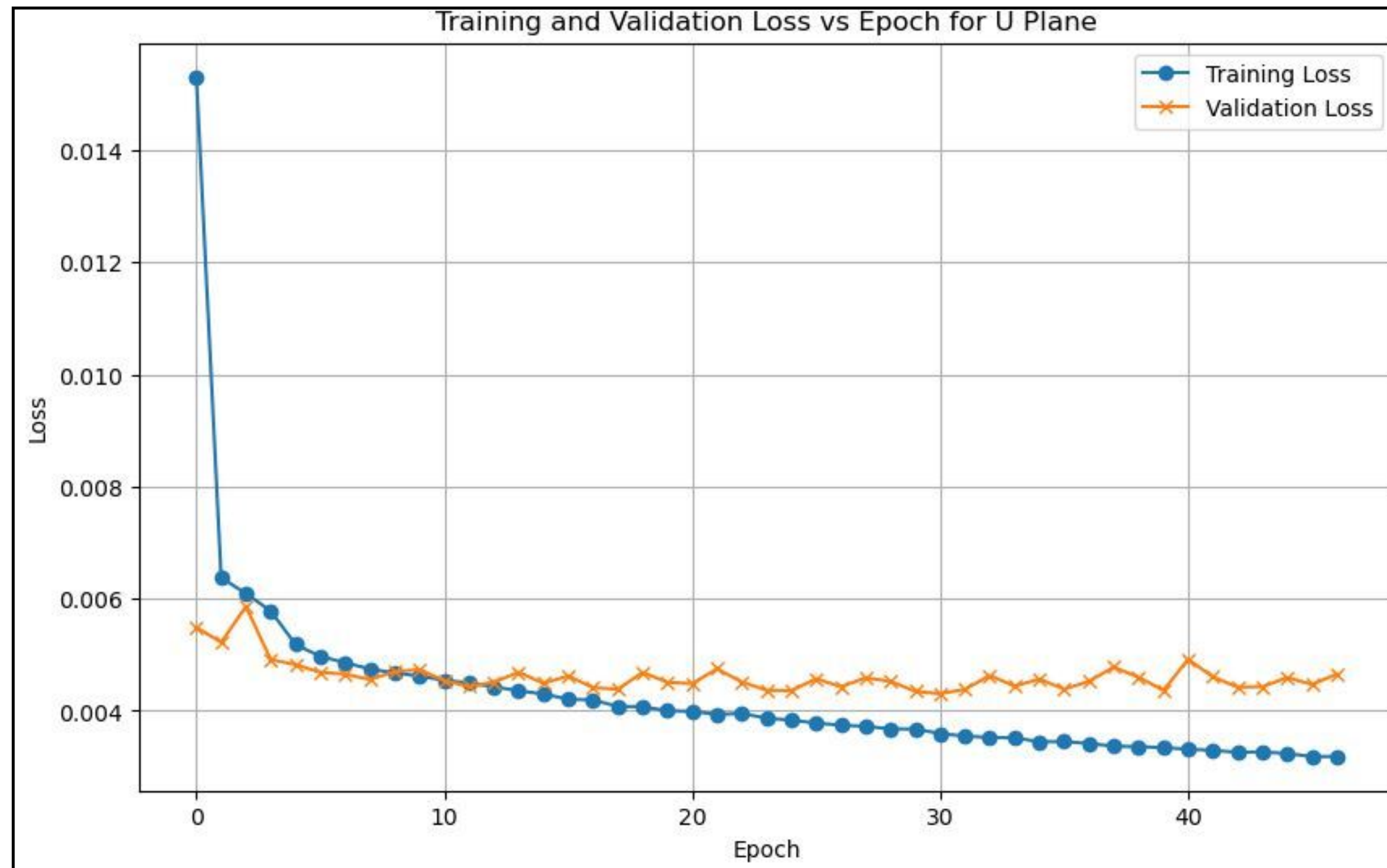


DNN ROI Training Samples



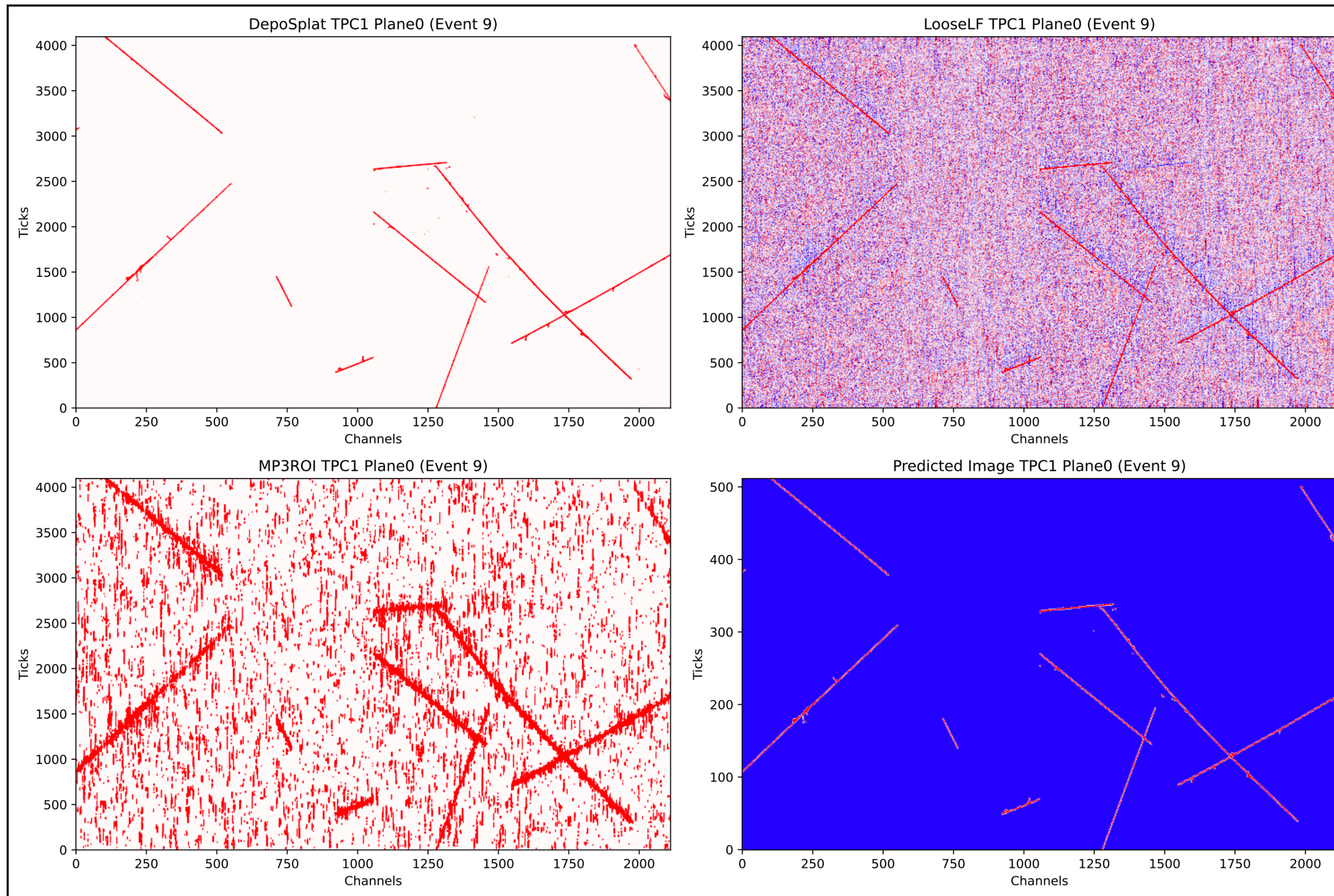
Decided to drop MP2
and only use Loose LF and MP3

DNN Training

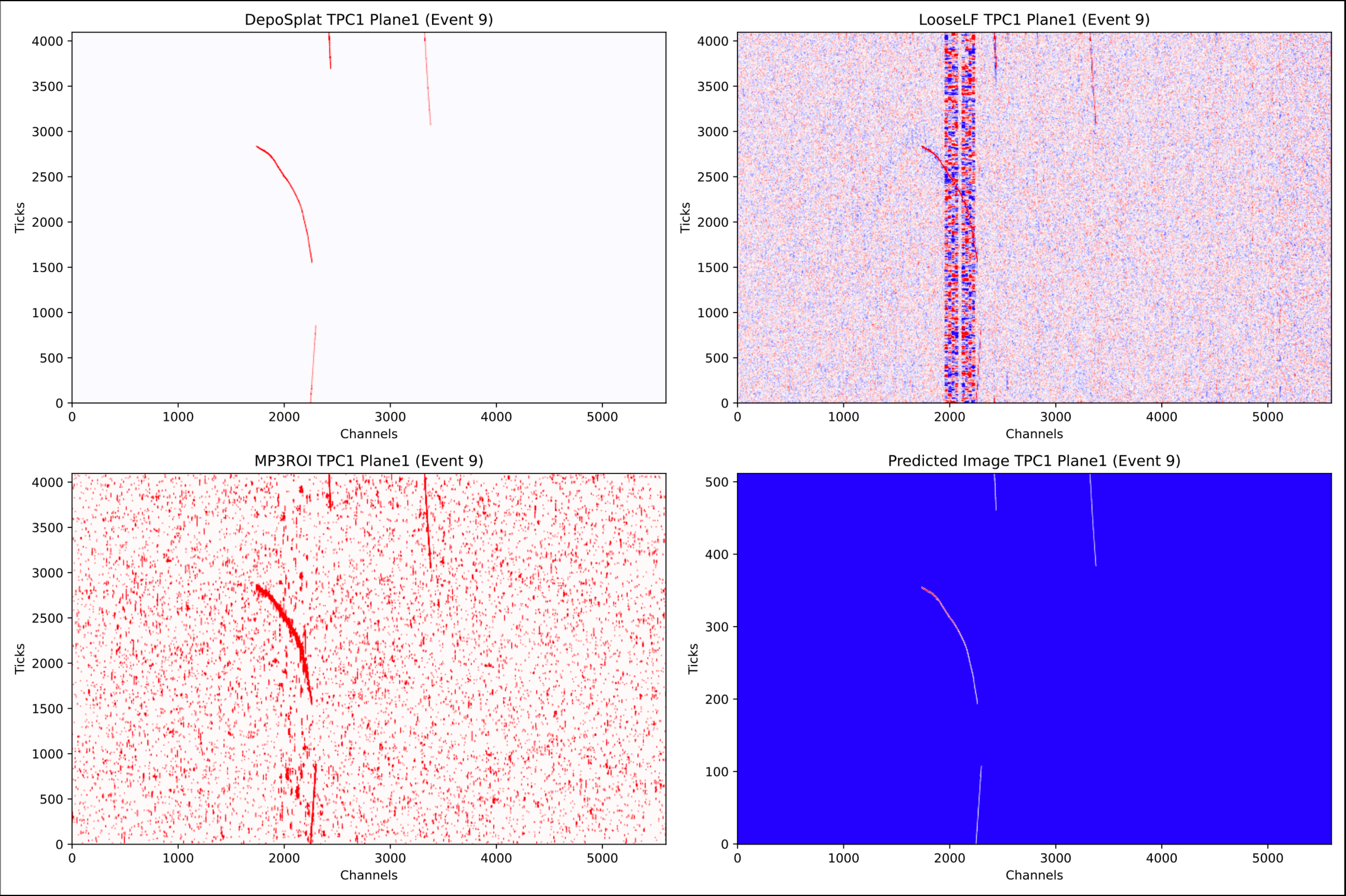


- Trained two different models using UNet architecture, Adam Optimizer with learning rate of 0.001
- Dataset consists of ~1000 BNB + Cosmic events
- Validation loss curve drops for the first 10 epochs and then saturates.
- Saved the least loss and highest dice score models for prediction for both planes.
- Took about 50 minutes per epoch for U Plane and about 2 hours for V Plane

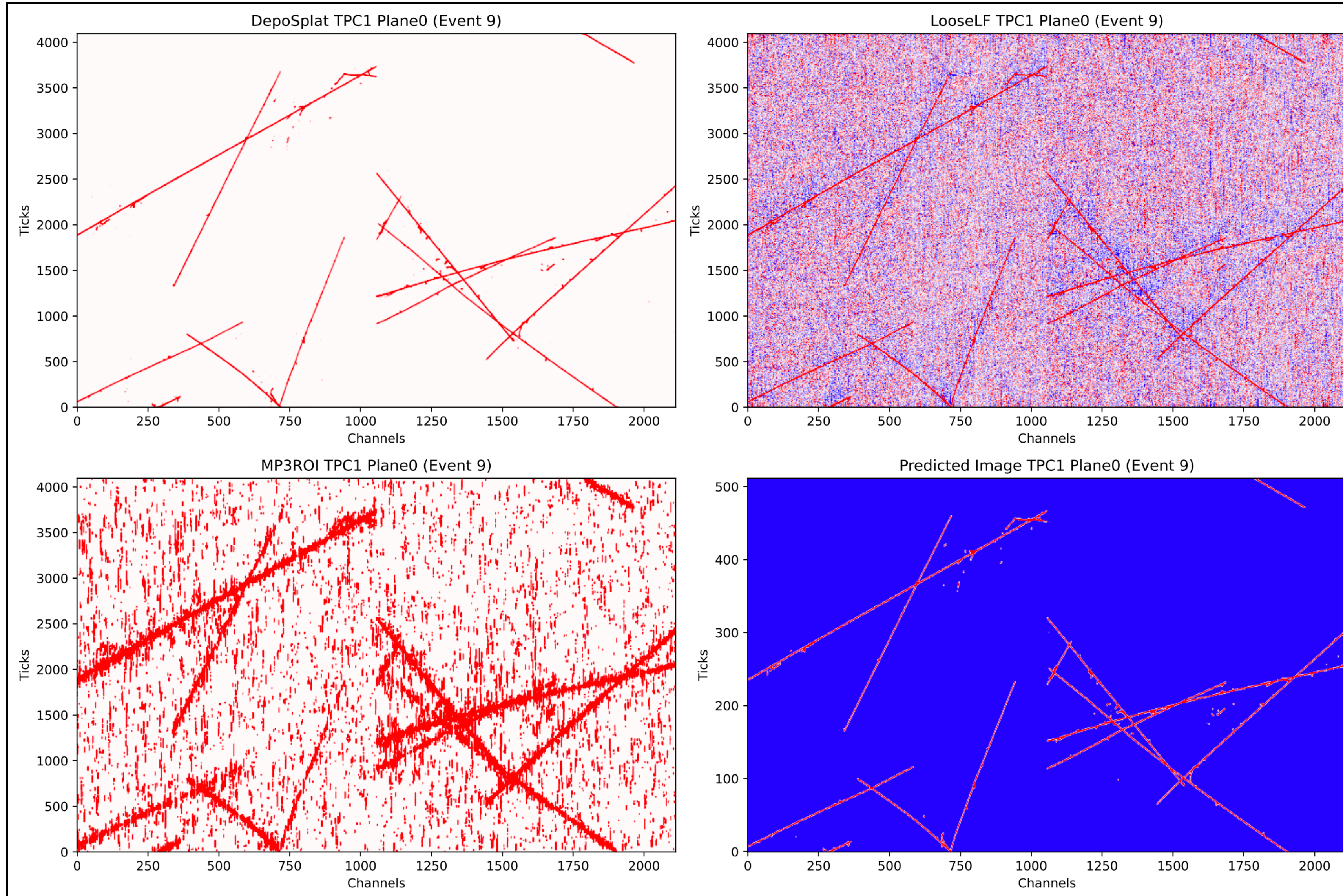
DNN ROI Prediction



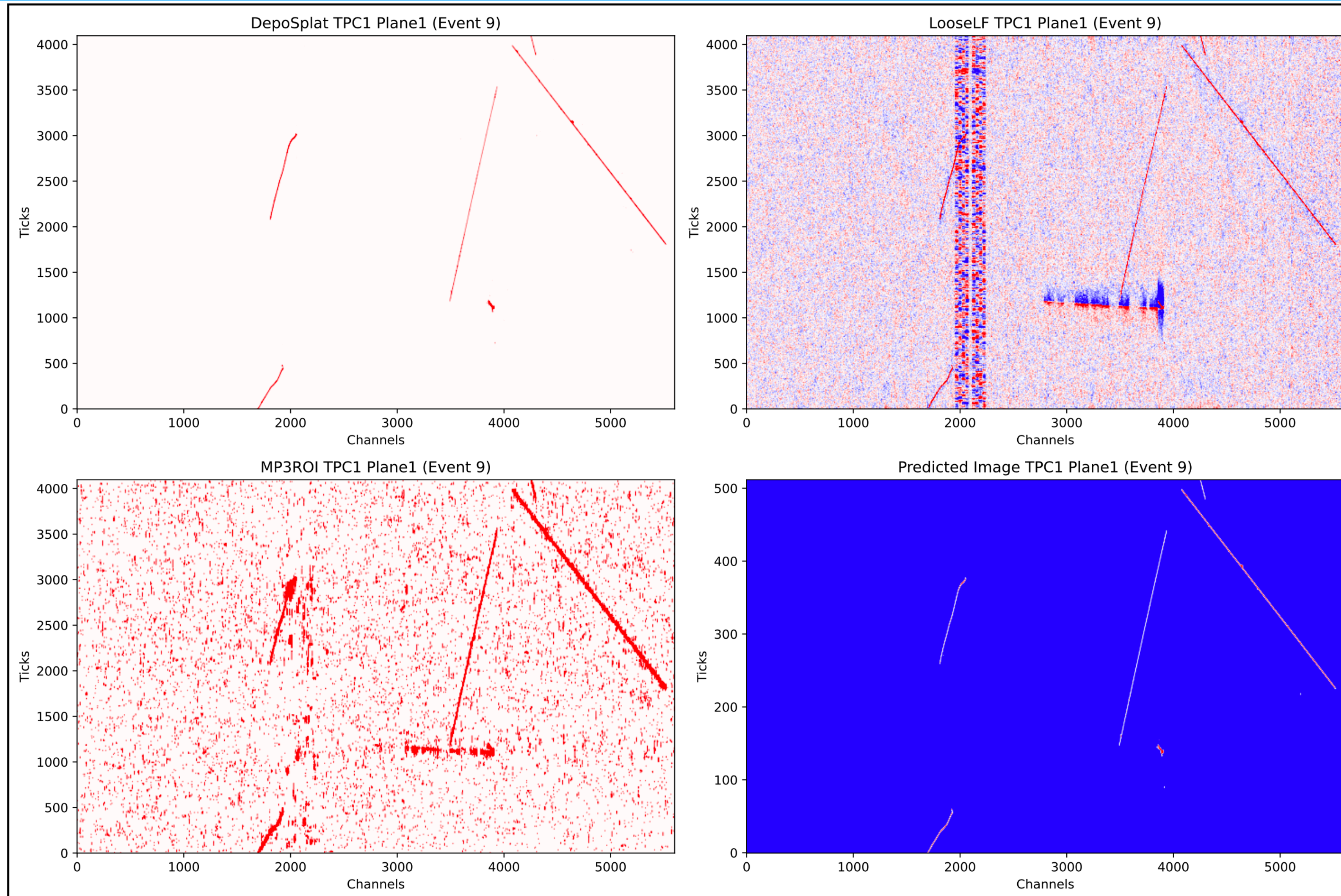
DNN ROI Prediction



DNN ROI Prediction



DNN ROI Prediction



Summary and Future Plans

- First pass at DNN ROI Prediction on ICARUS looks very promising.
- Need to further optimize the training process and pre-processing of samples for training.
- Gray has been working on producing new samples with optimized filter parameters as well as smear values for filters.
- Would like to train the model with samples containing varying noise conditions within ICARUS.
- Want to estimate the ROI pixel efficiency and purity.
- Would like to start working on inference to get a sense of how charge extraction looks like



DNN-ROI Finding for SBND

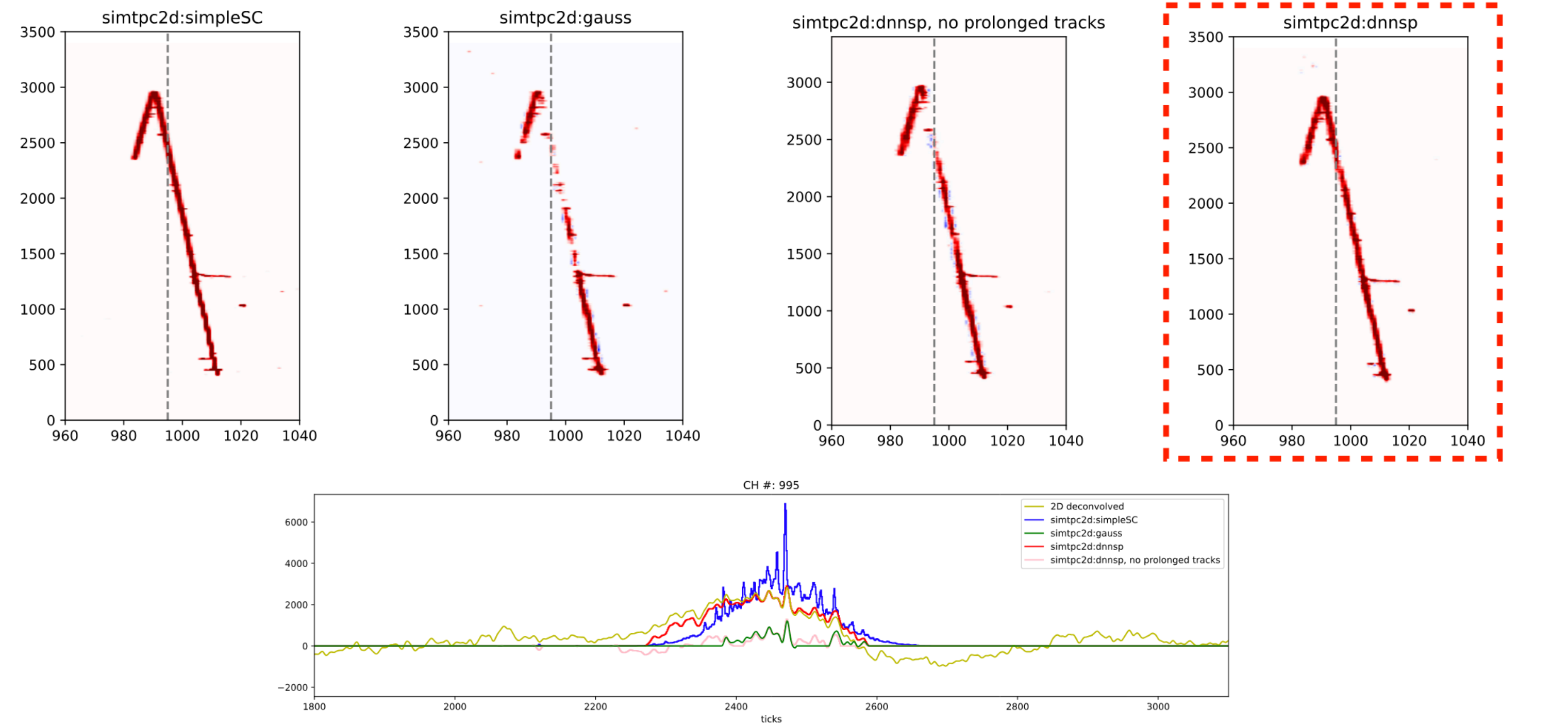
WireCell Meeting (09/19/2024)

Avinay Bhat

SBND DNN ROI Performance on PTs

DNN ROI @ SBND SP - MC

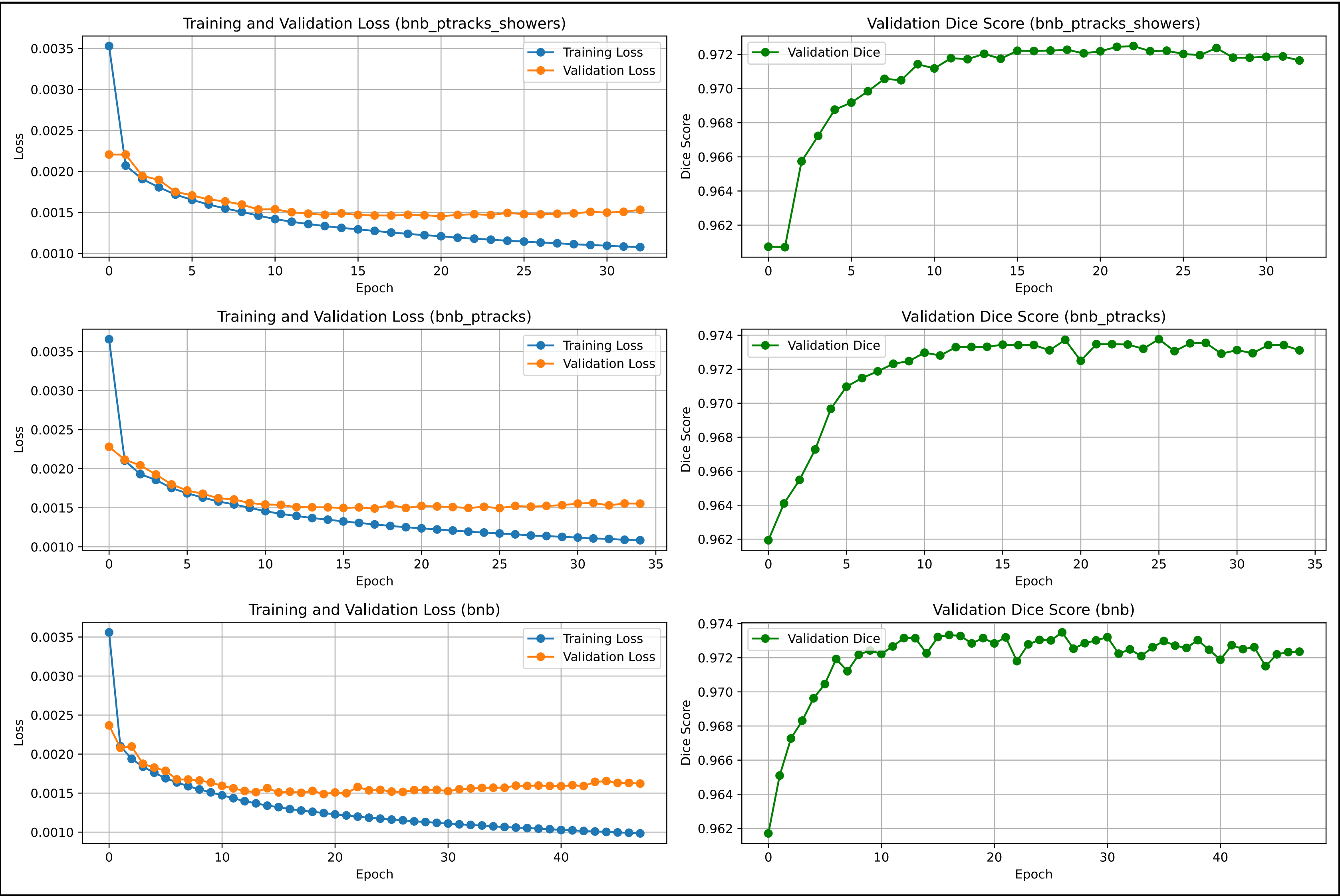
- UResNet trained for SBND



Performance on Showers

- Trained three different models using UResNet architecture
- Used Adam optimizer to optimized the learning rate (started with $lr=0.001$)
- Used BCE to calculate the loss
- Trained and evaluated on Augmented sample set of:
 - BNB Cosmics (5000 events)
 - Prolonged Tracks (θ_{xz} [75-87]) (800 events)
 - BNB Electron Showers (400 events)
- Showers were generated with binned energies between [0.15-1.5 GeV] with bin size of 0.3 GeV
- Tested on a test sample of showers

Performance on Showers



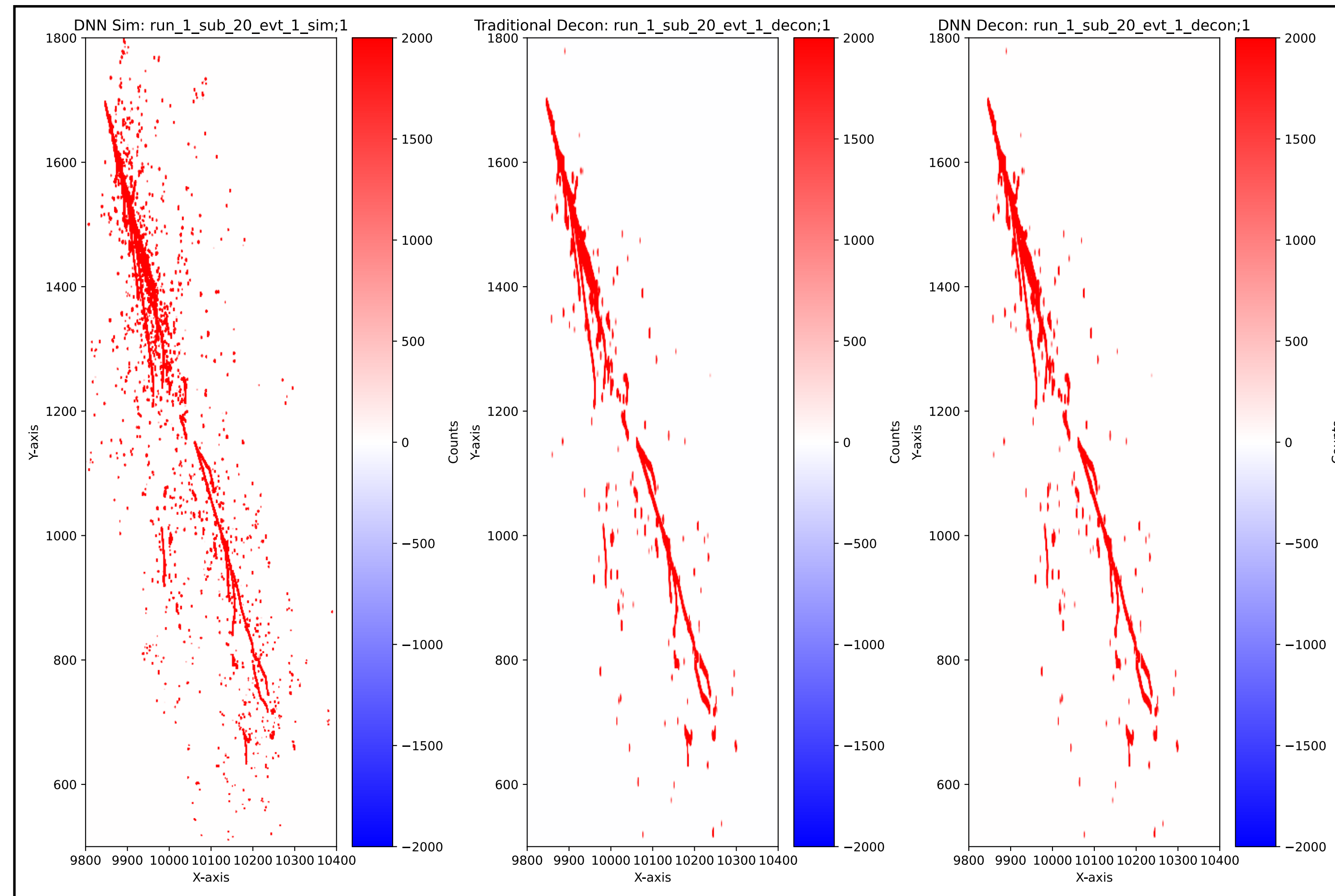
BNB + Cosmics + PT+
Showers

BNB + Cosmics + PT

Only BNB + Cosmics

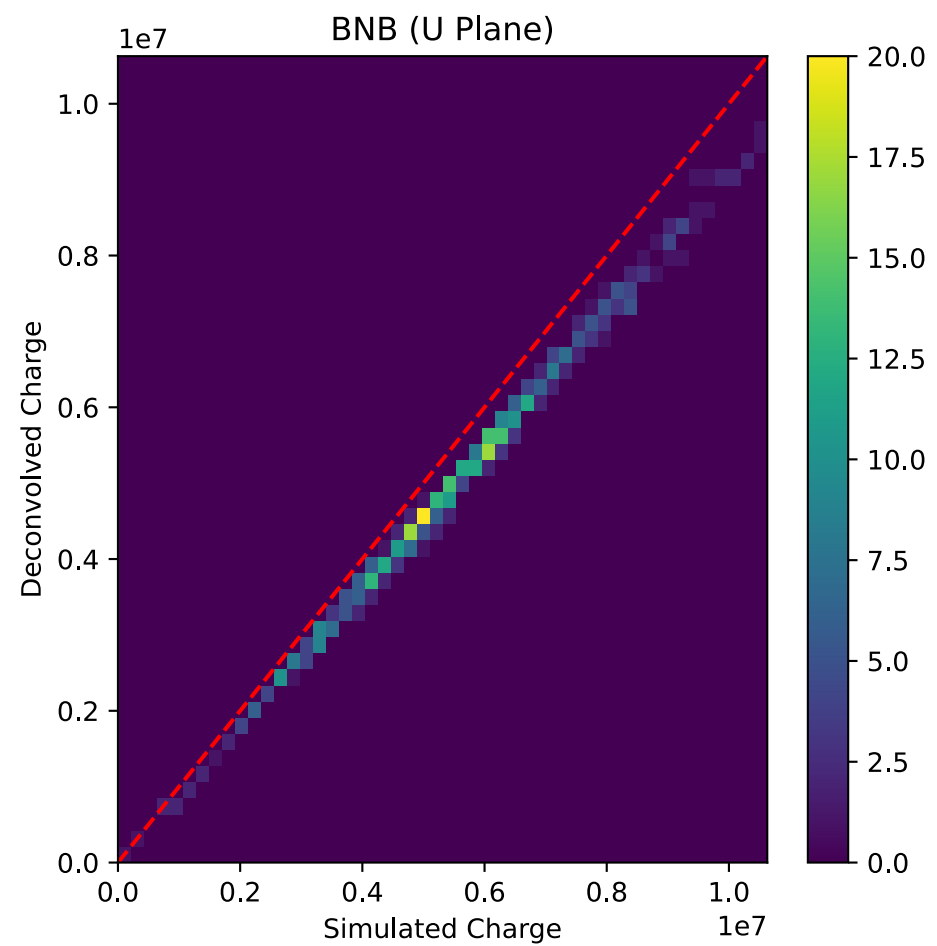
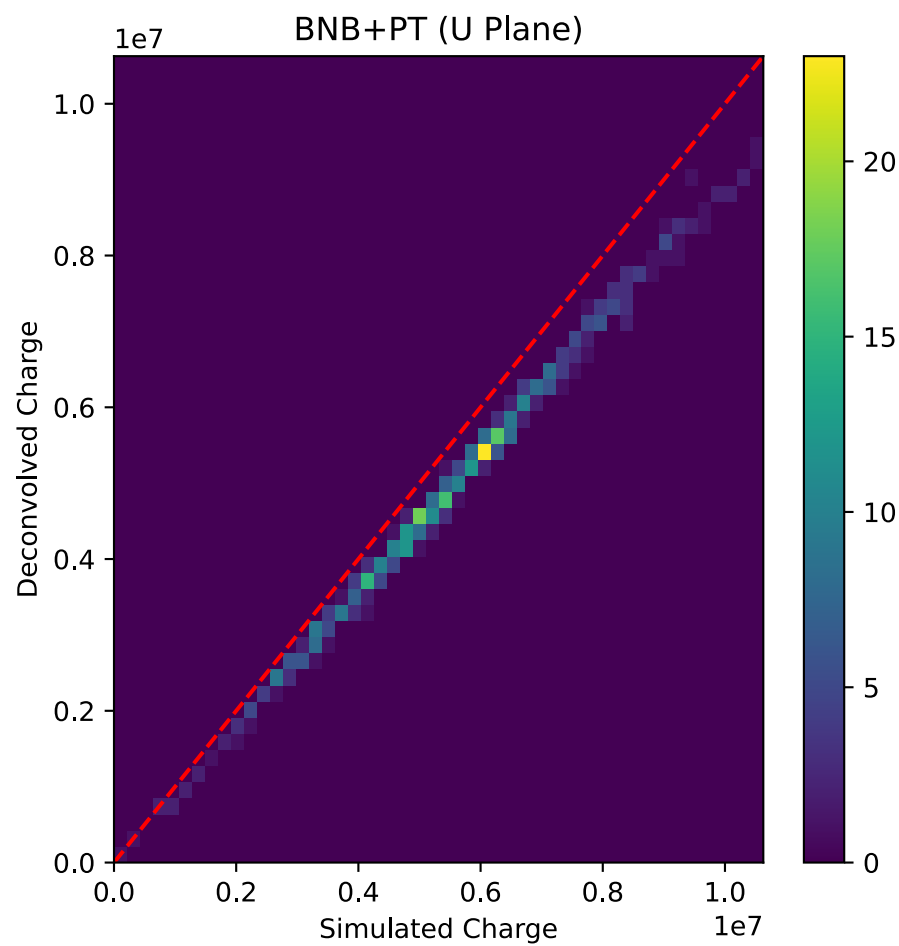
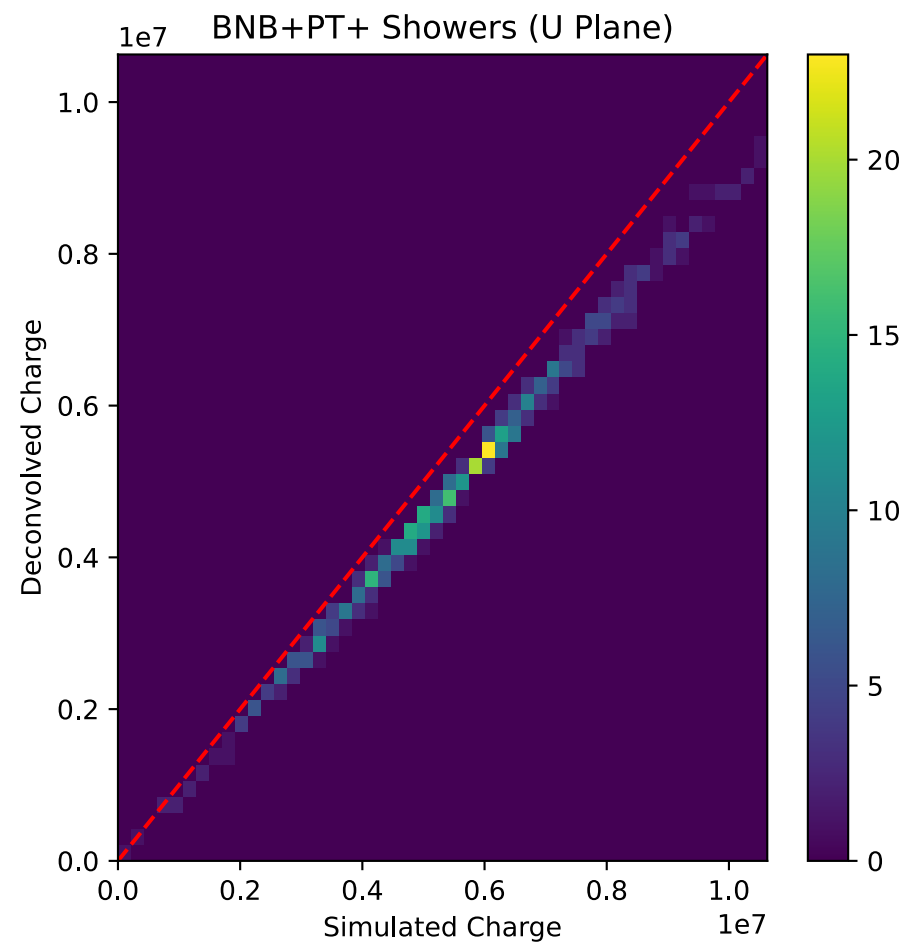
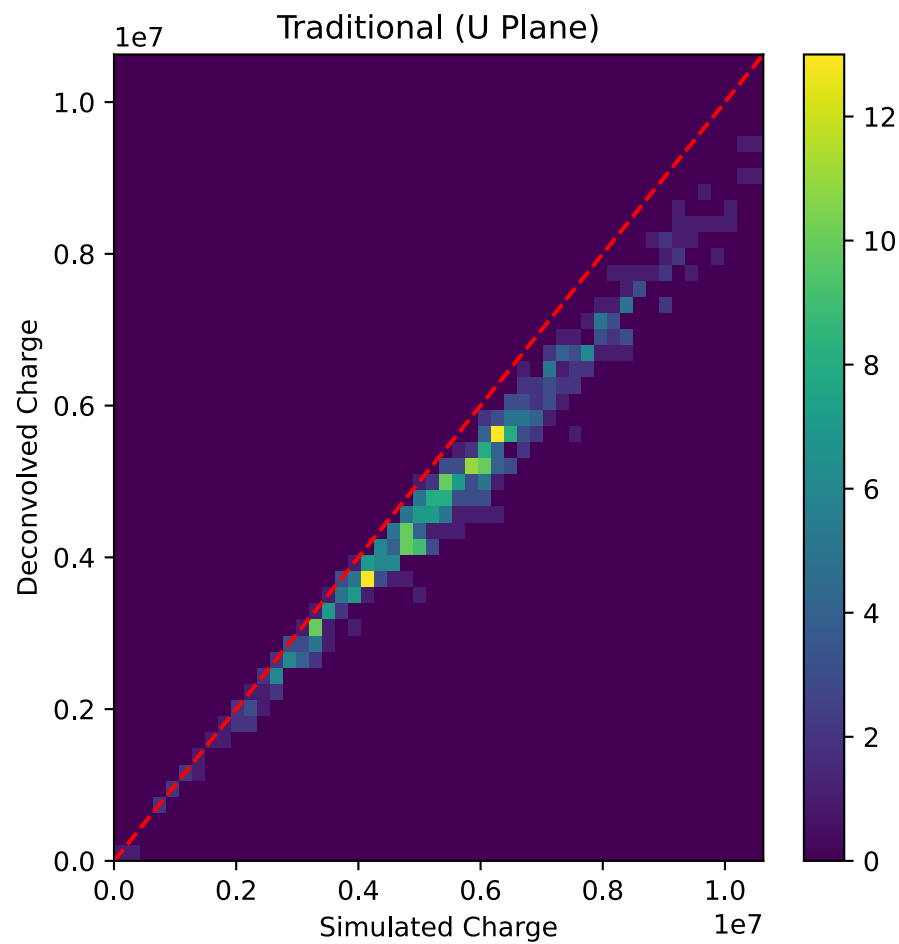
Performance on Showers

Example for
E [1.4 GeV-
1.5 Gev]

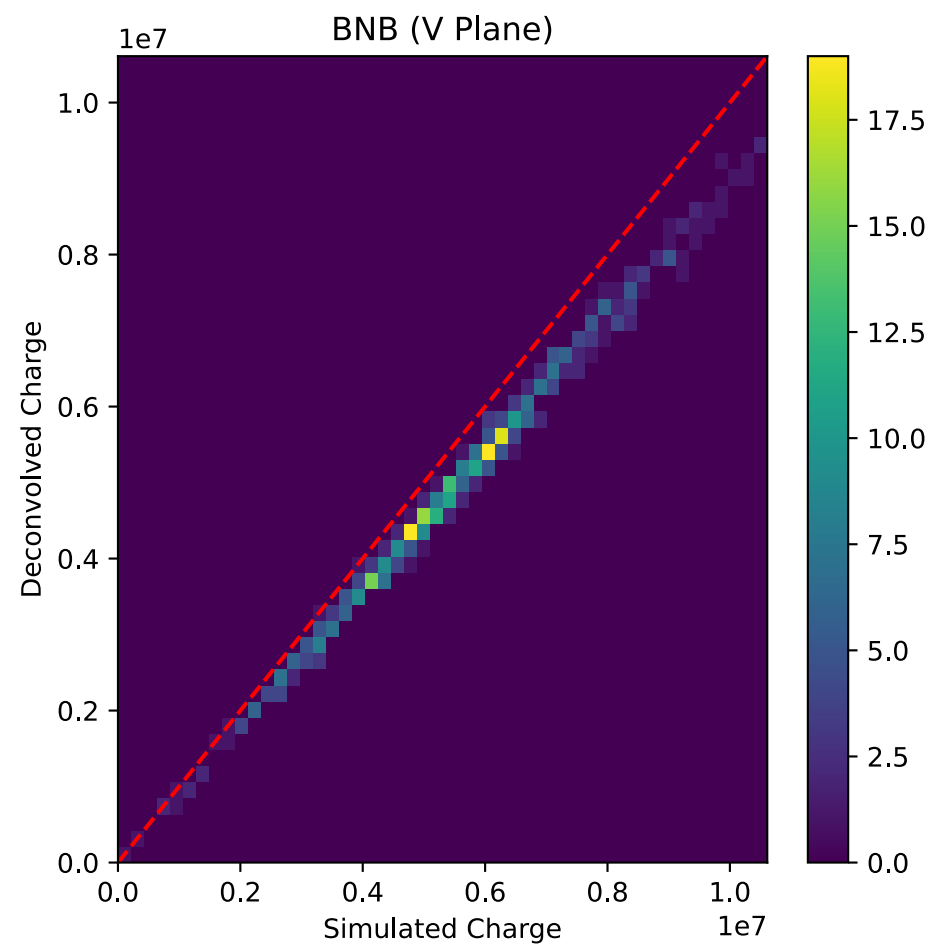
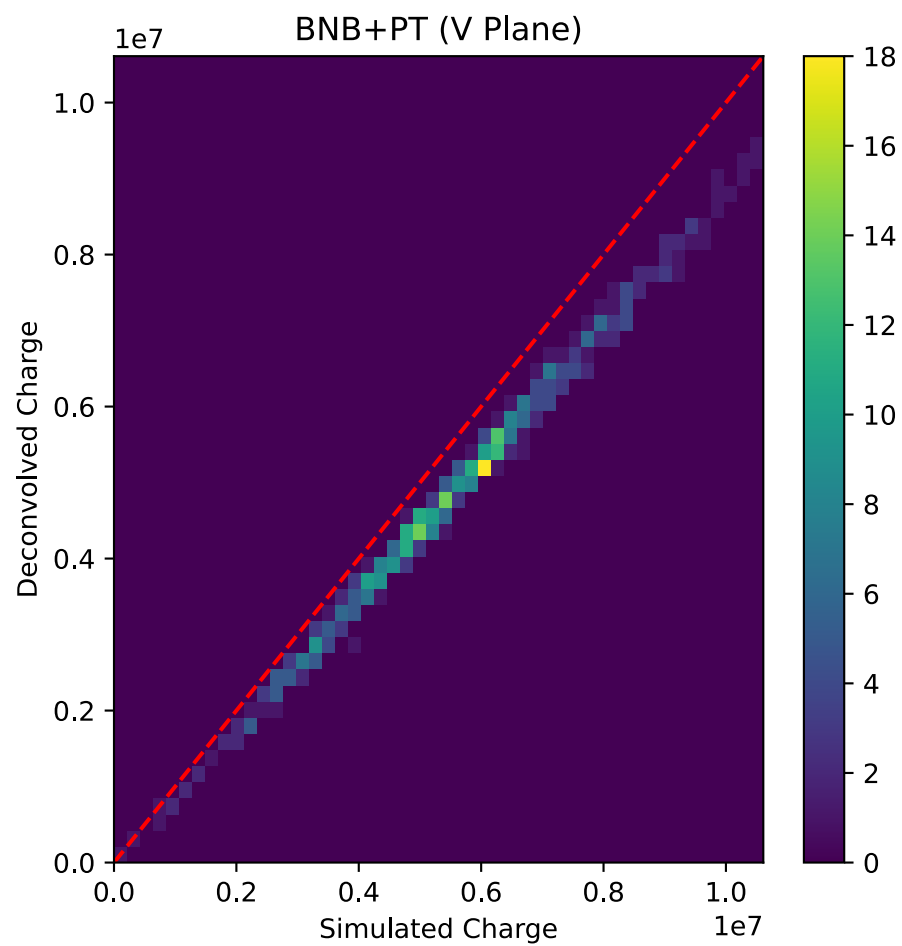
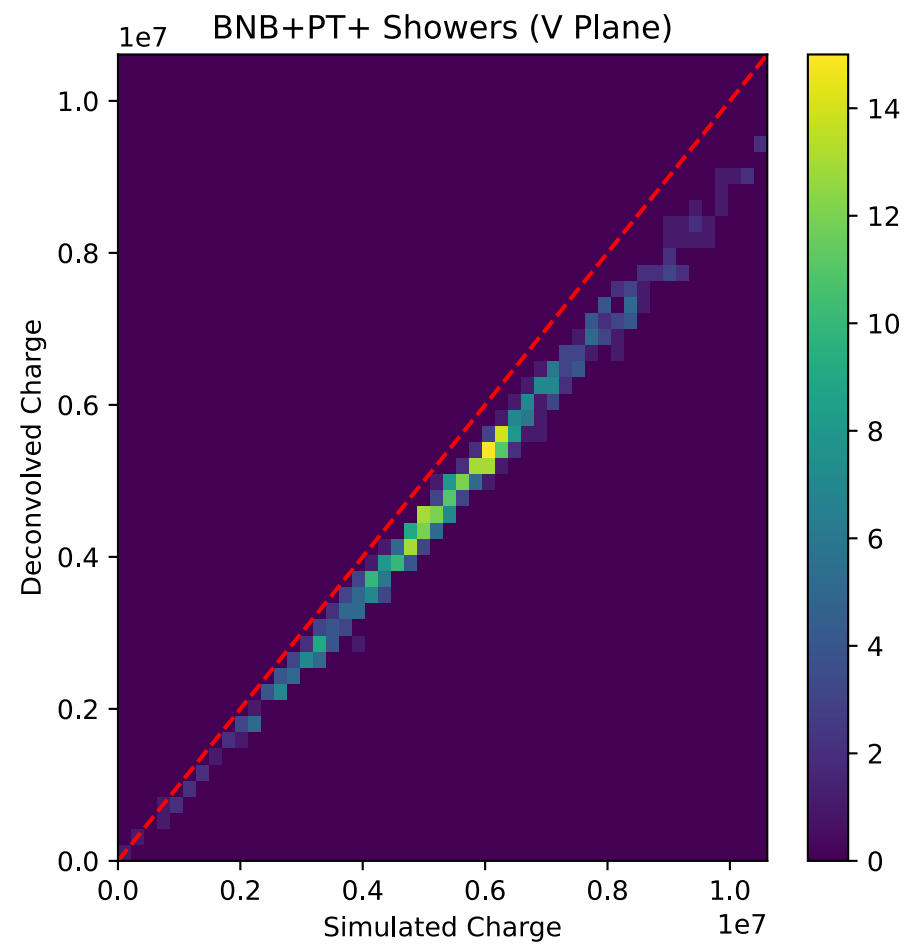
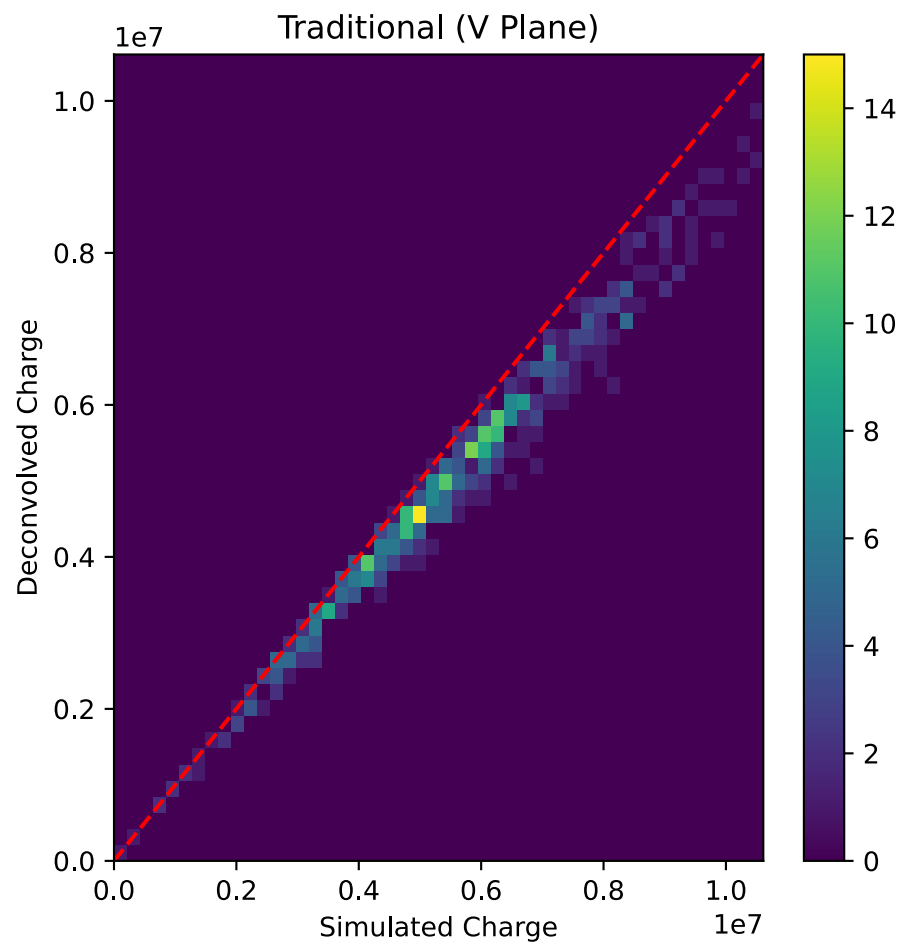


Charge Extraction on Showers

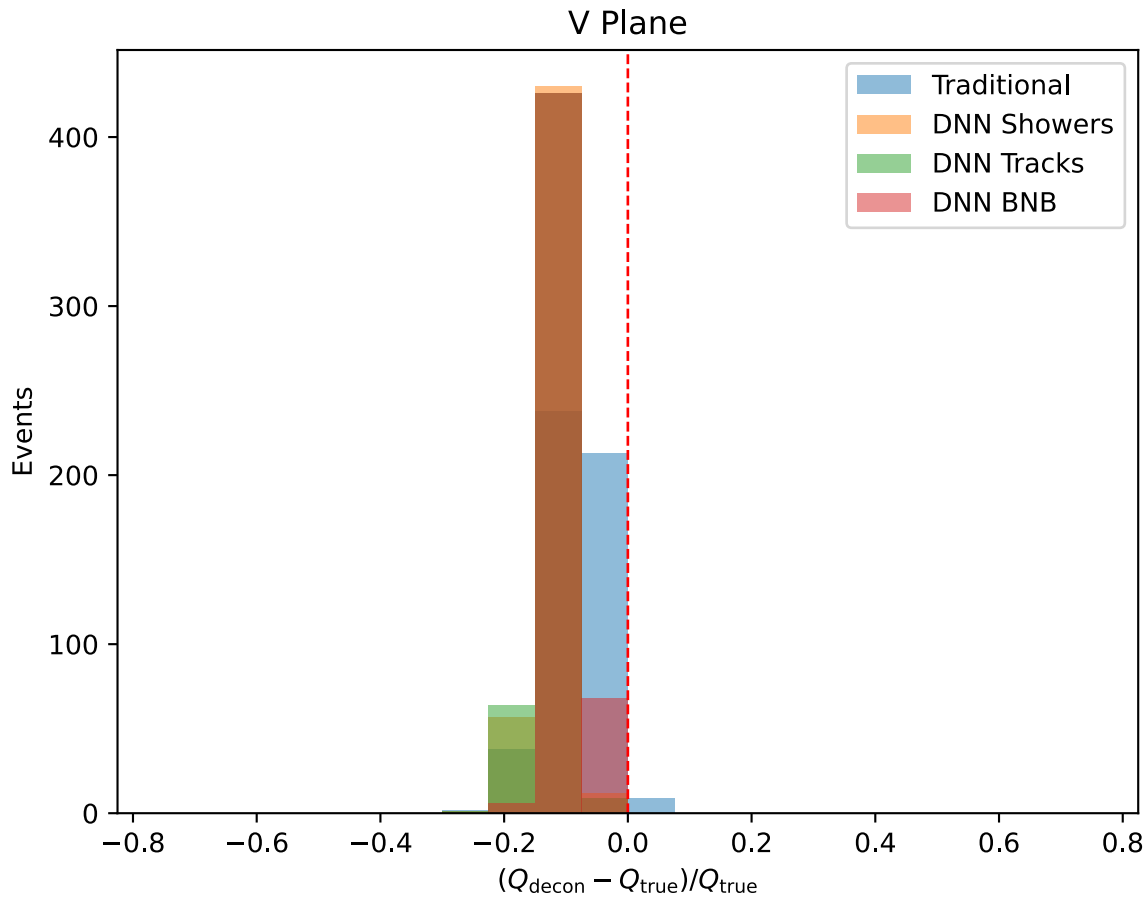
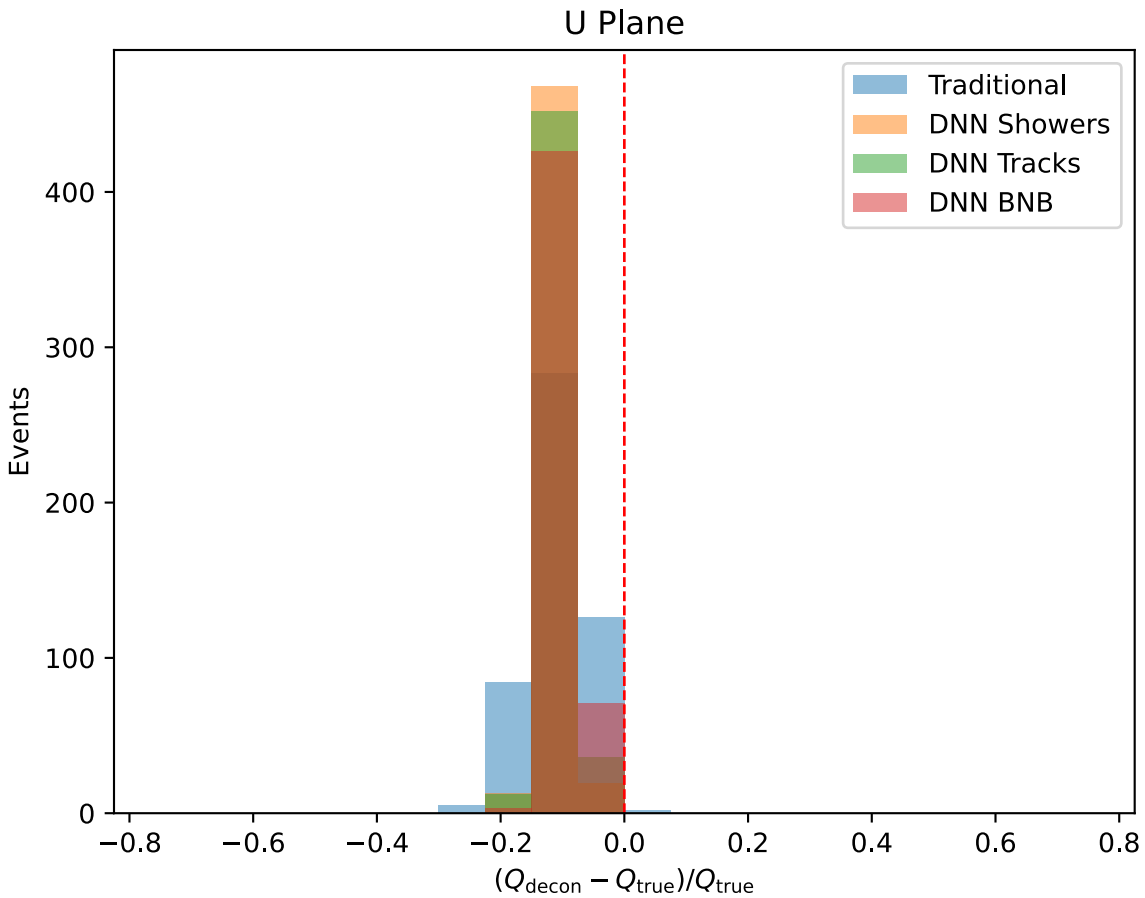
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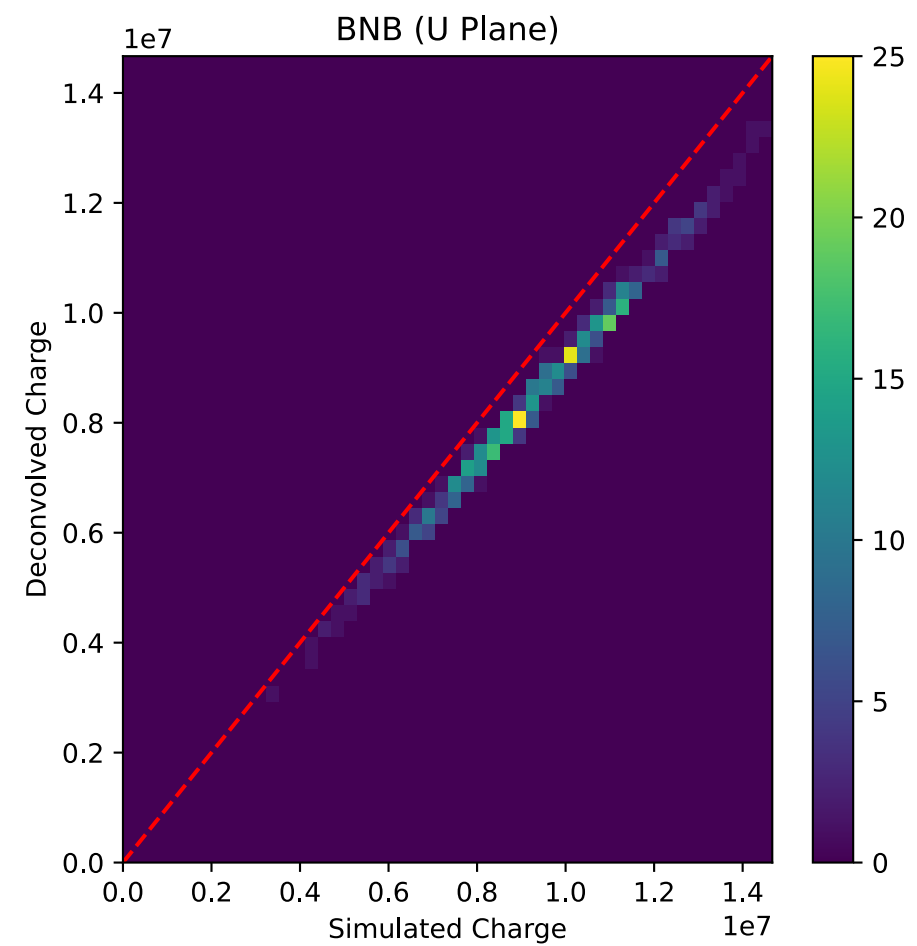
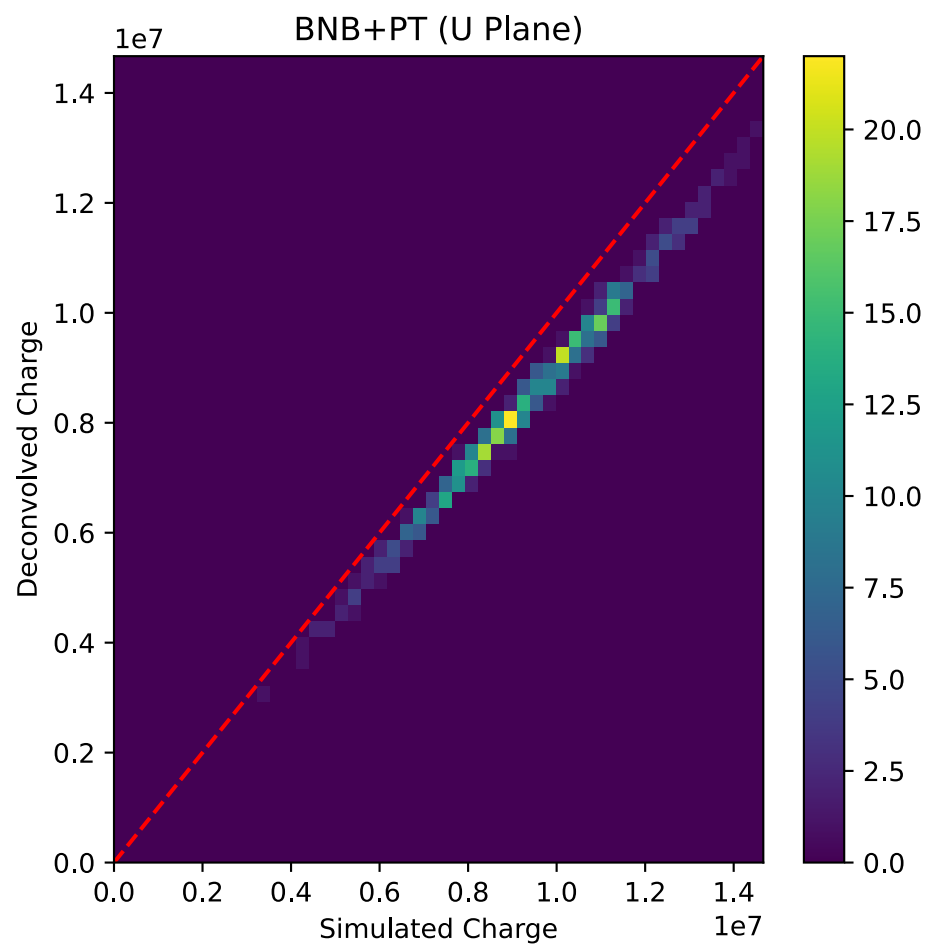
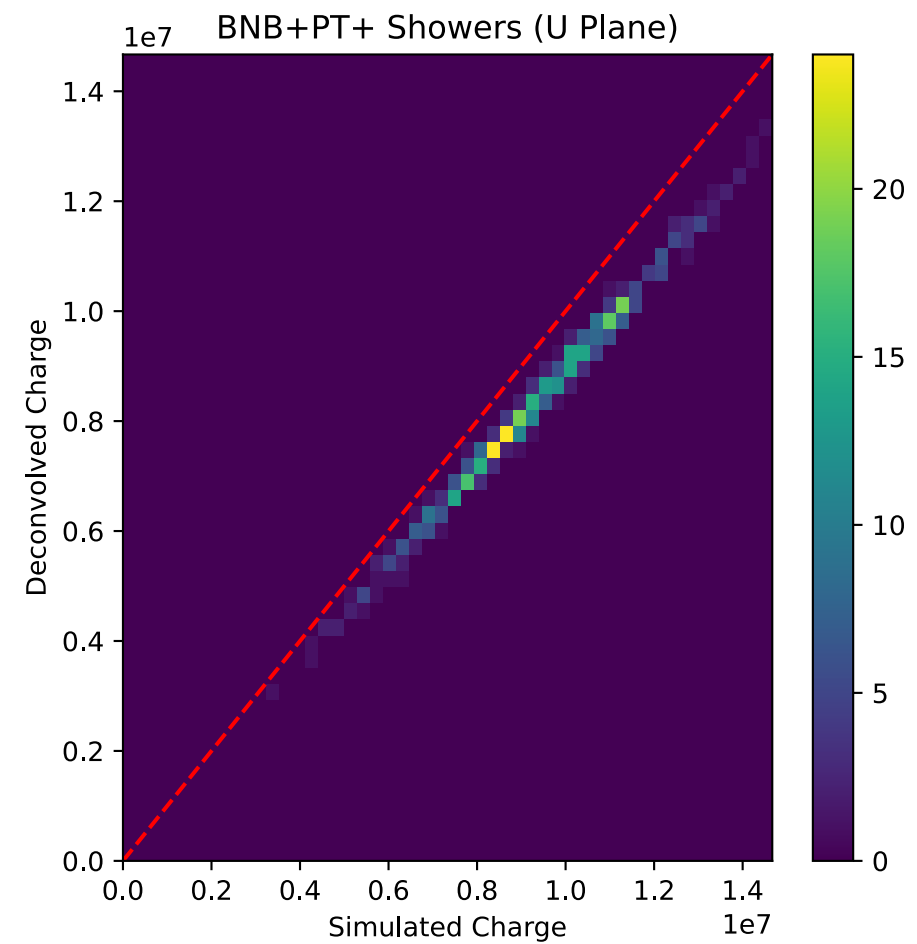
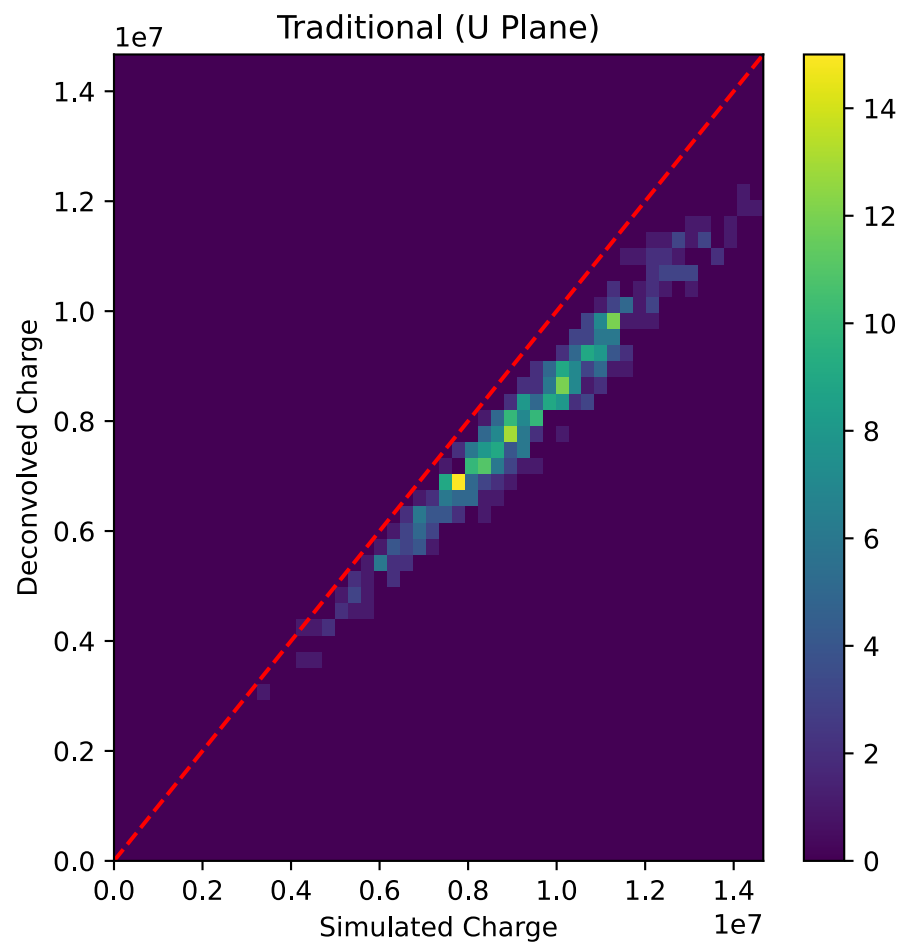
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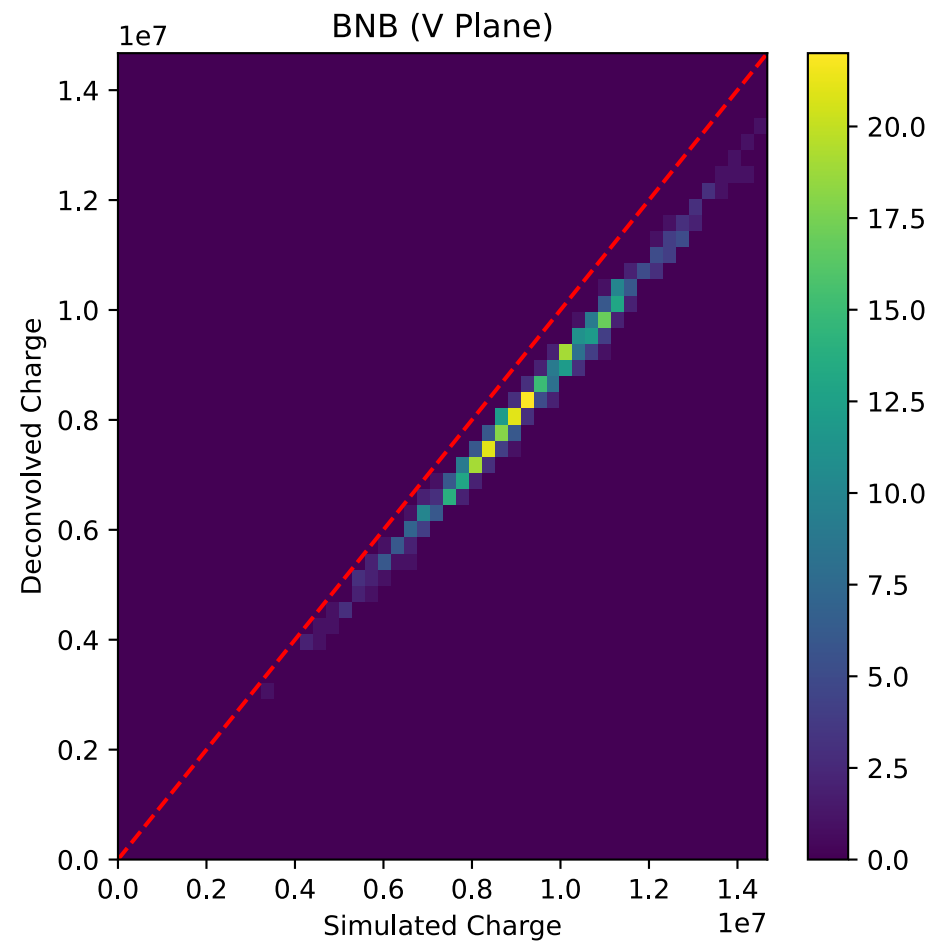
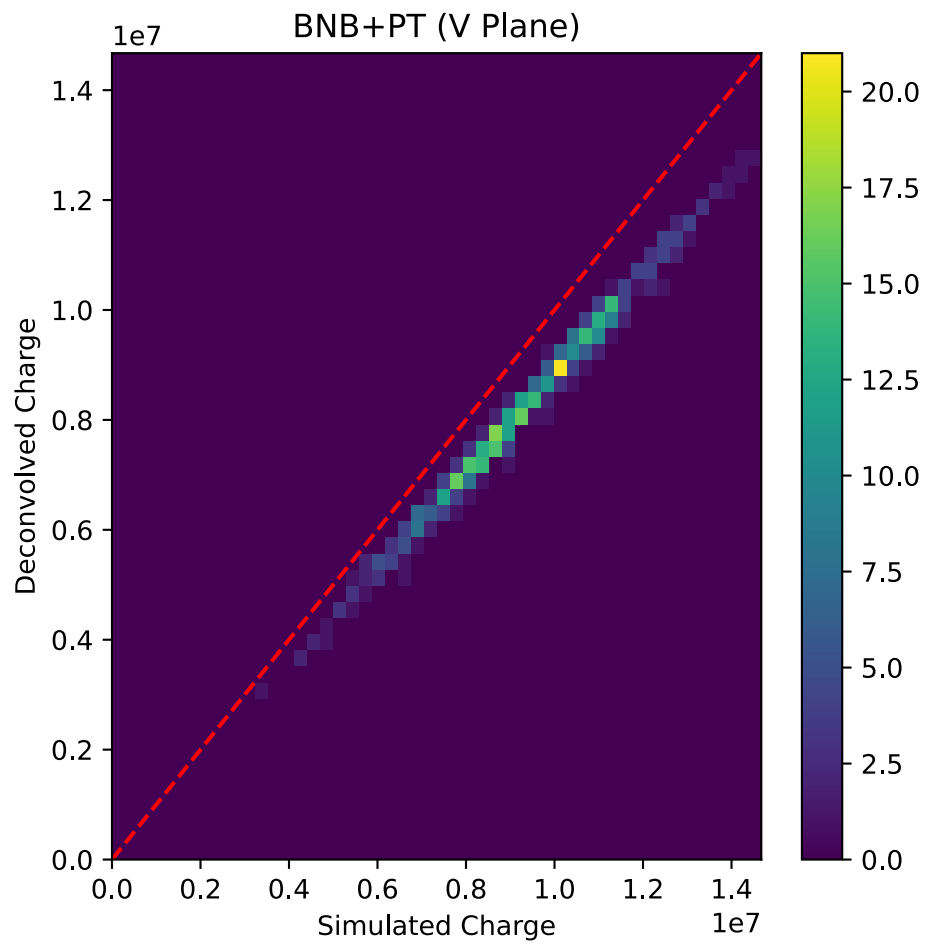
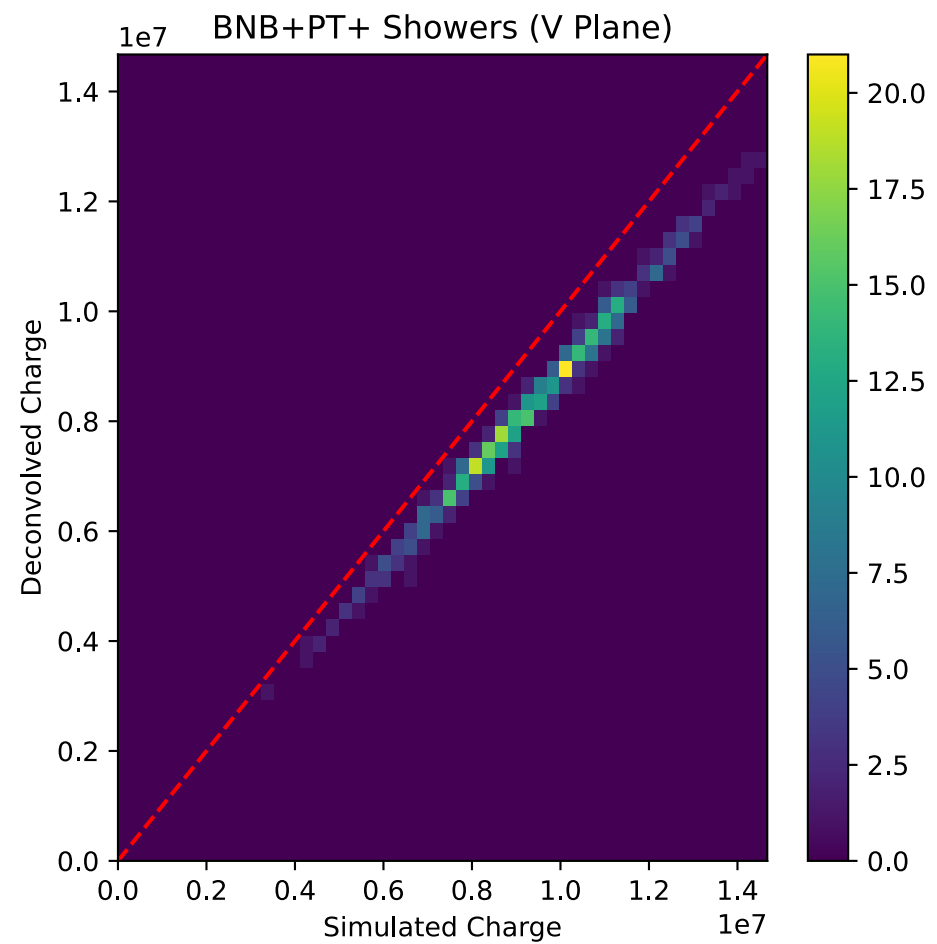
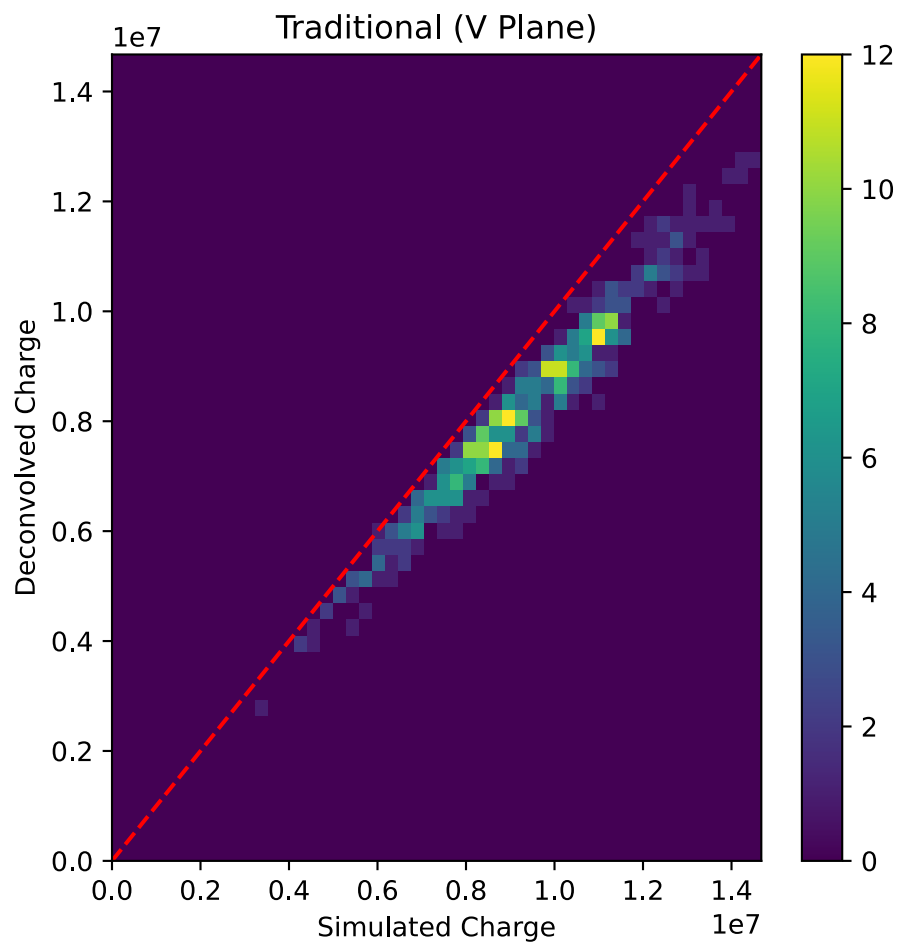
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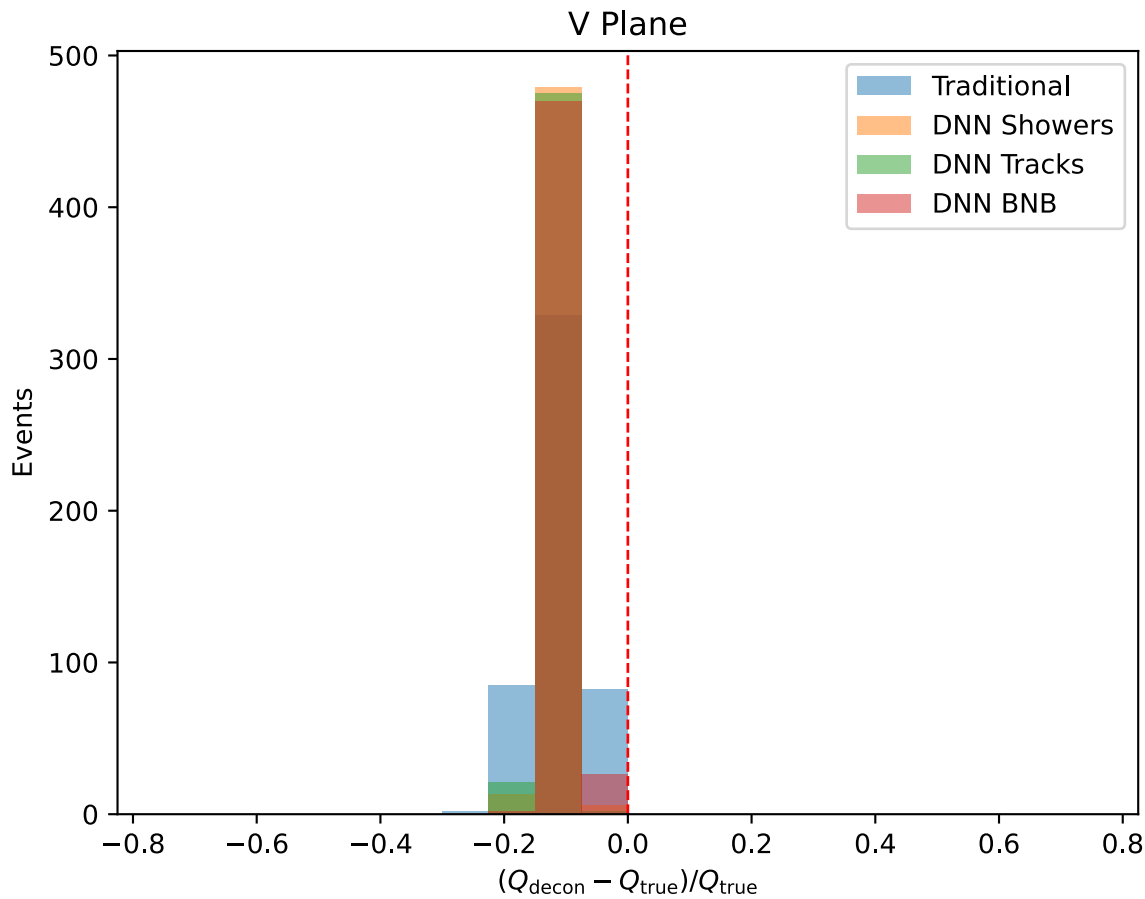
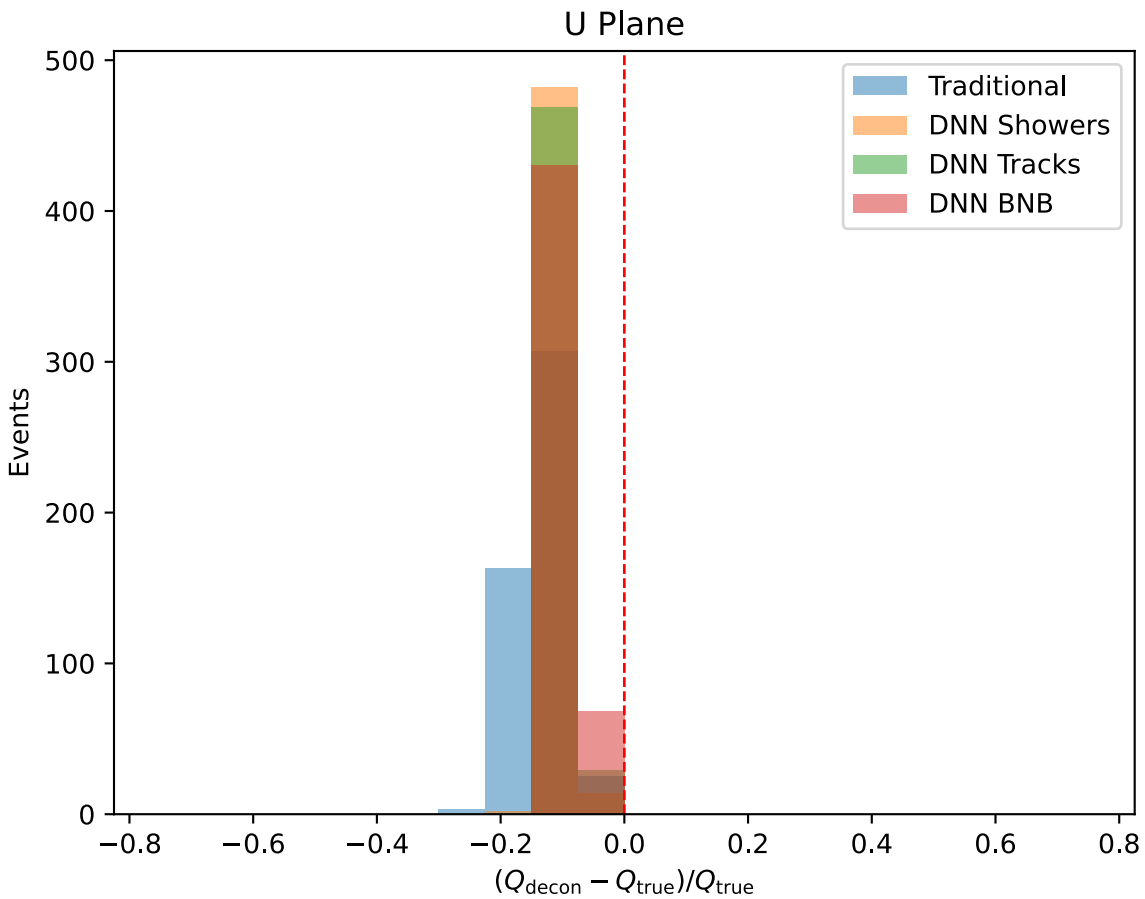
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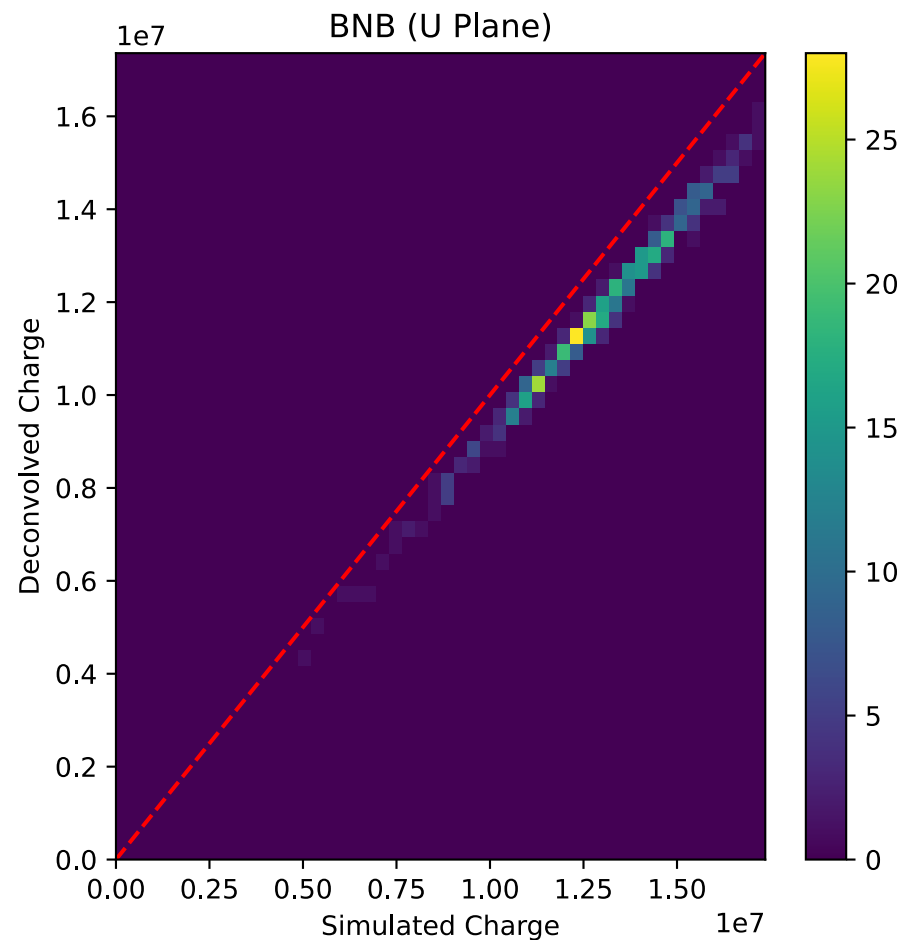
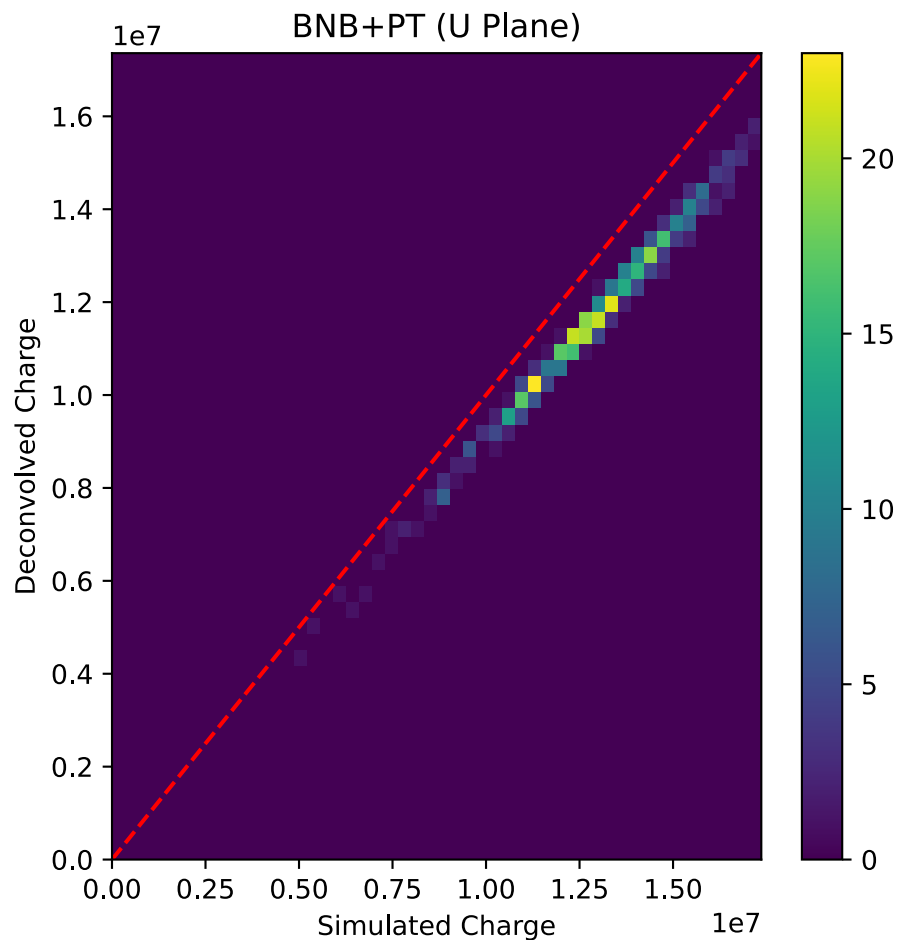
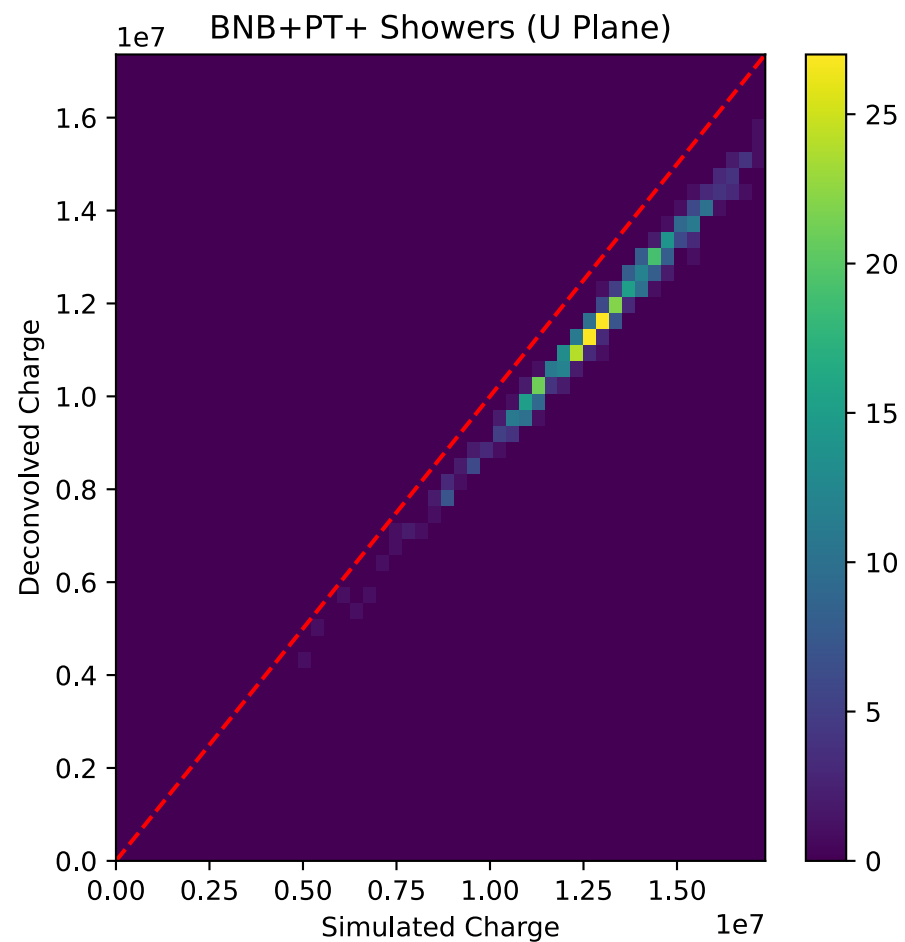
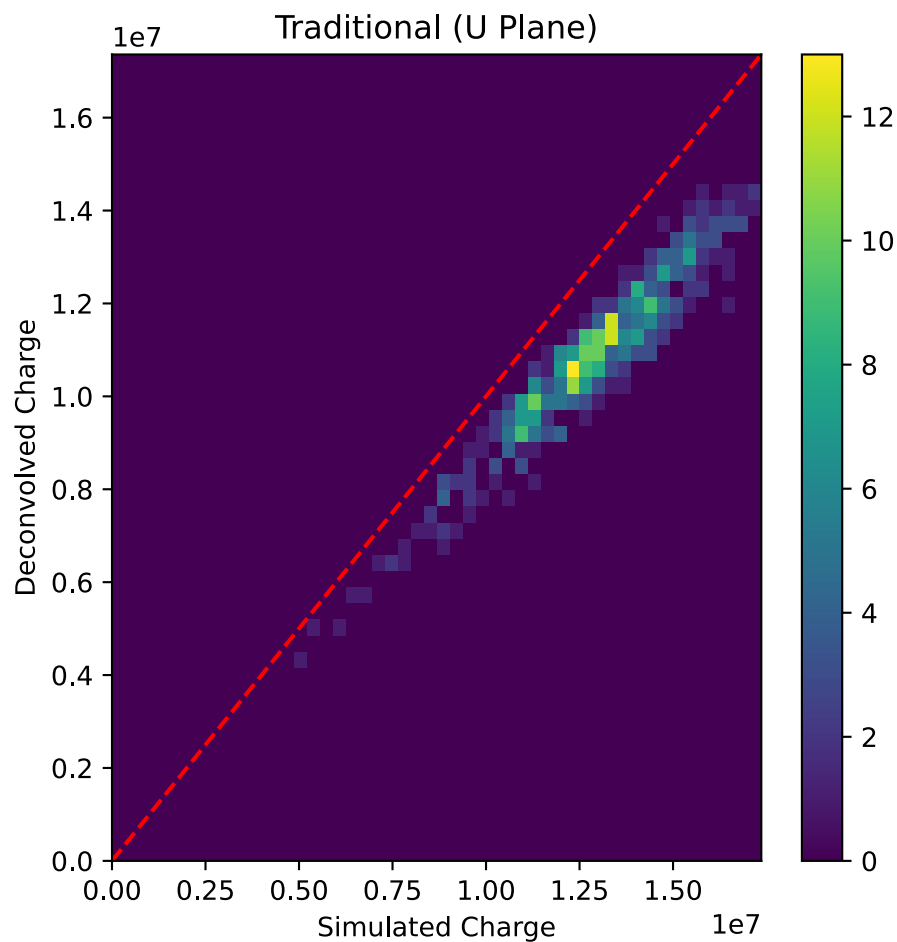
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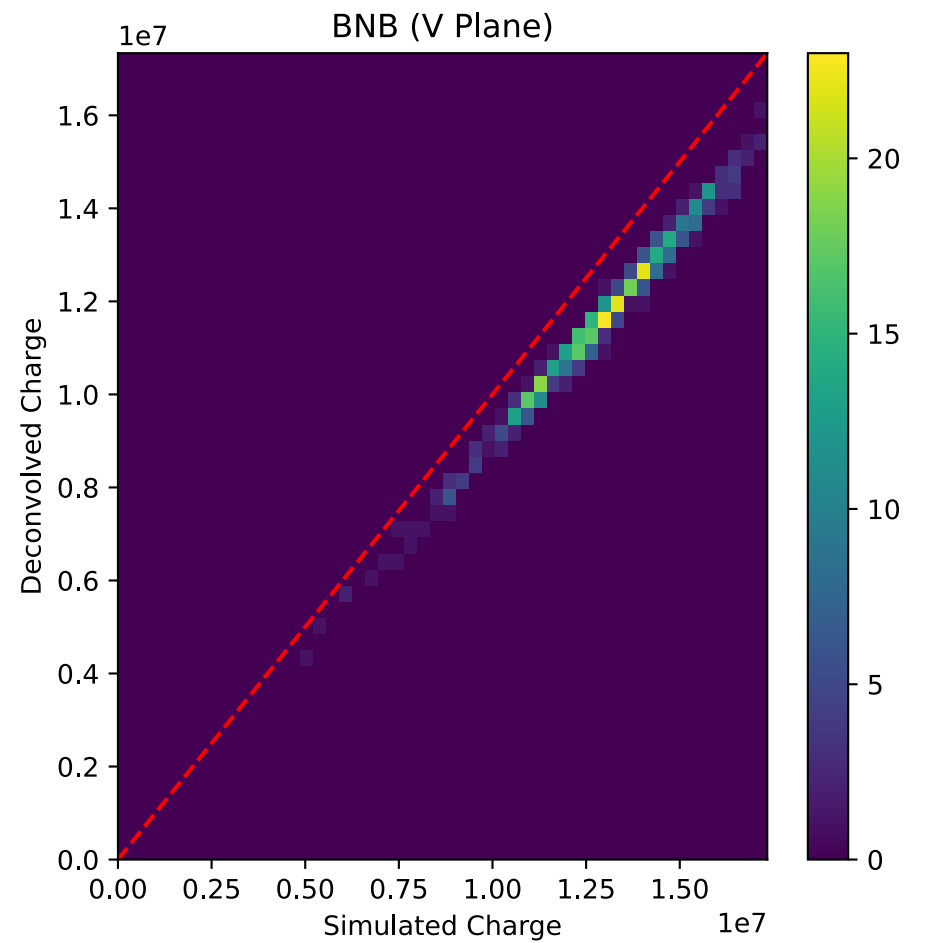
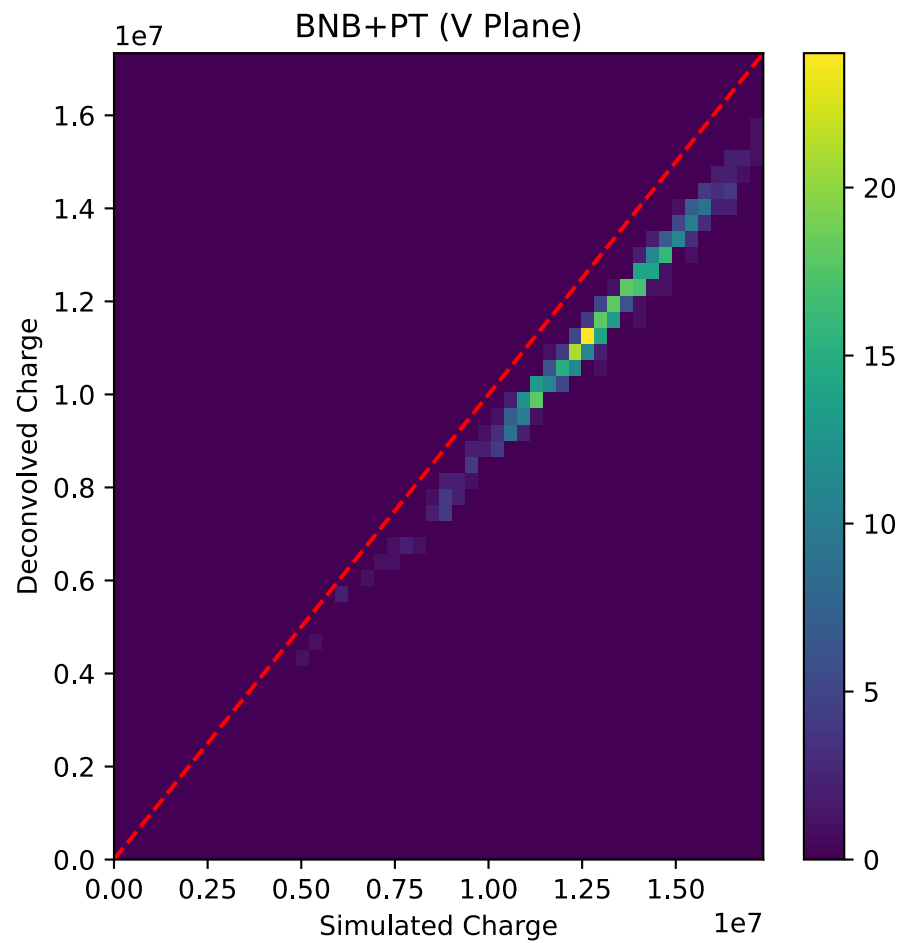
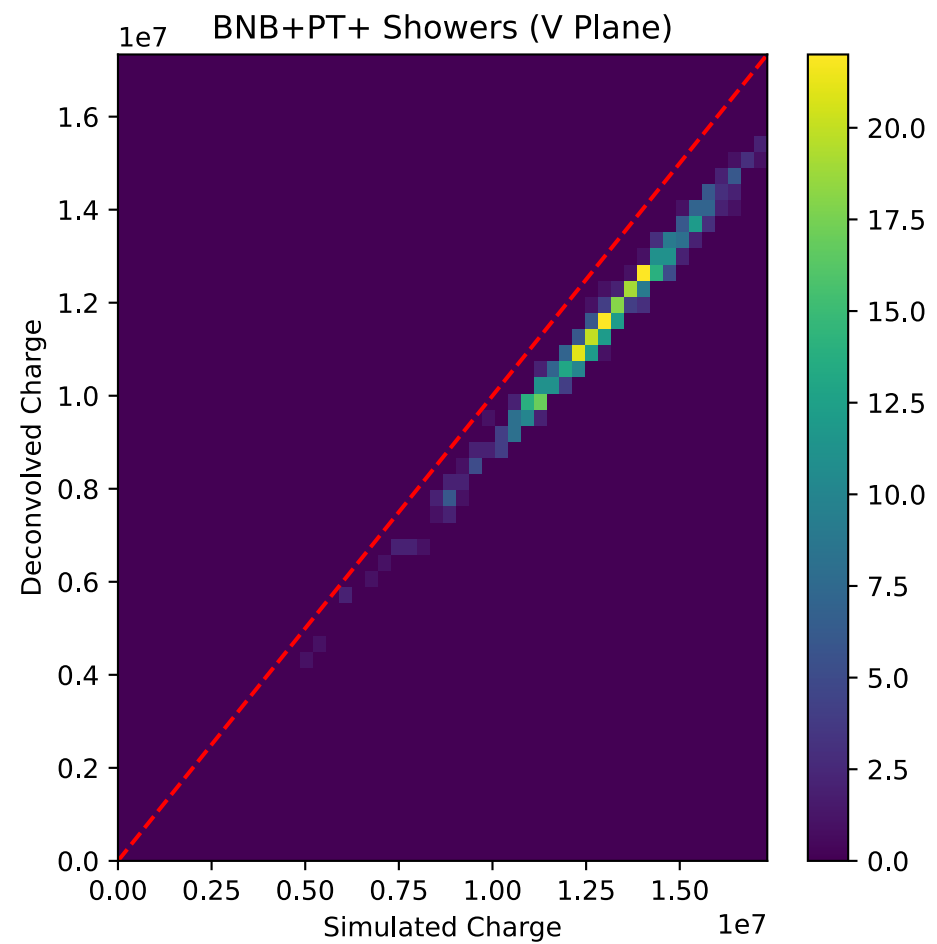
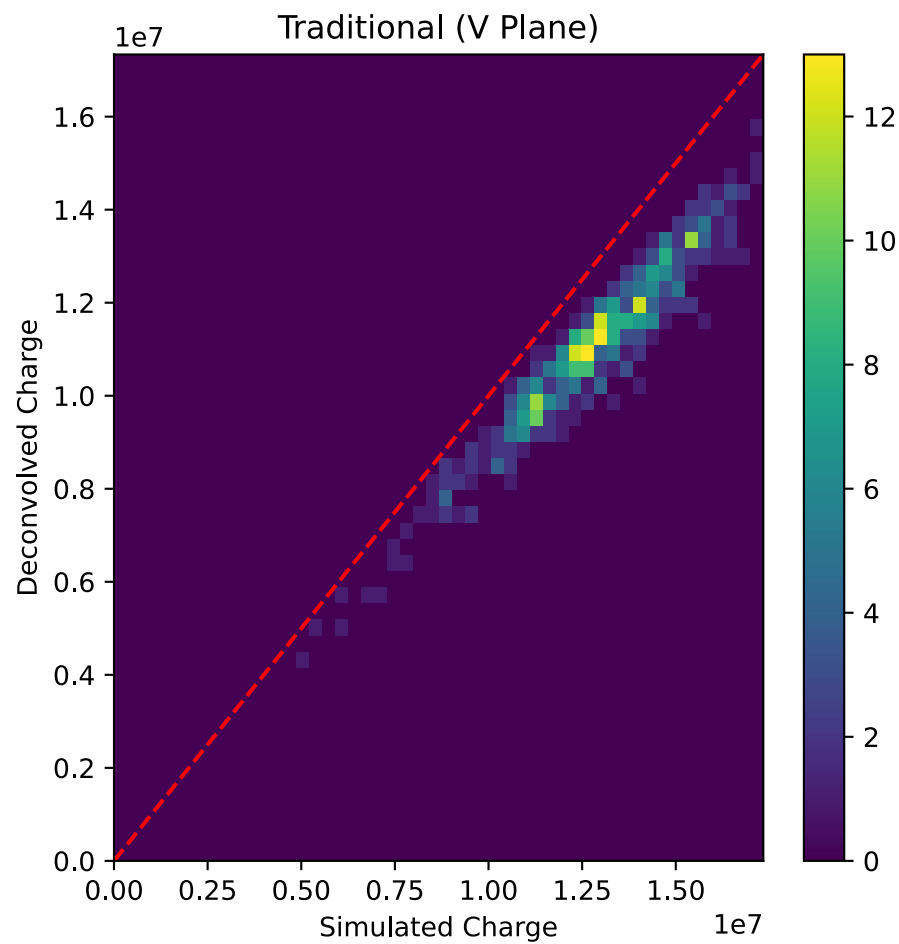
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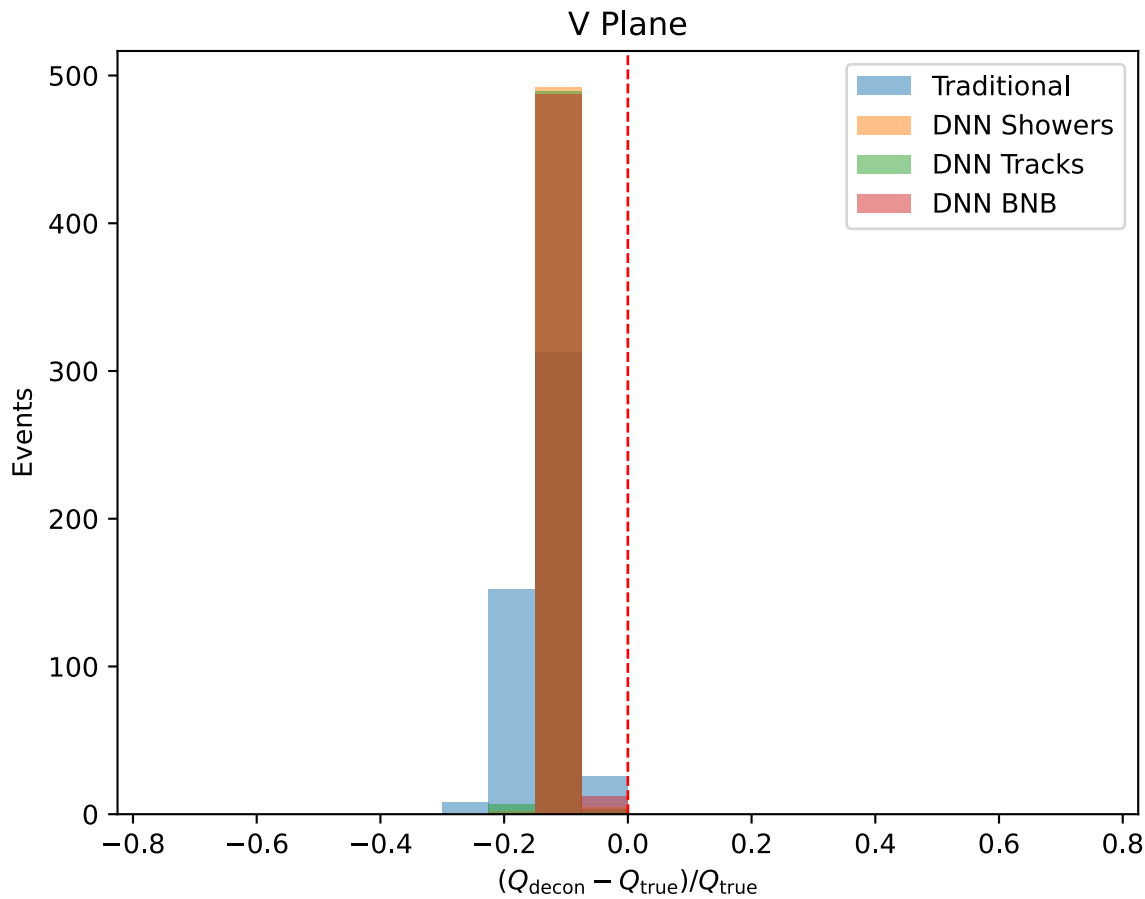
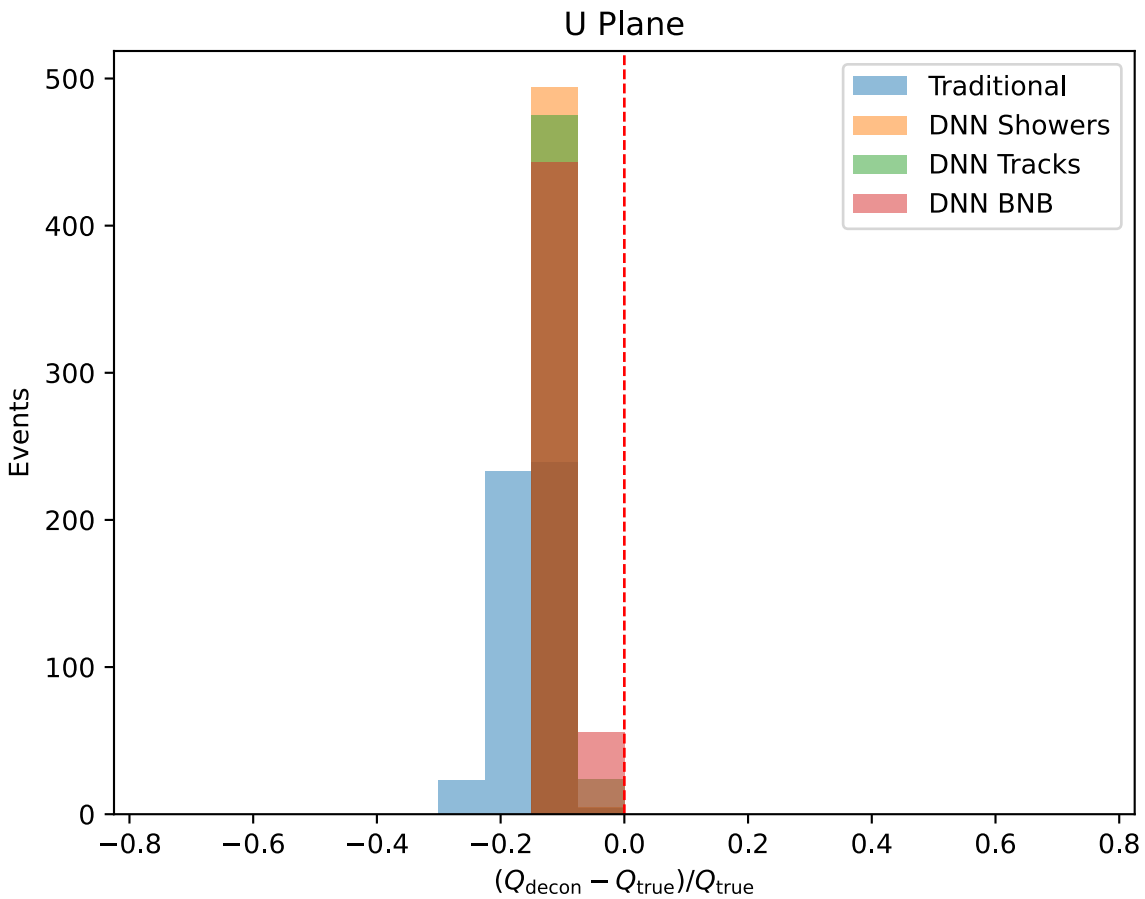
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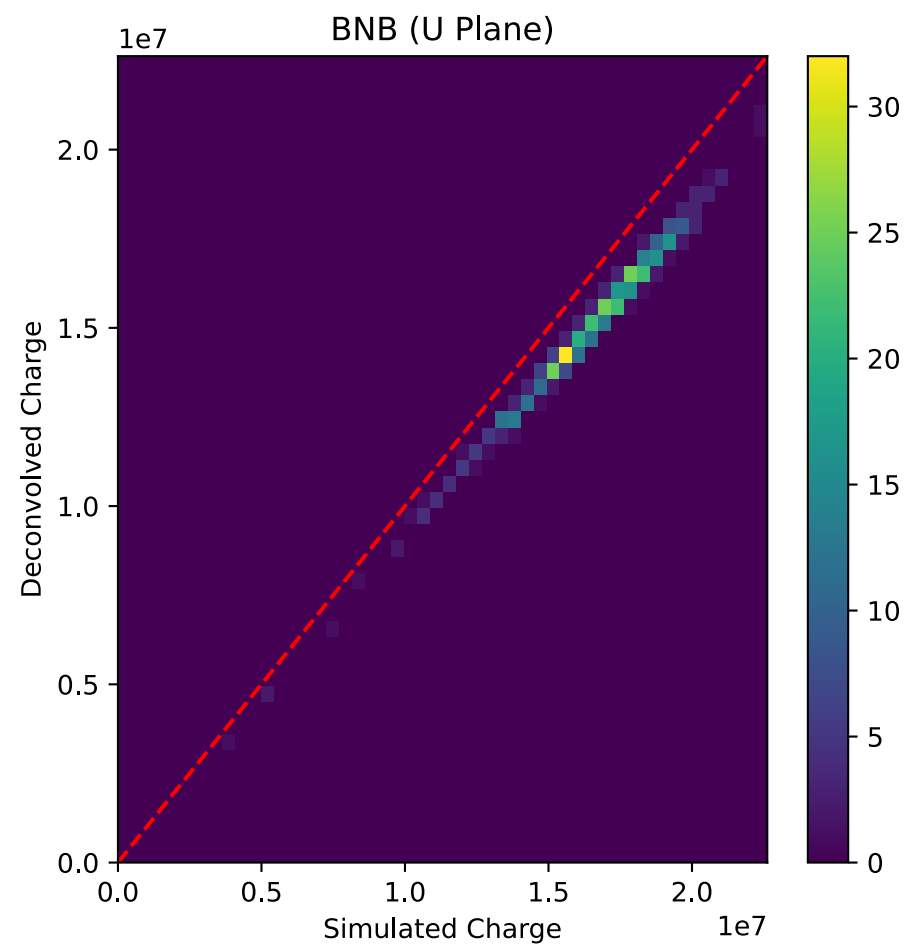
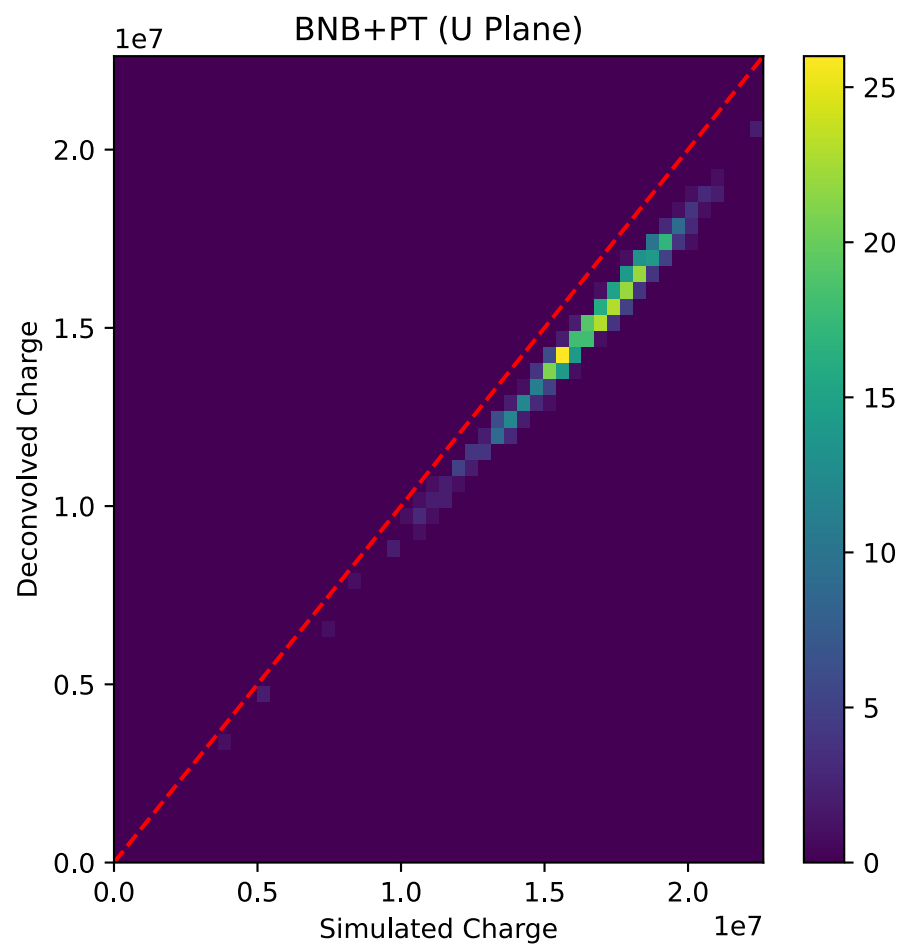
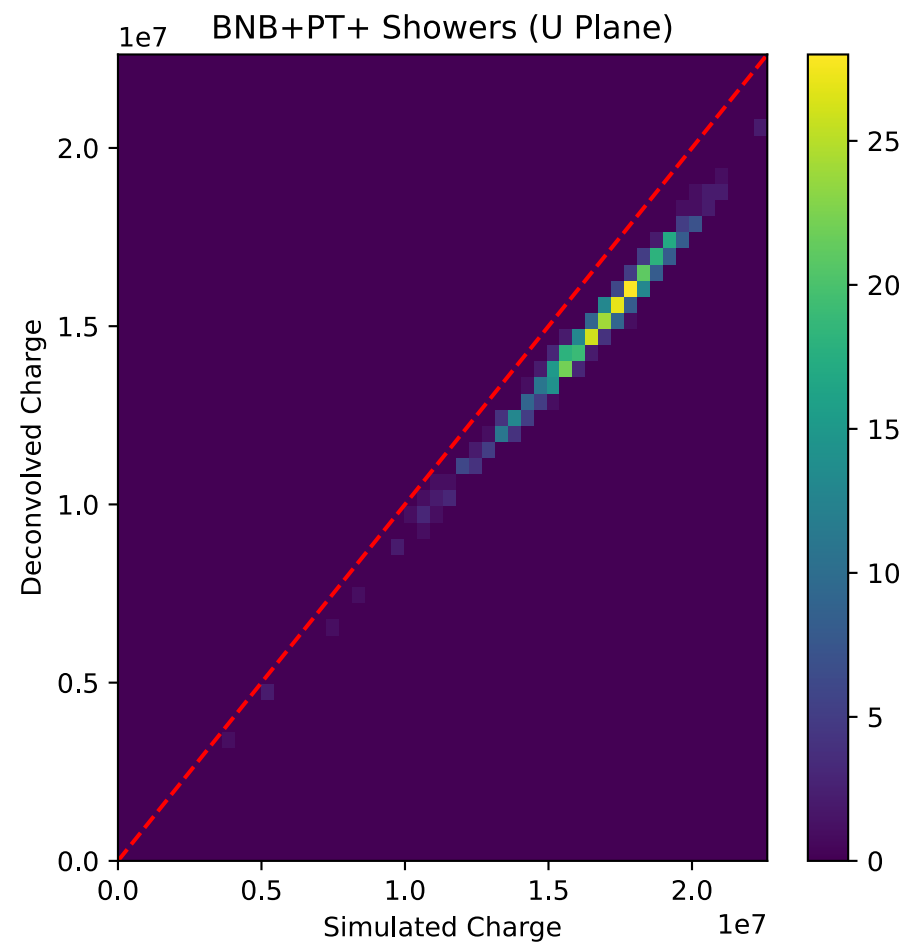
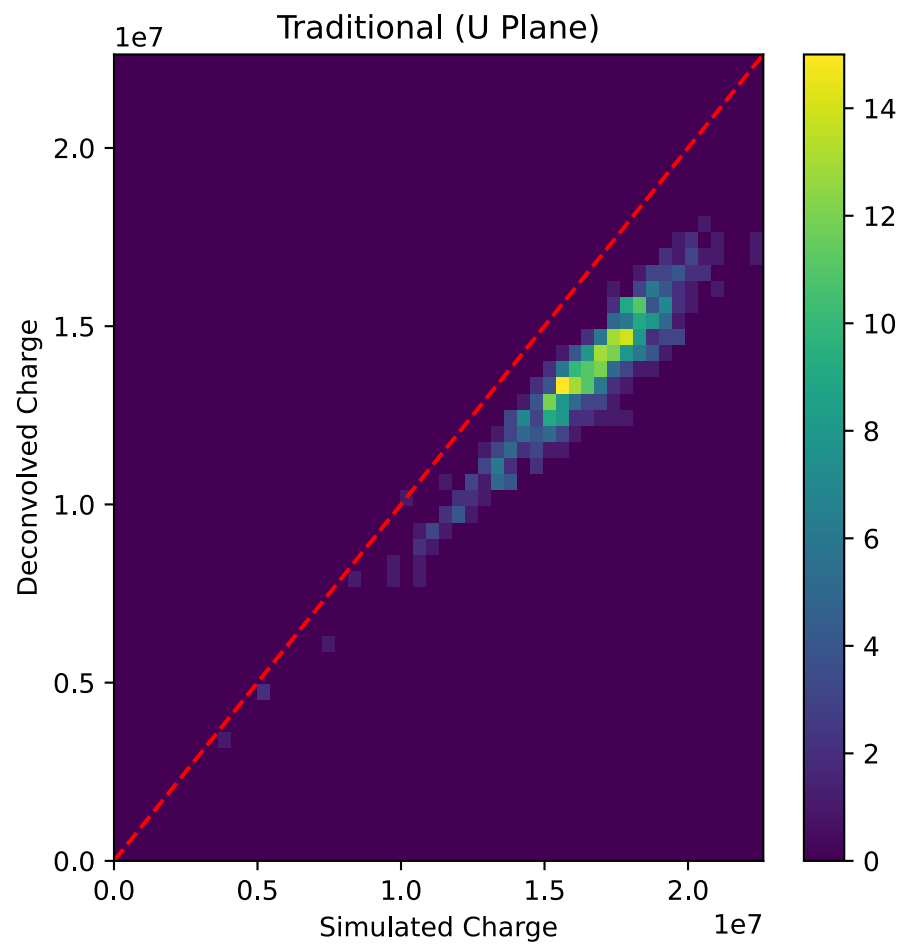
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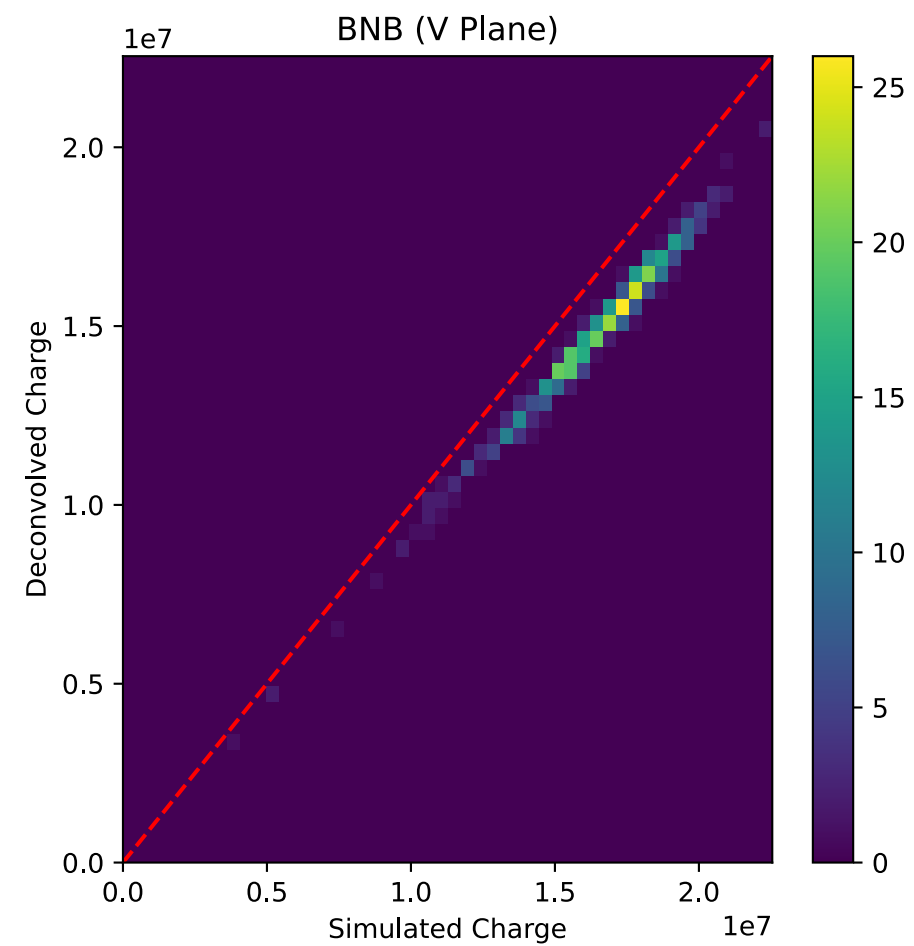
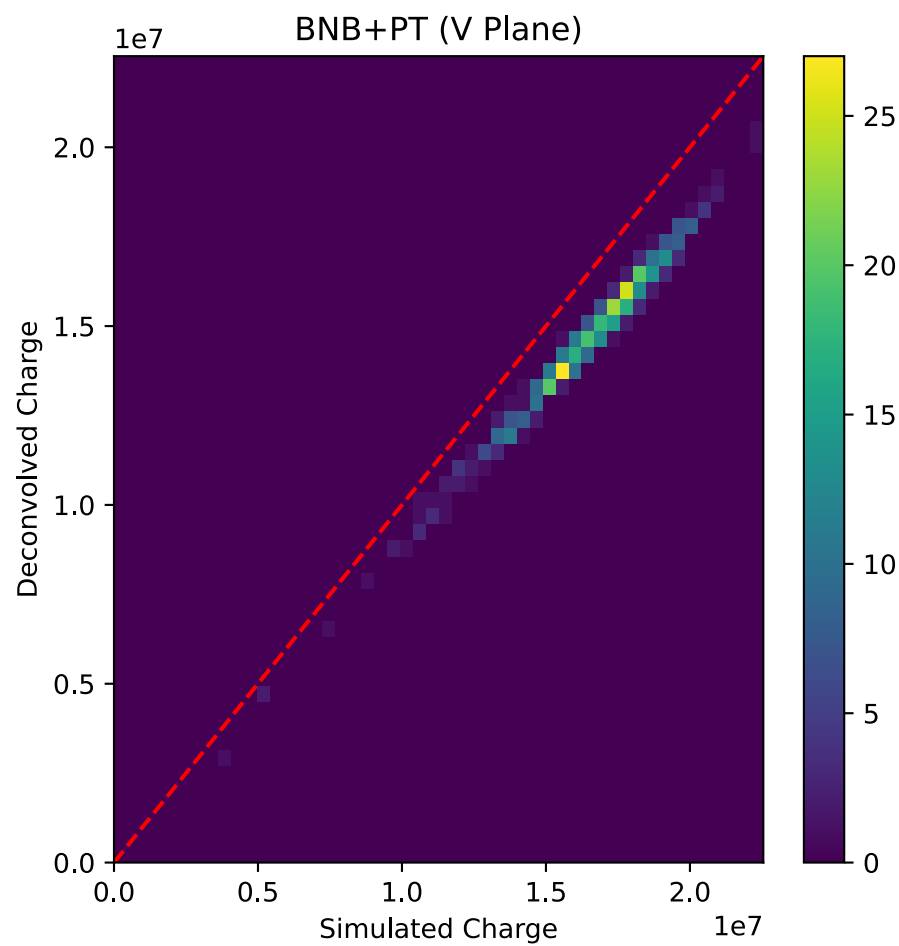
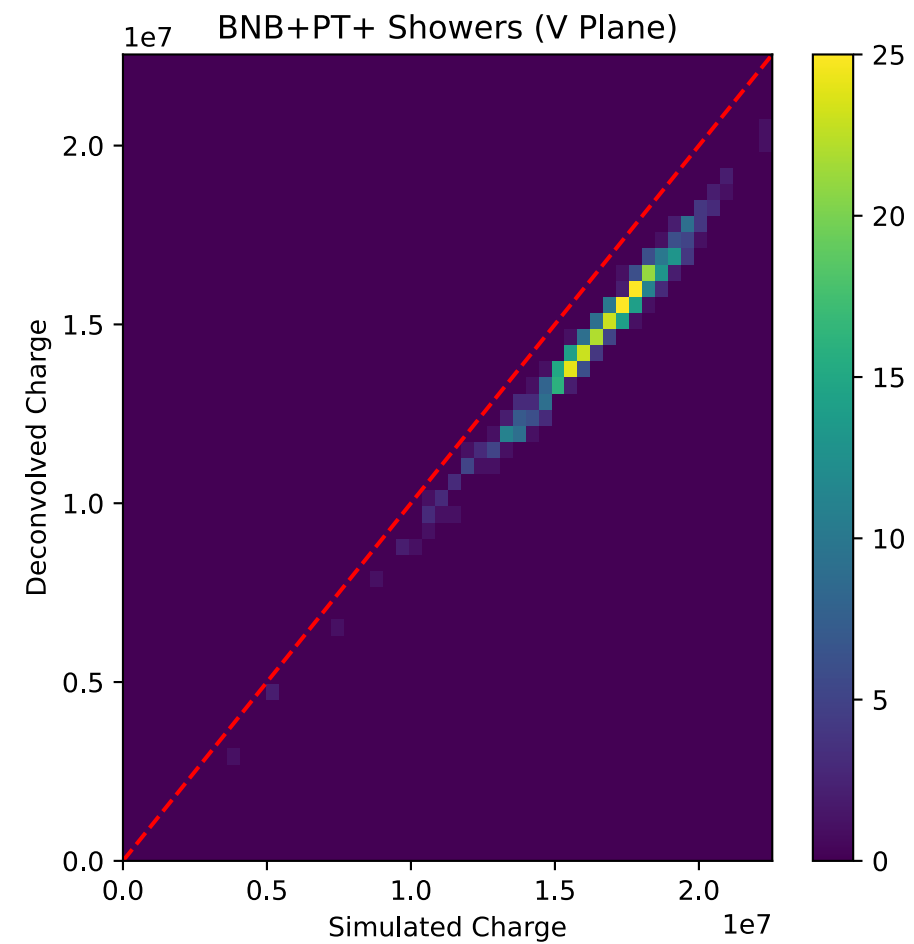
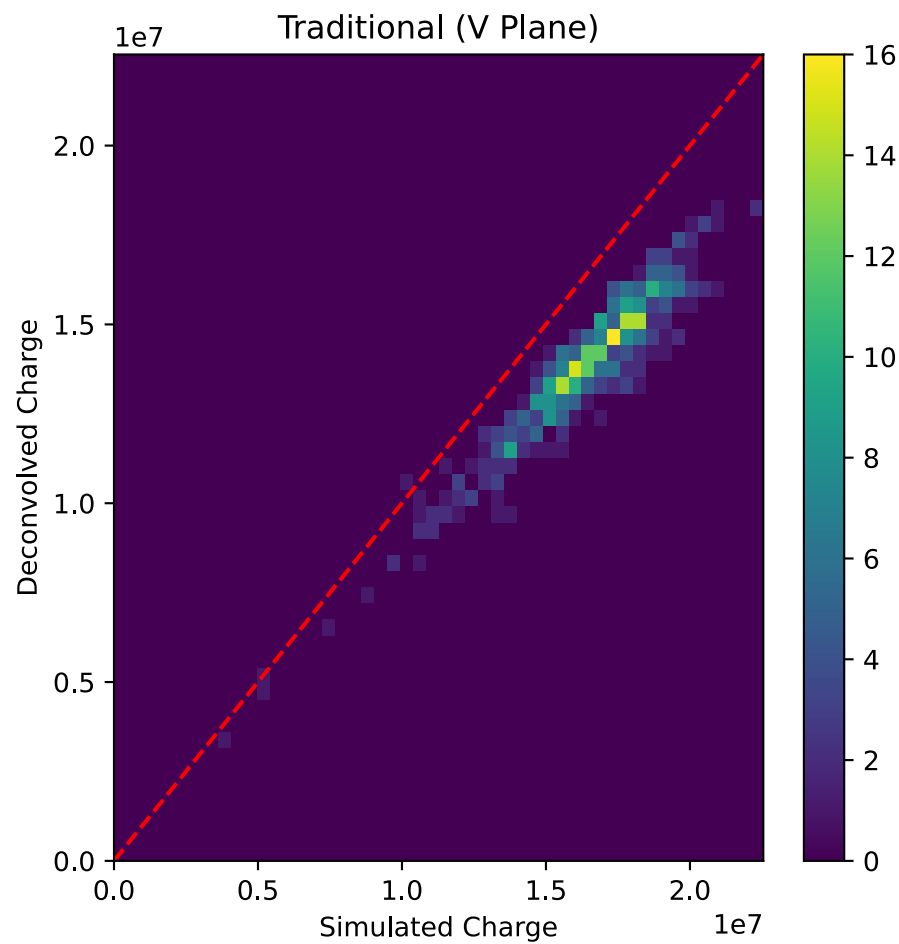
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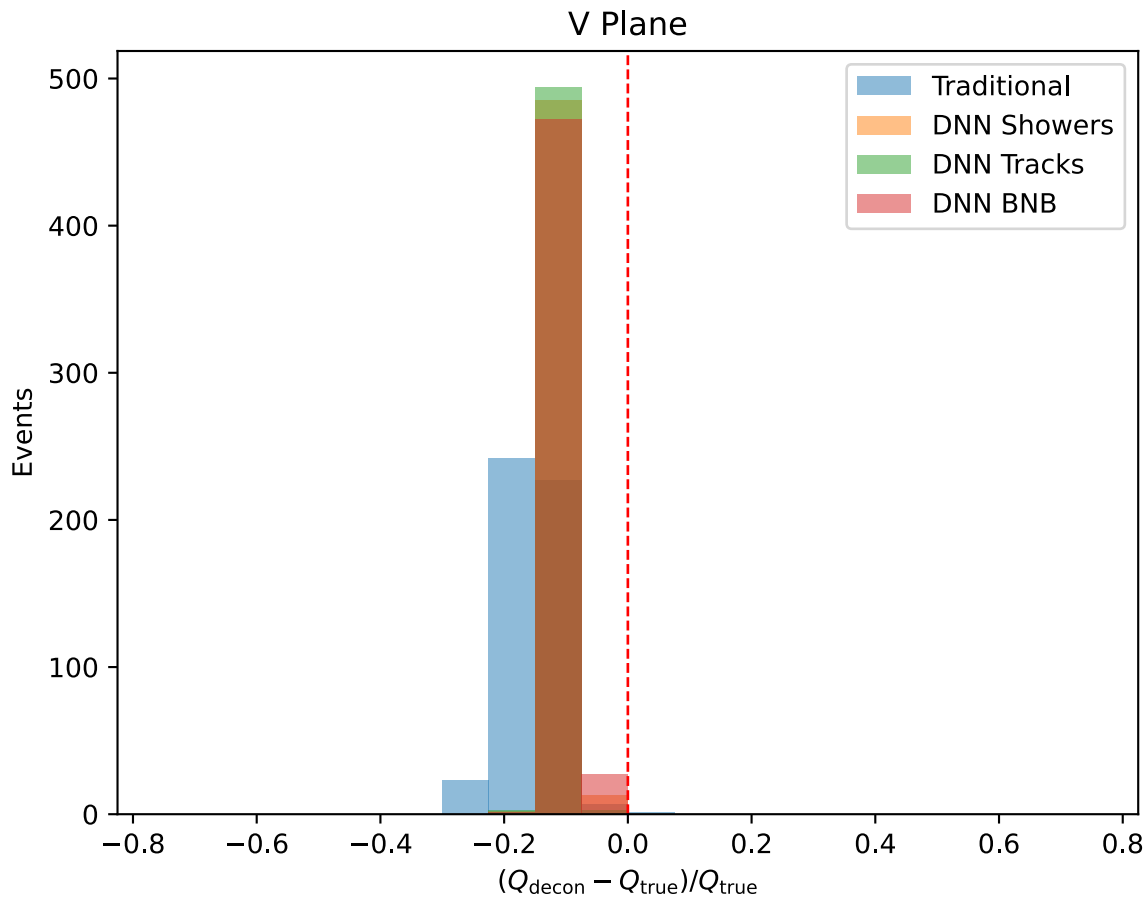
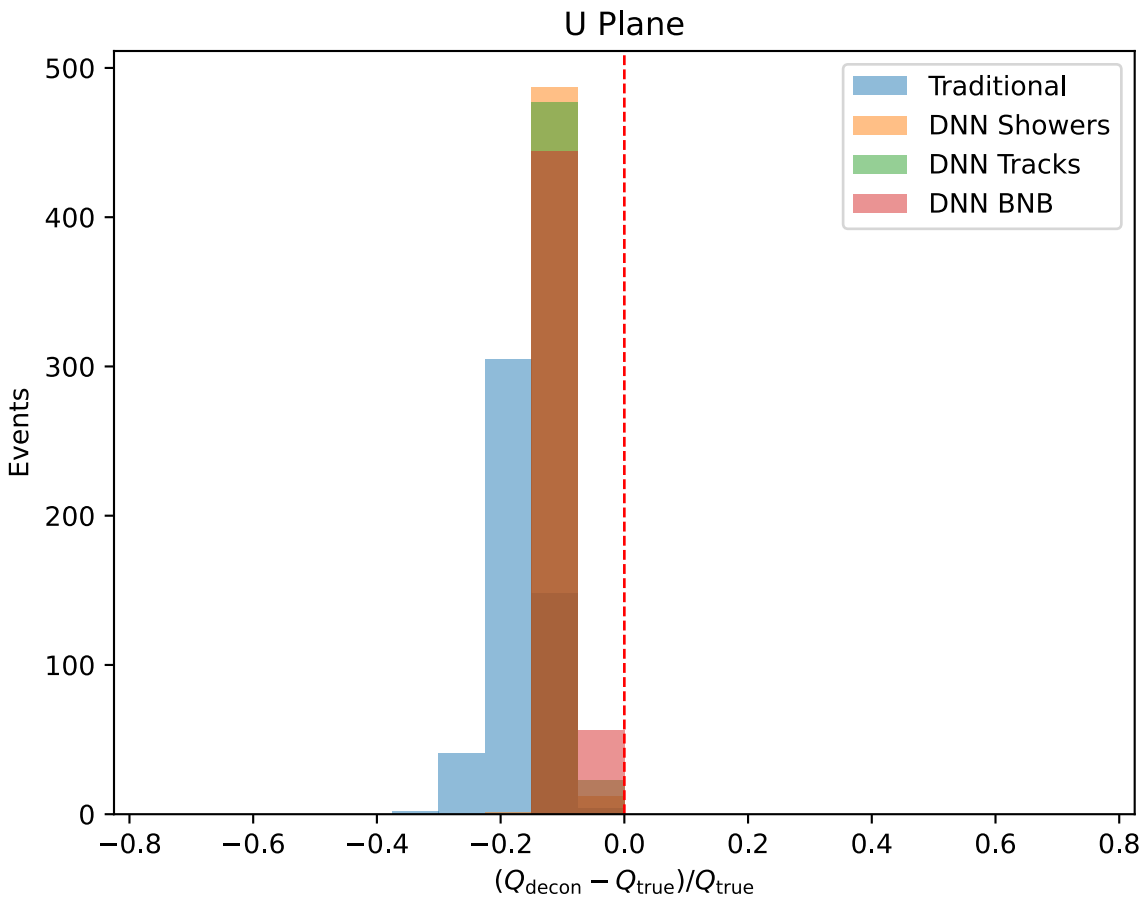
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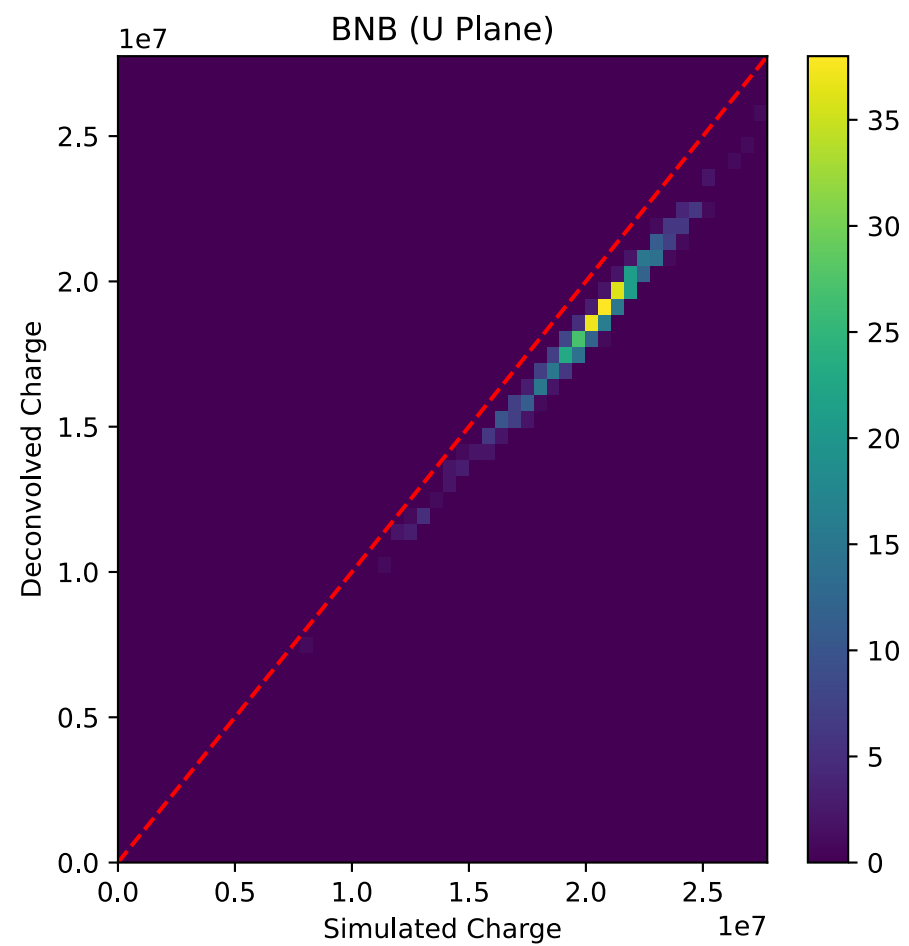
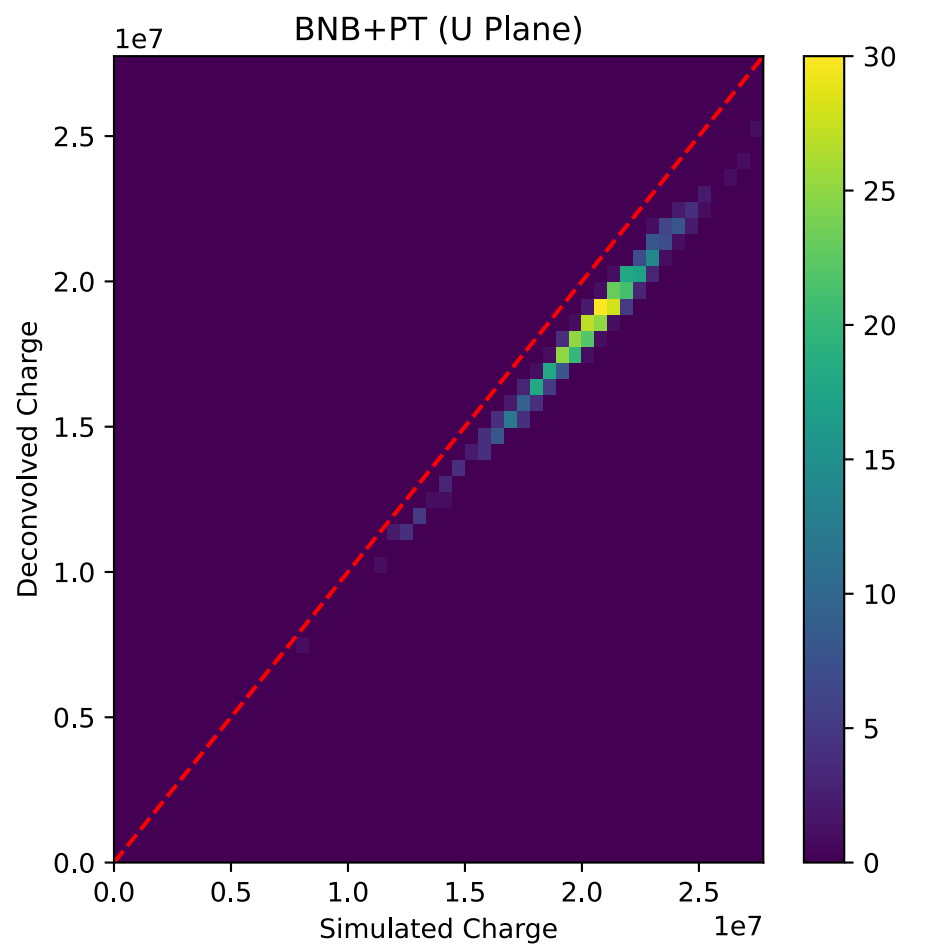
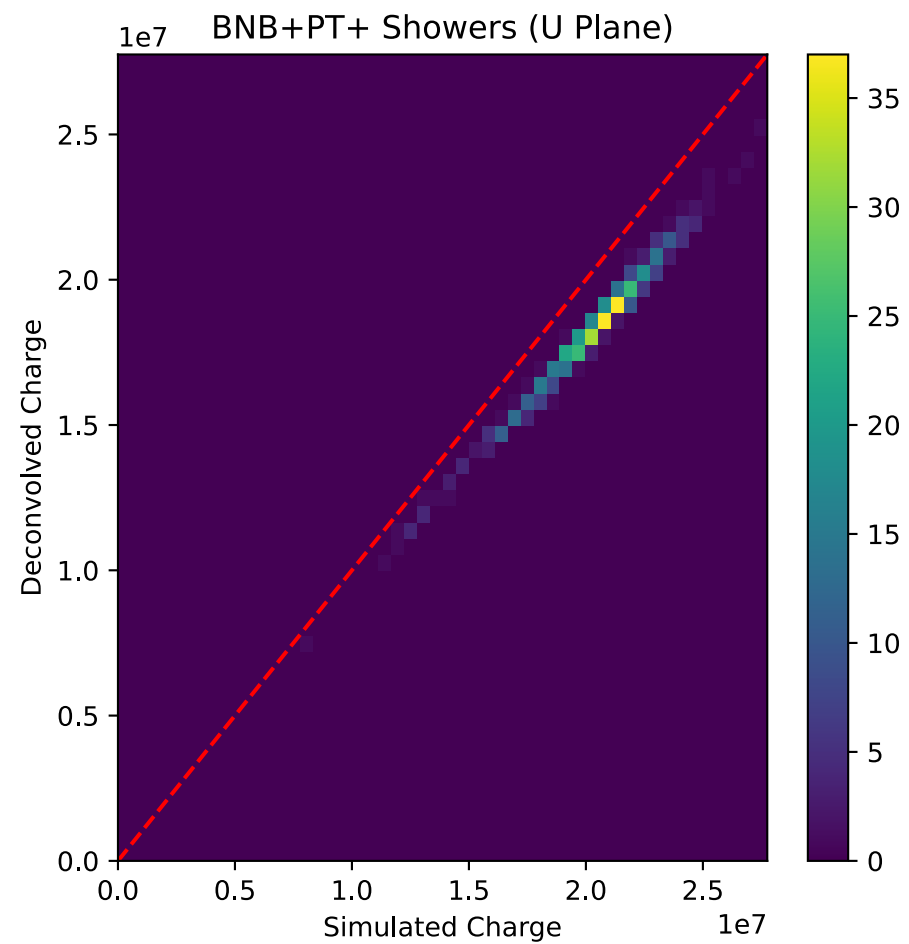
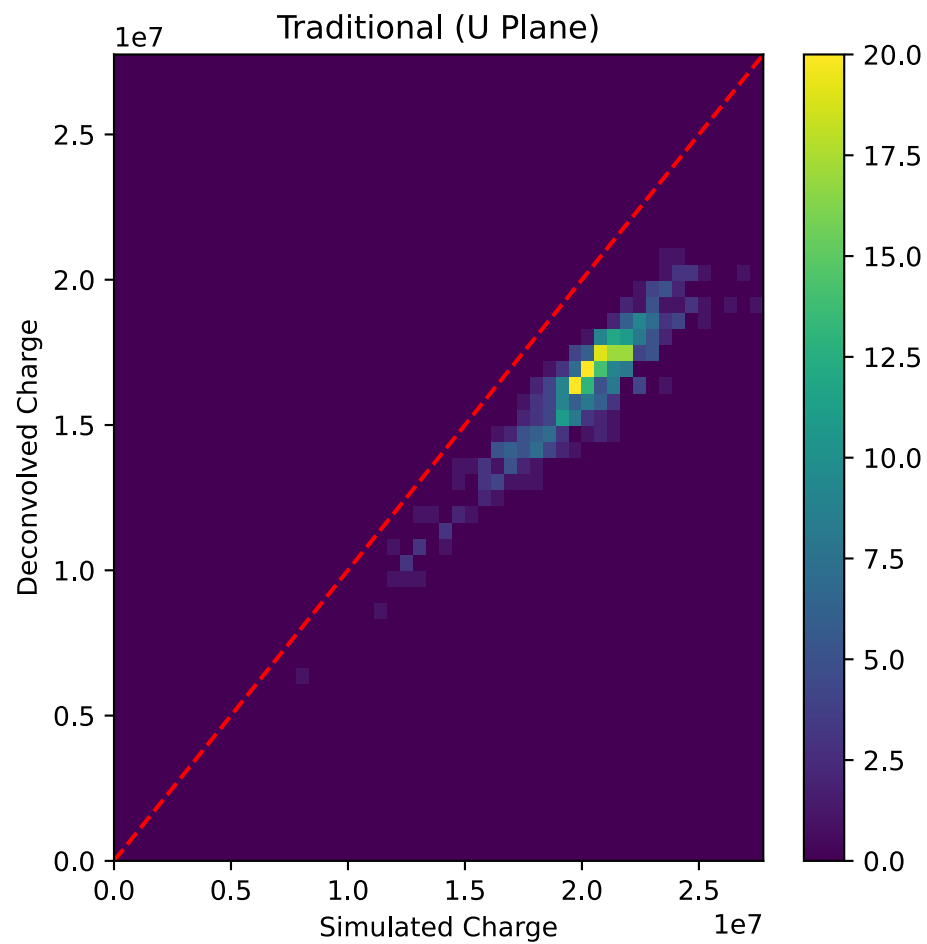
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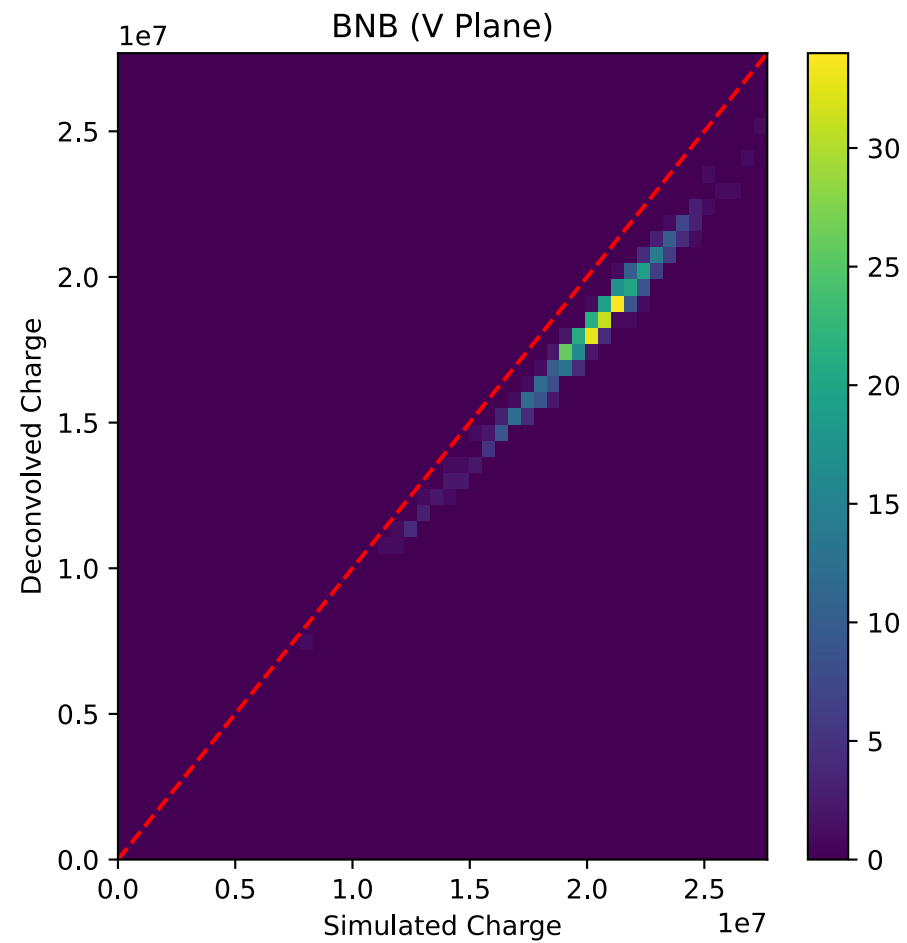
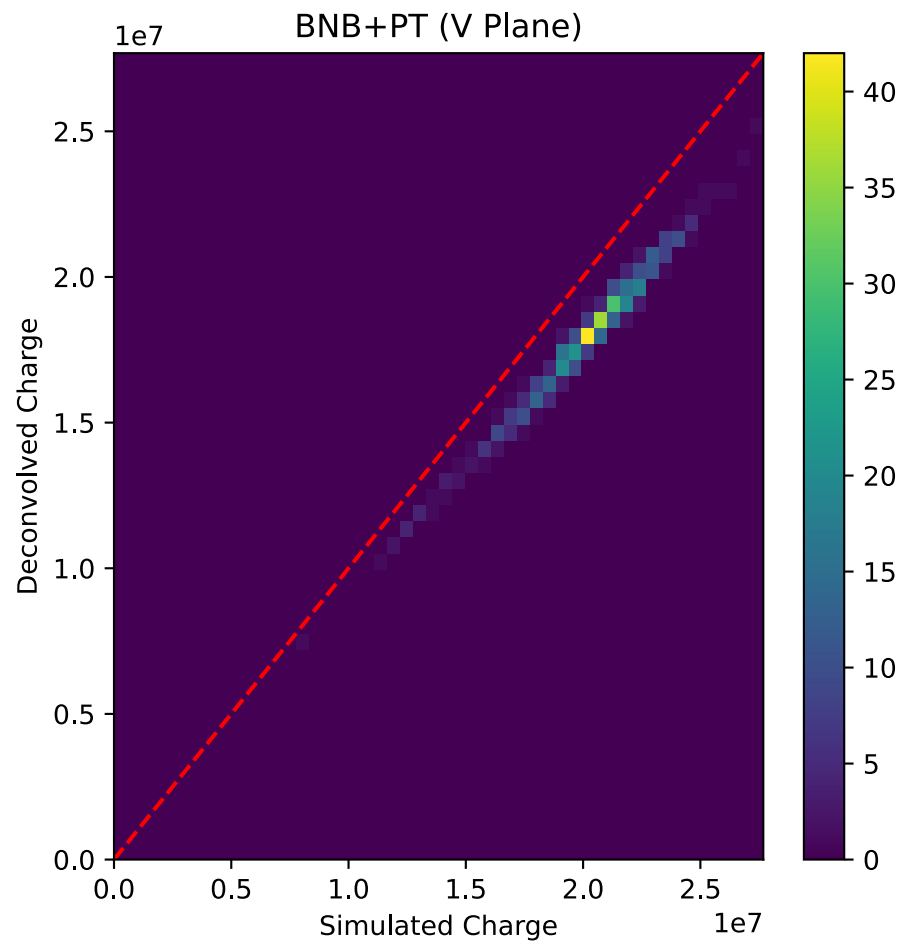
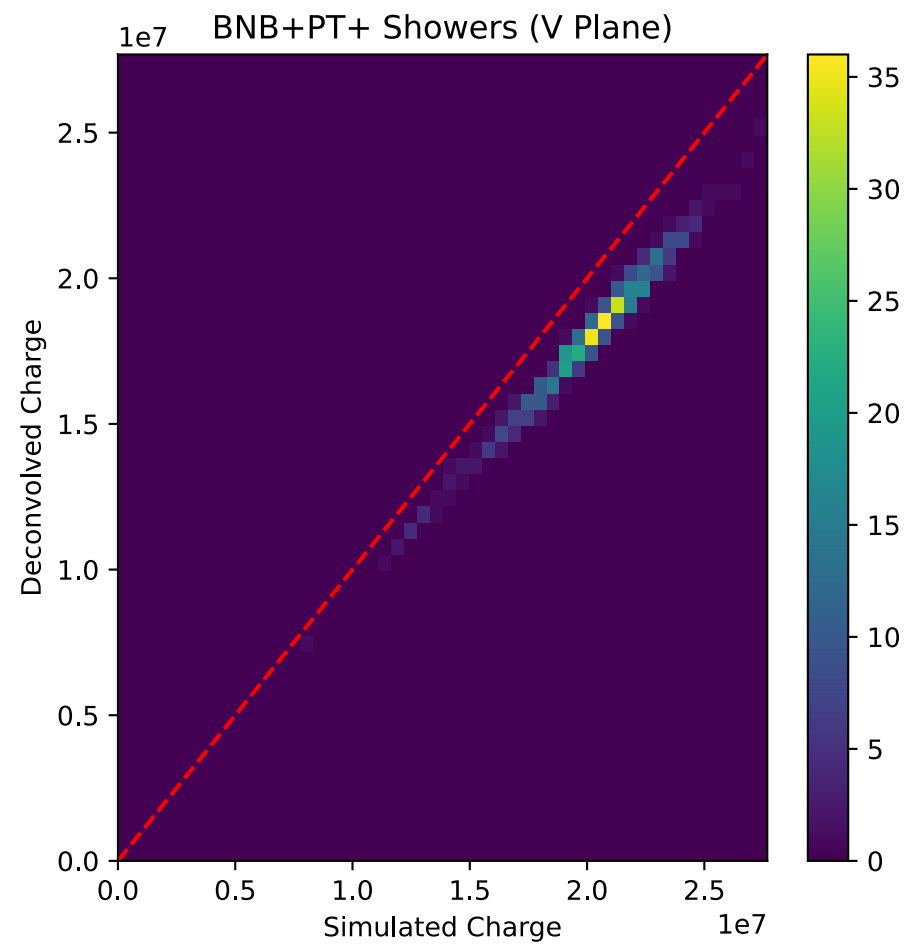
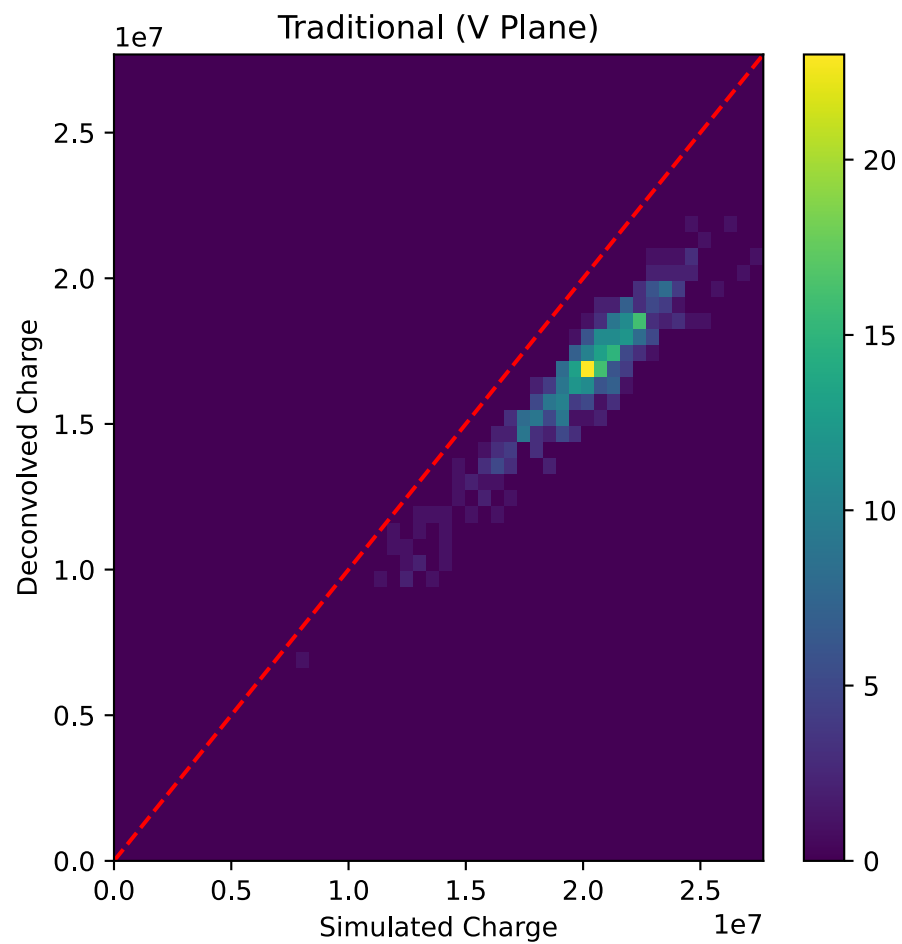
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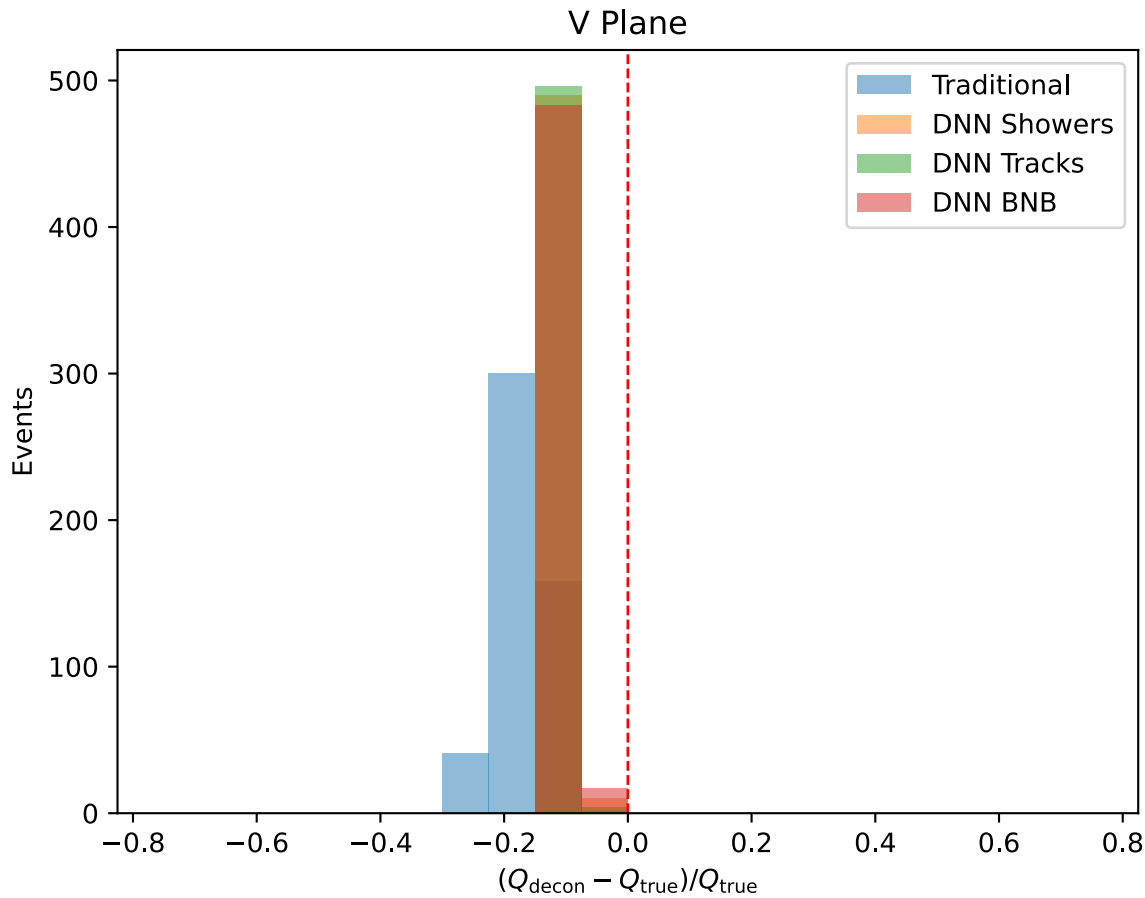
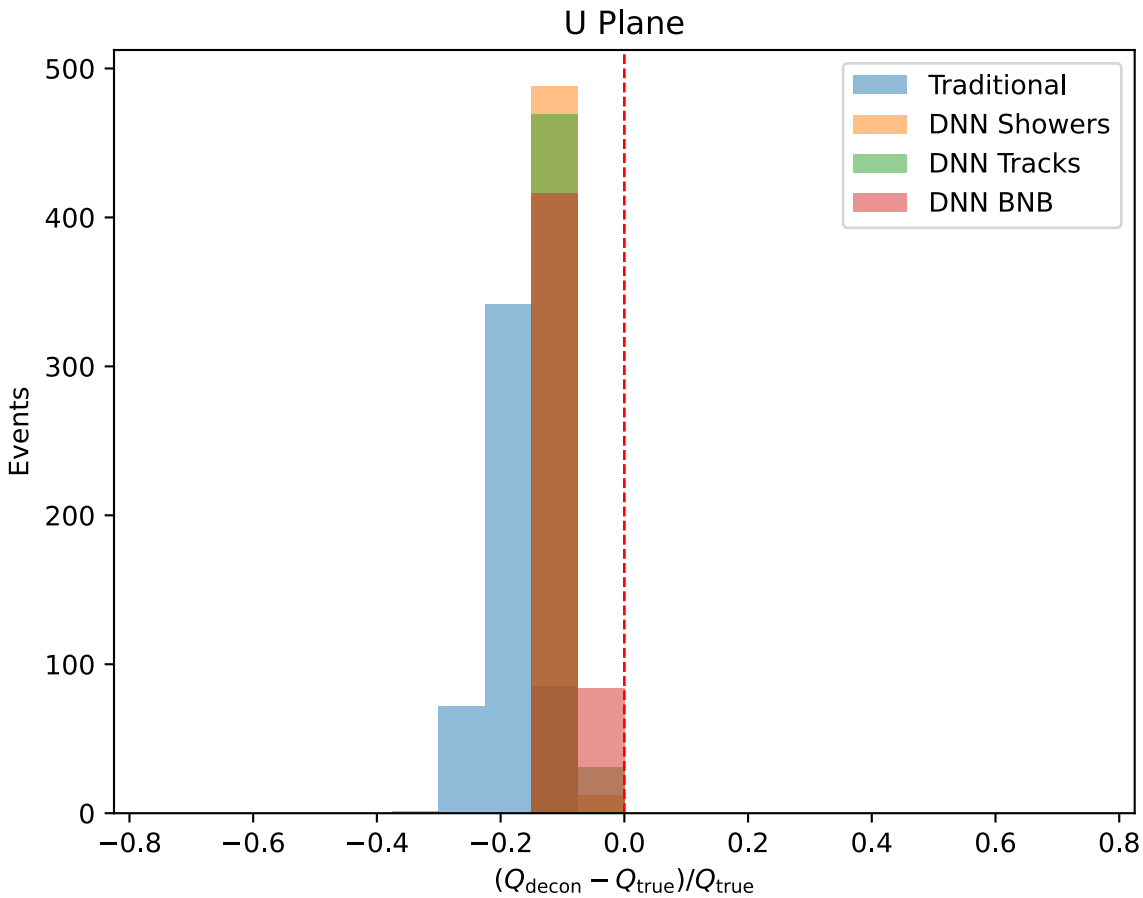
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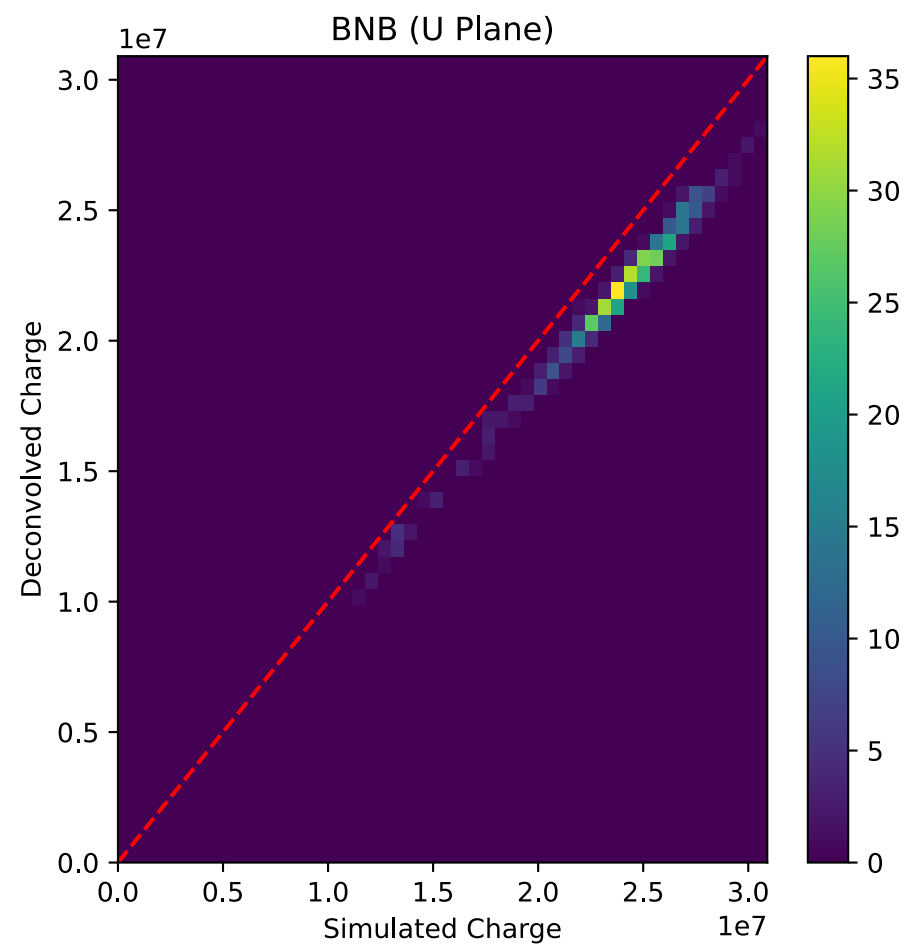
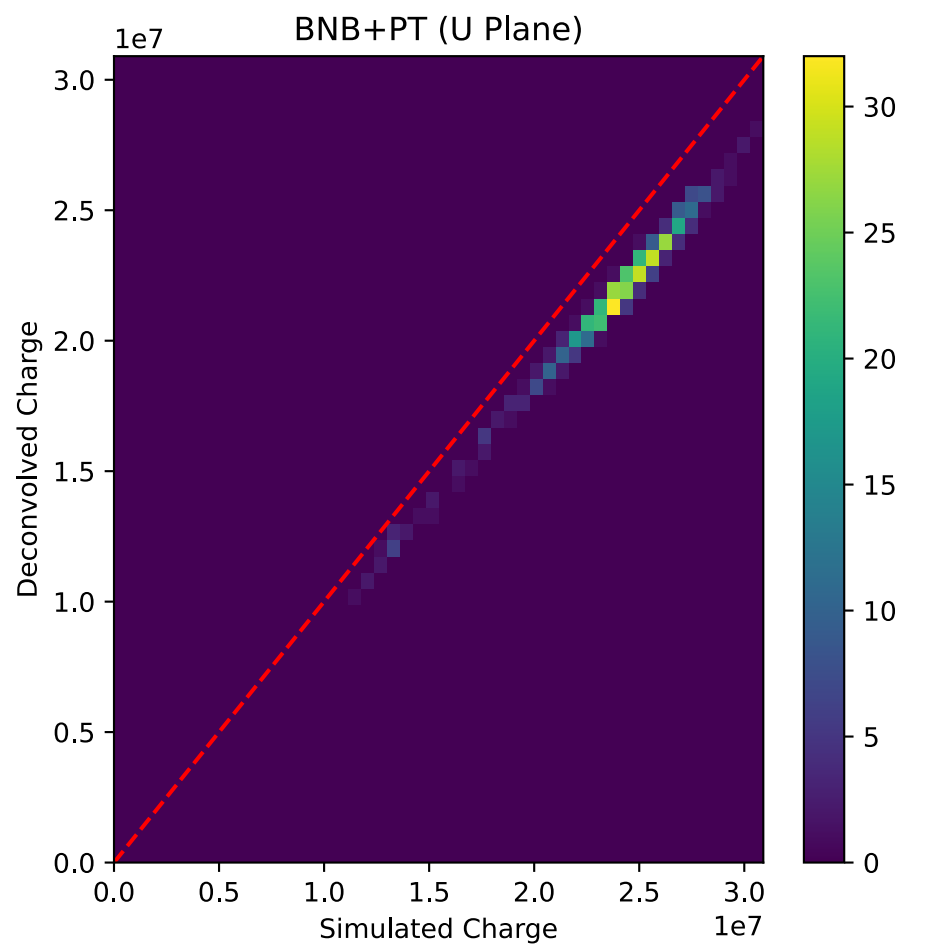
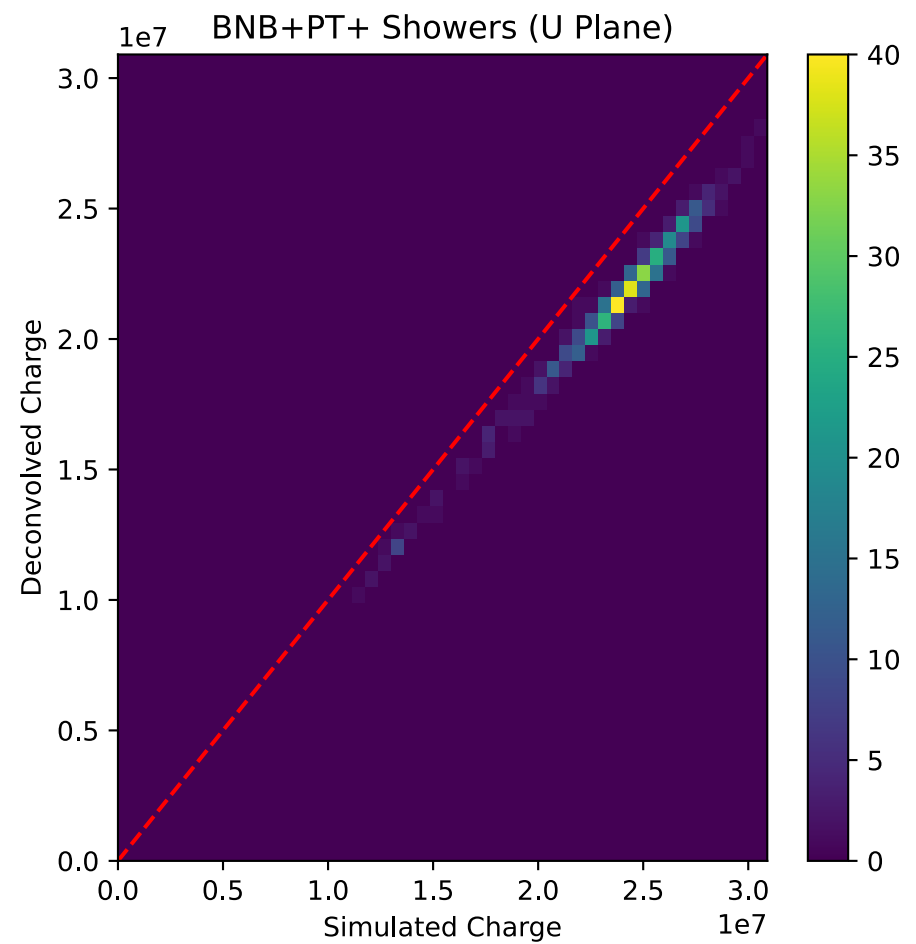
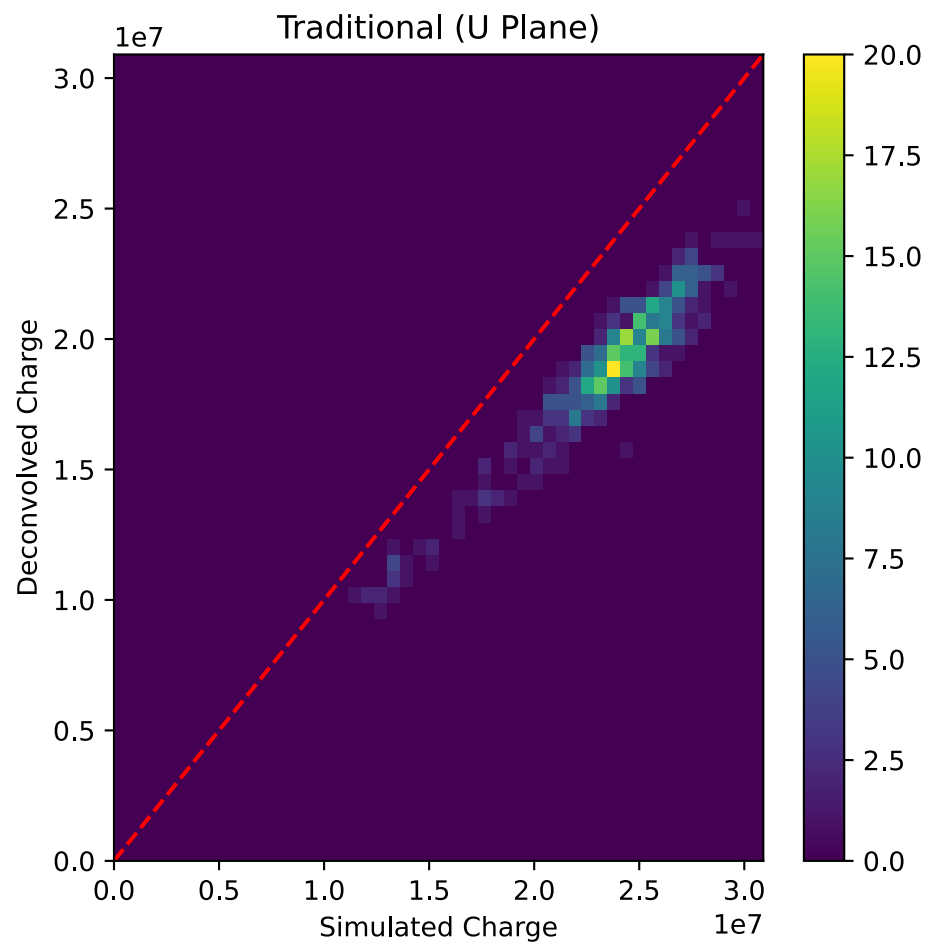
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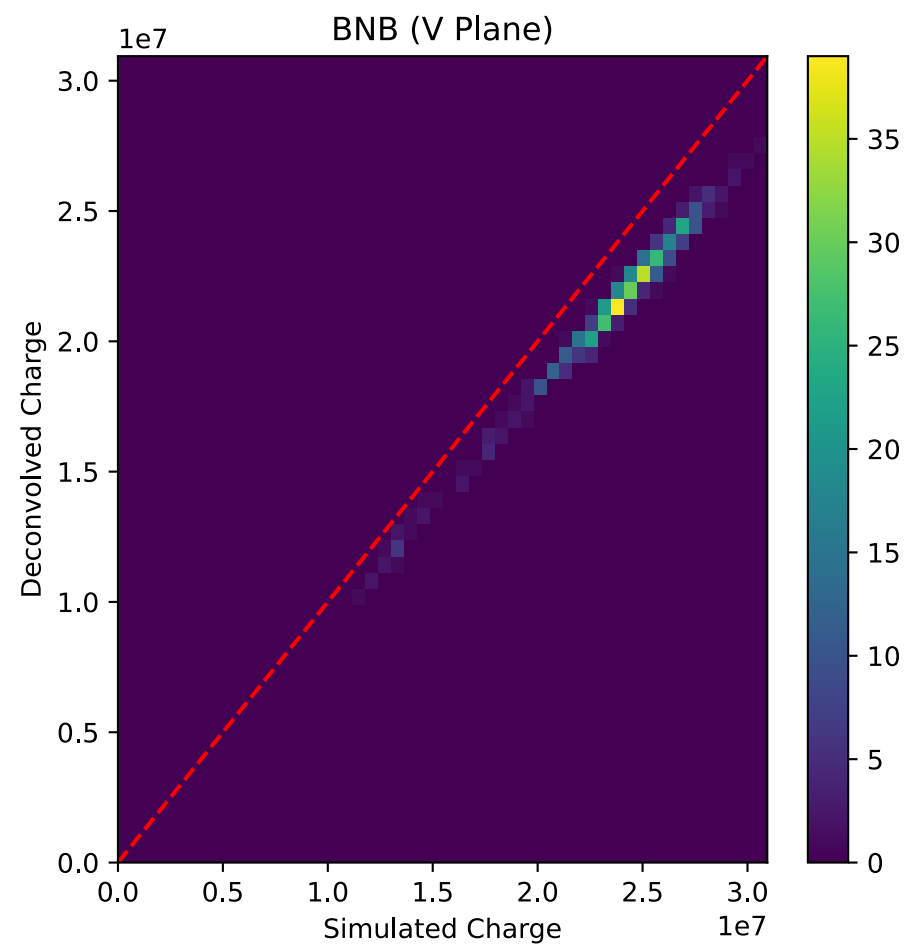
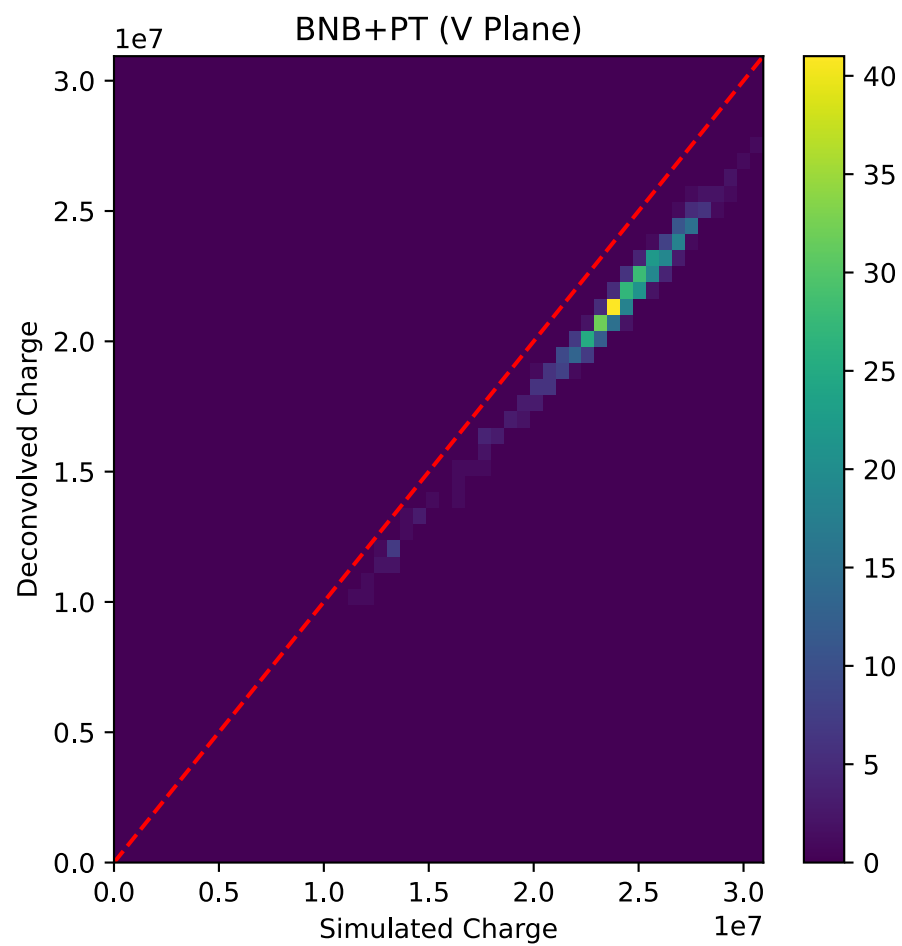
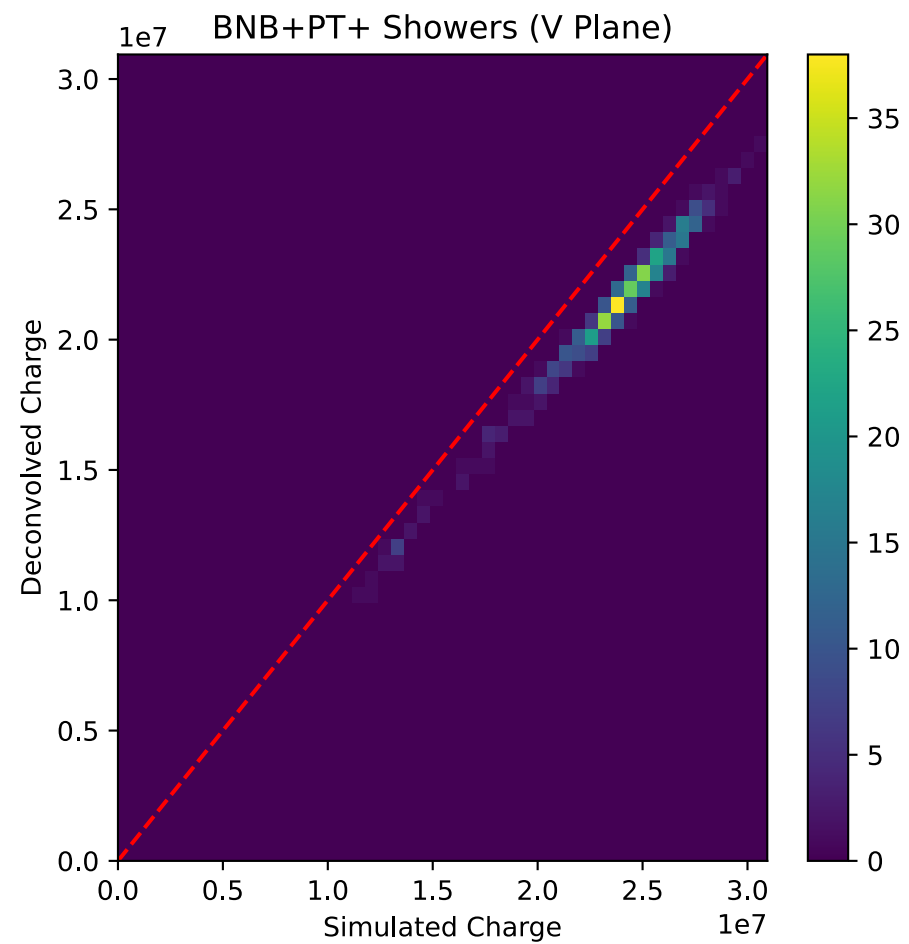
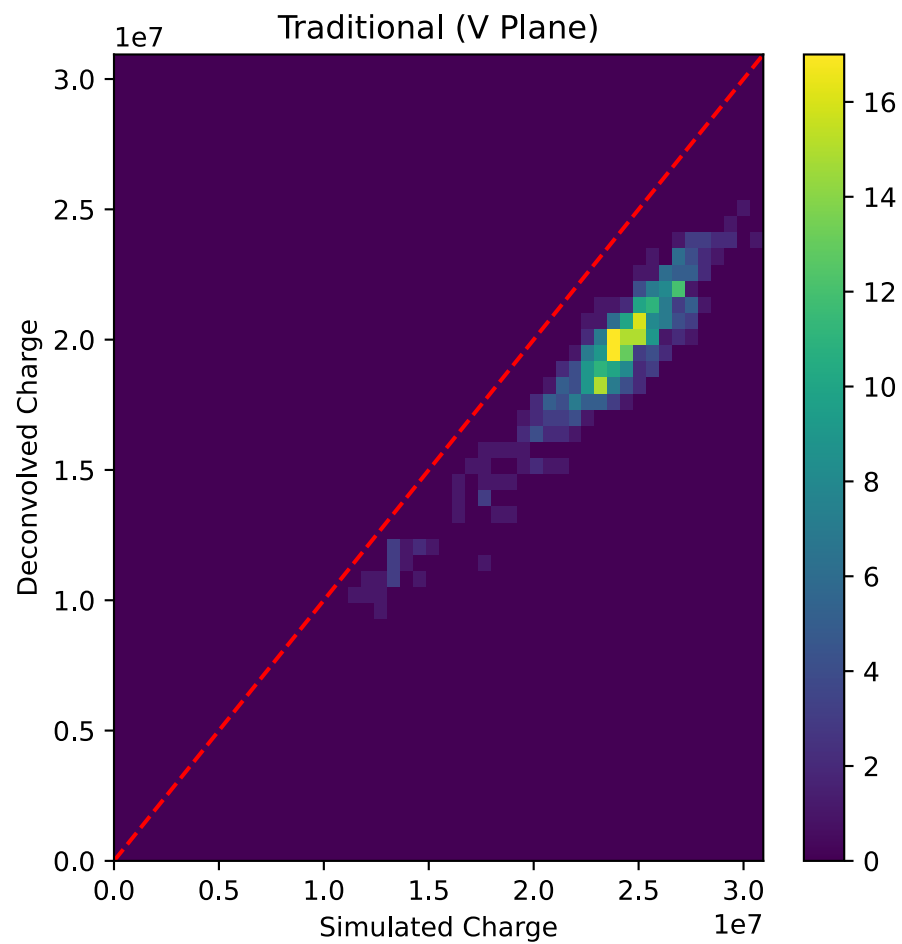
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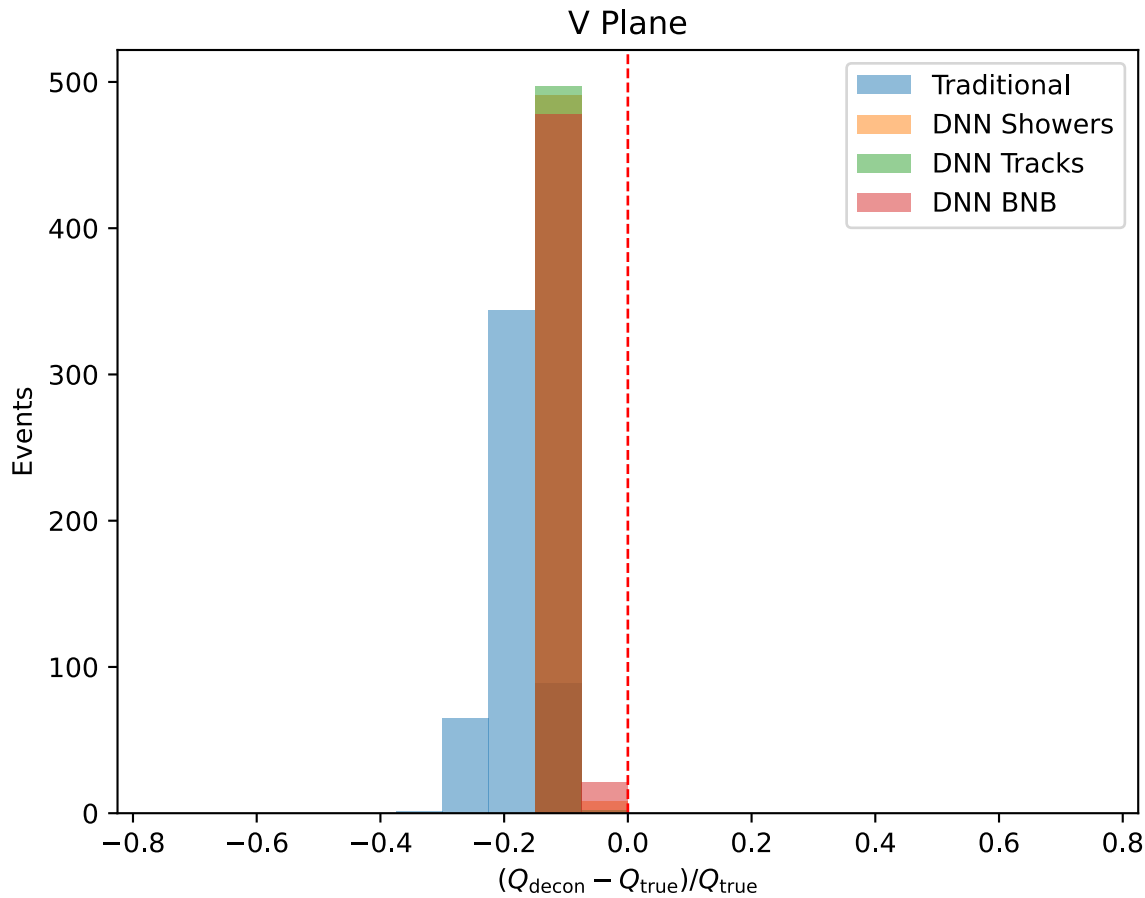
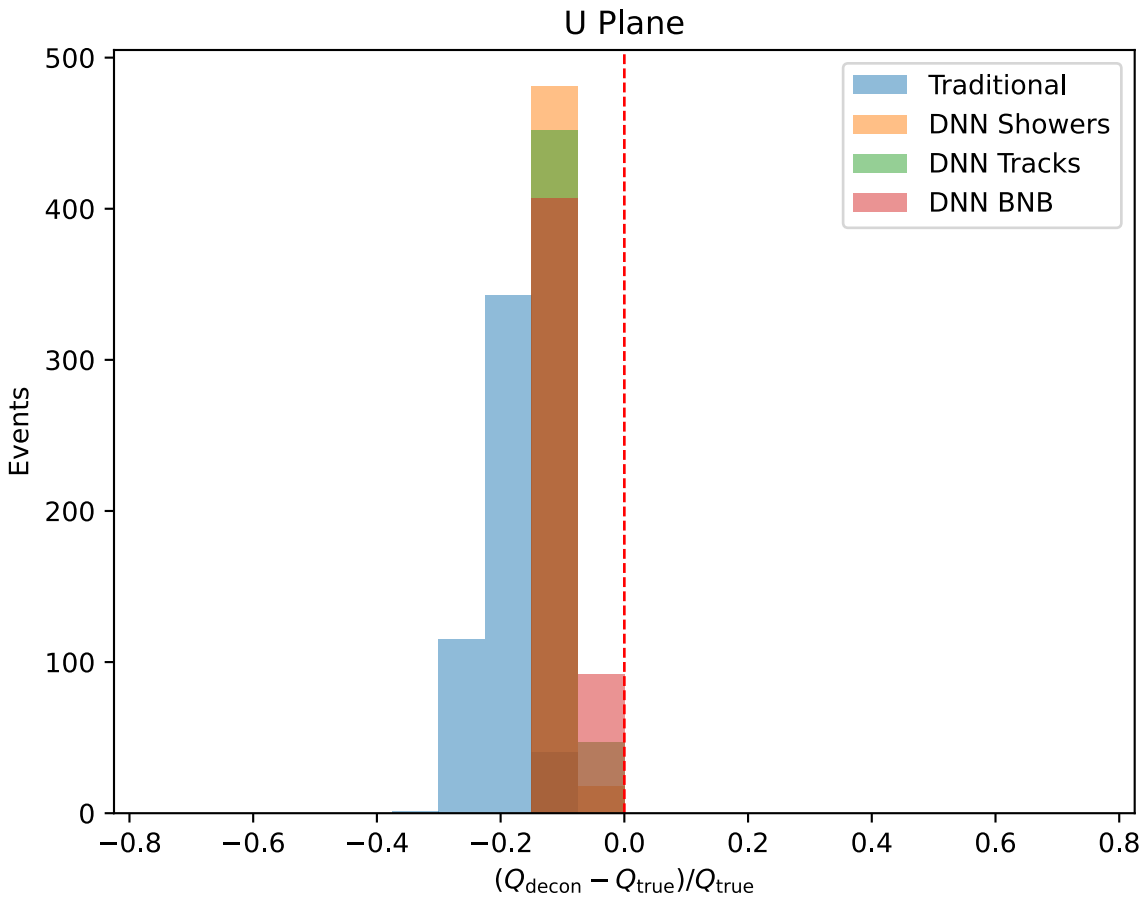
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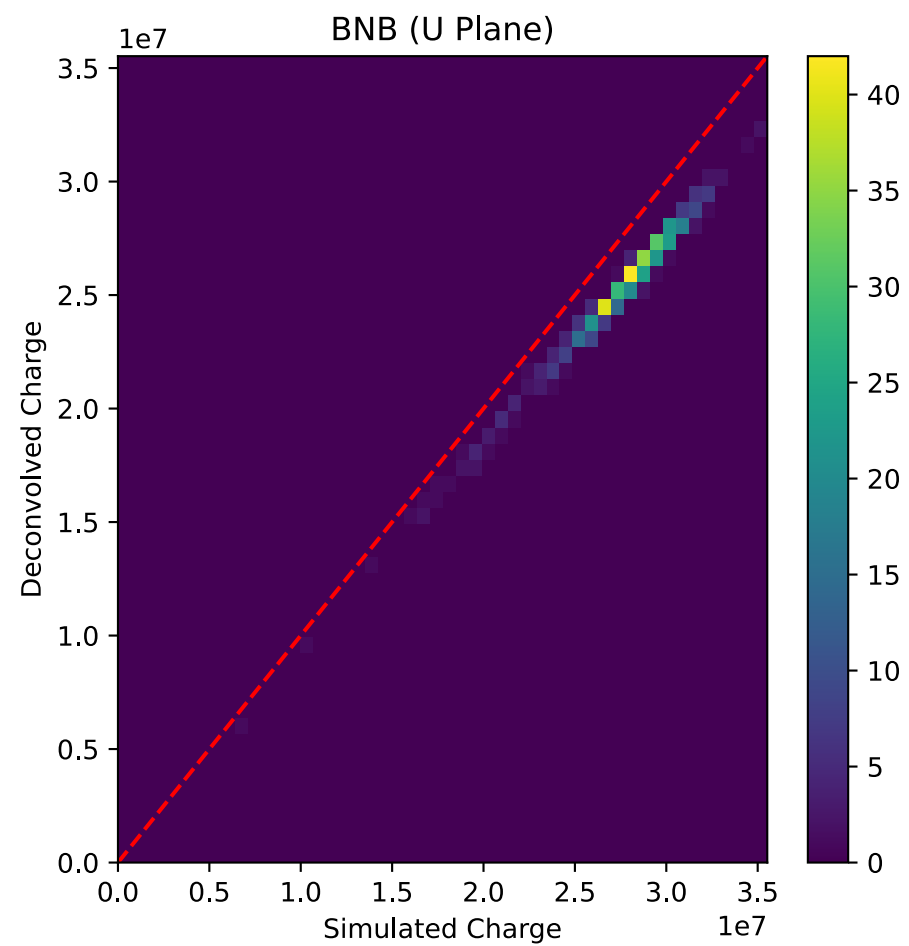
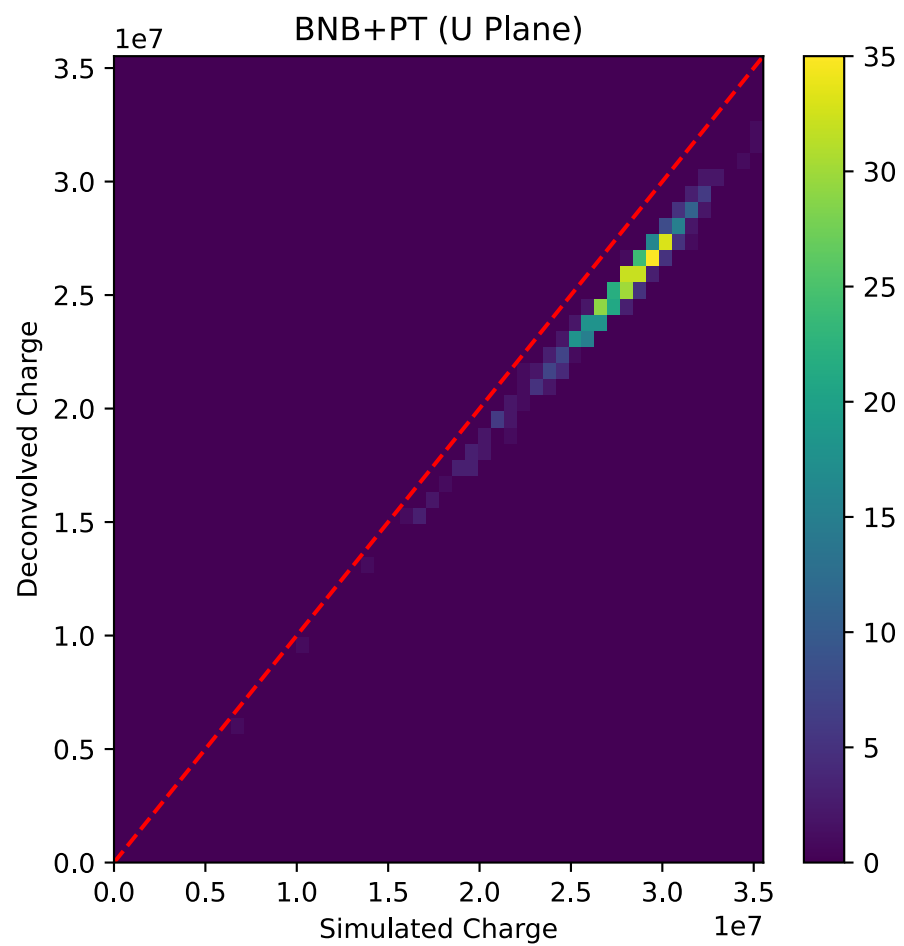
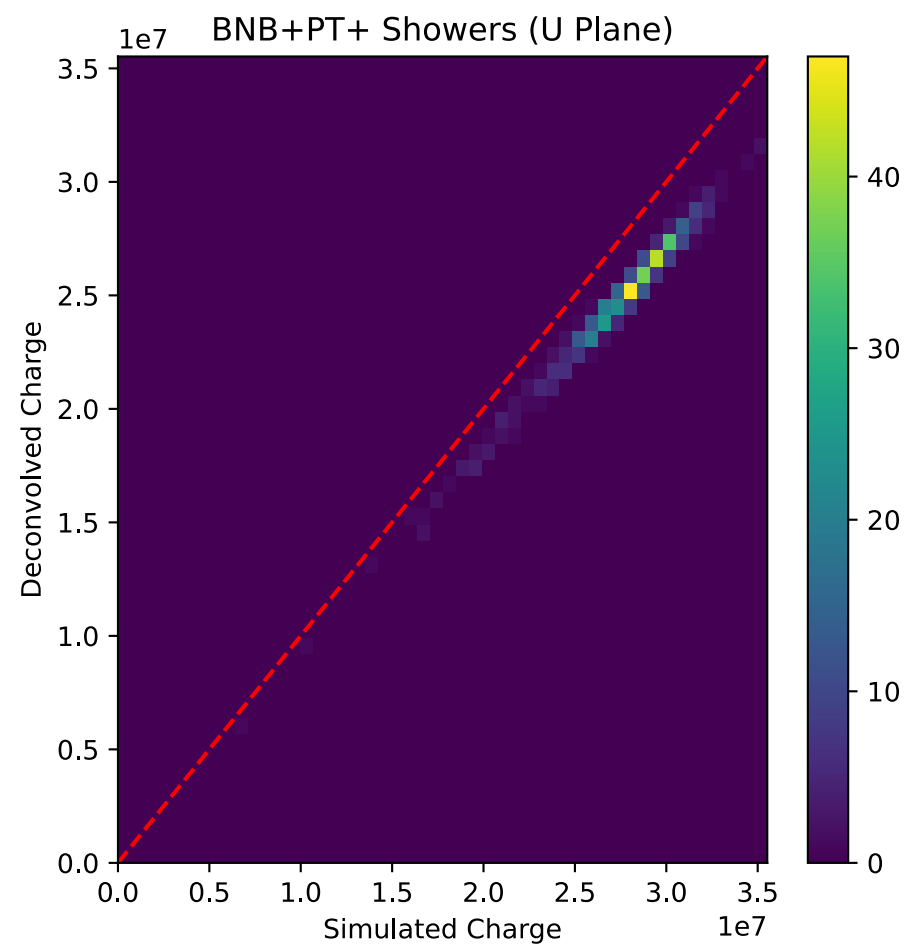
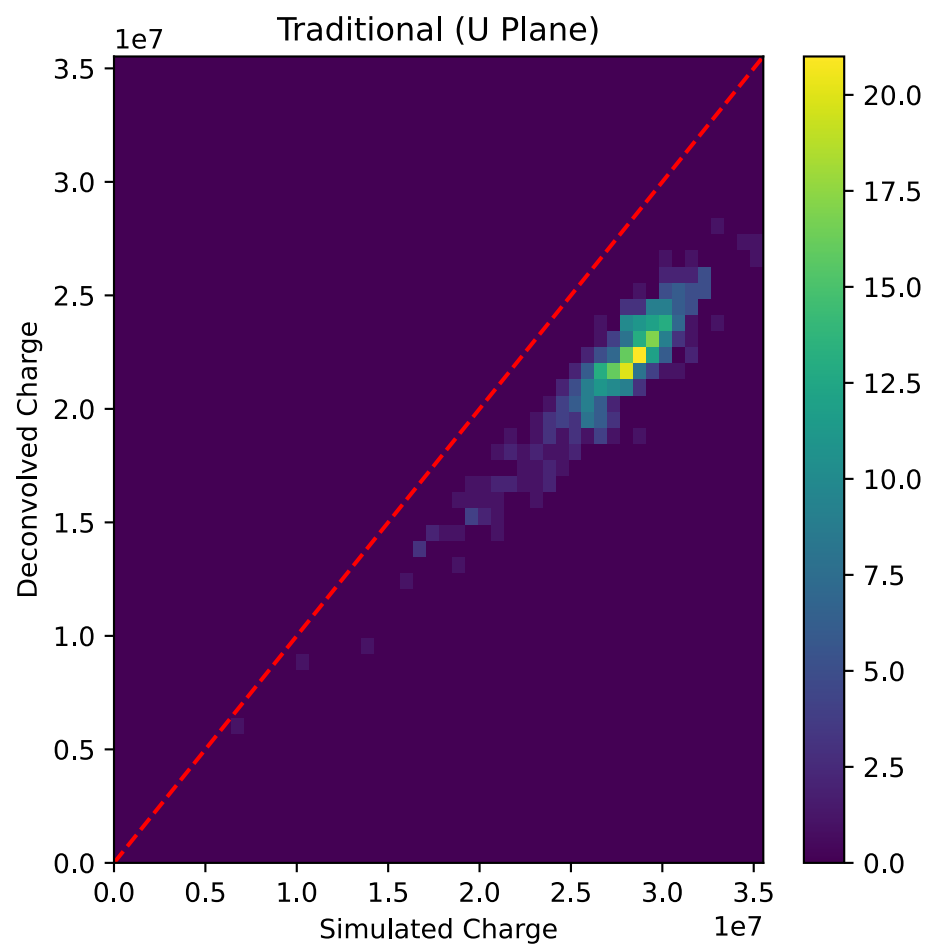
$E_{\text{shower}} \in [0.90 \text{ GeV}, 1.05 \text{ GeV}]$



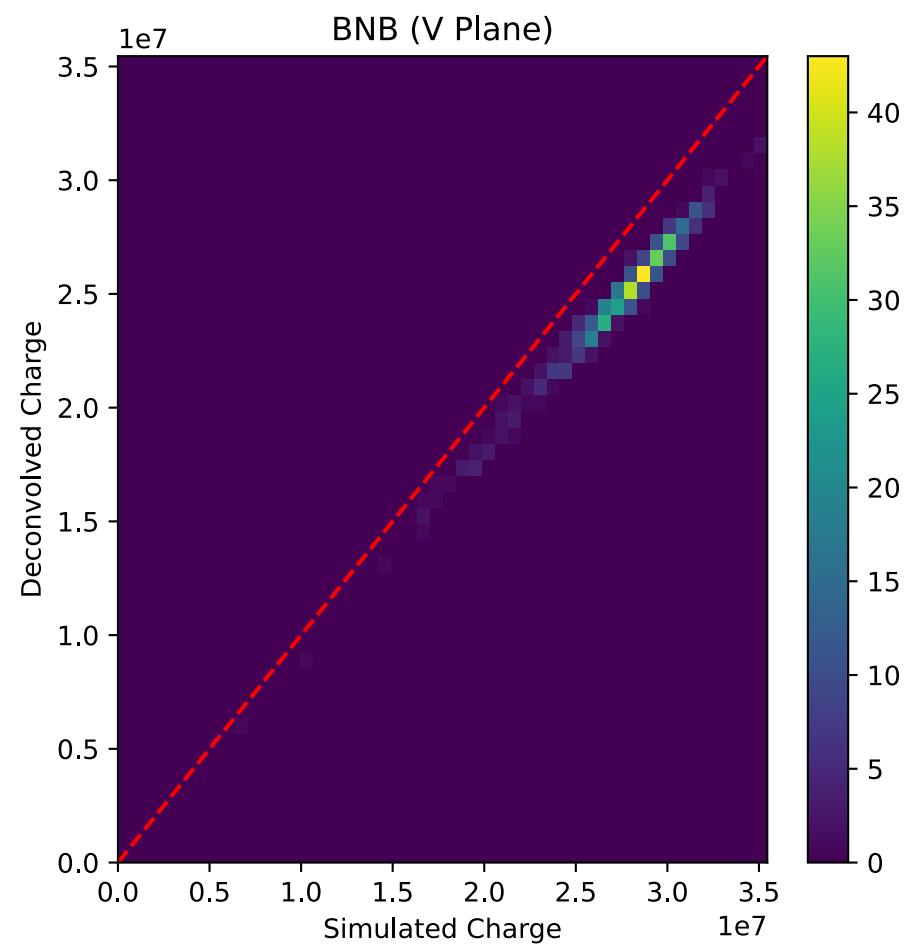
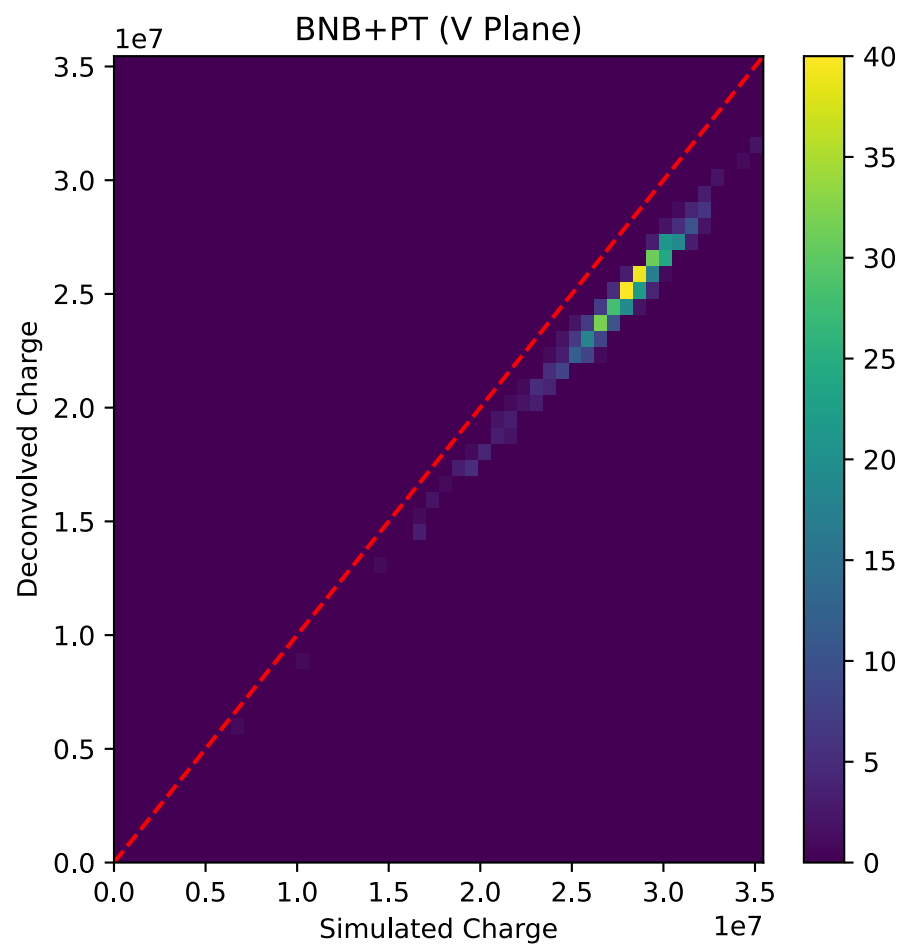
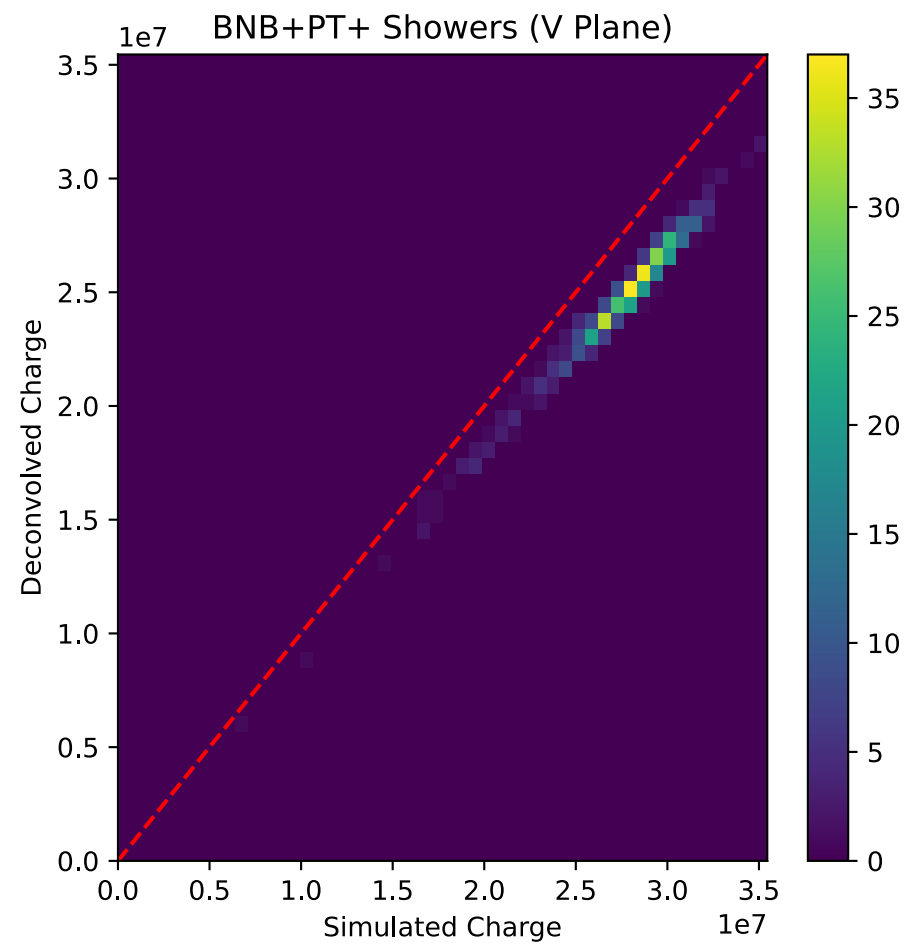
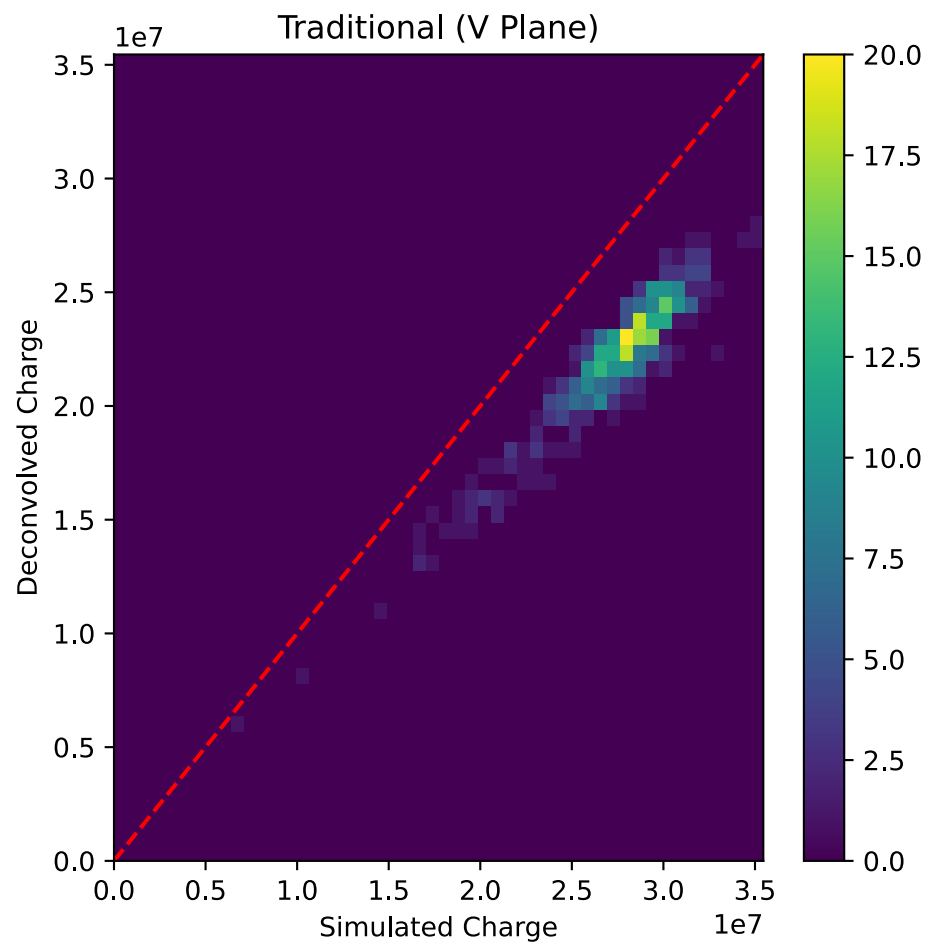
$E_{\text{shower}} \in [0.90 \text{ GeV}, 1.05 \text{ GeV}]$



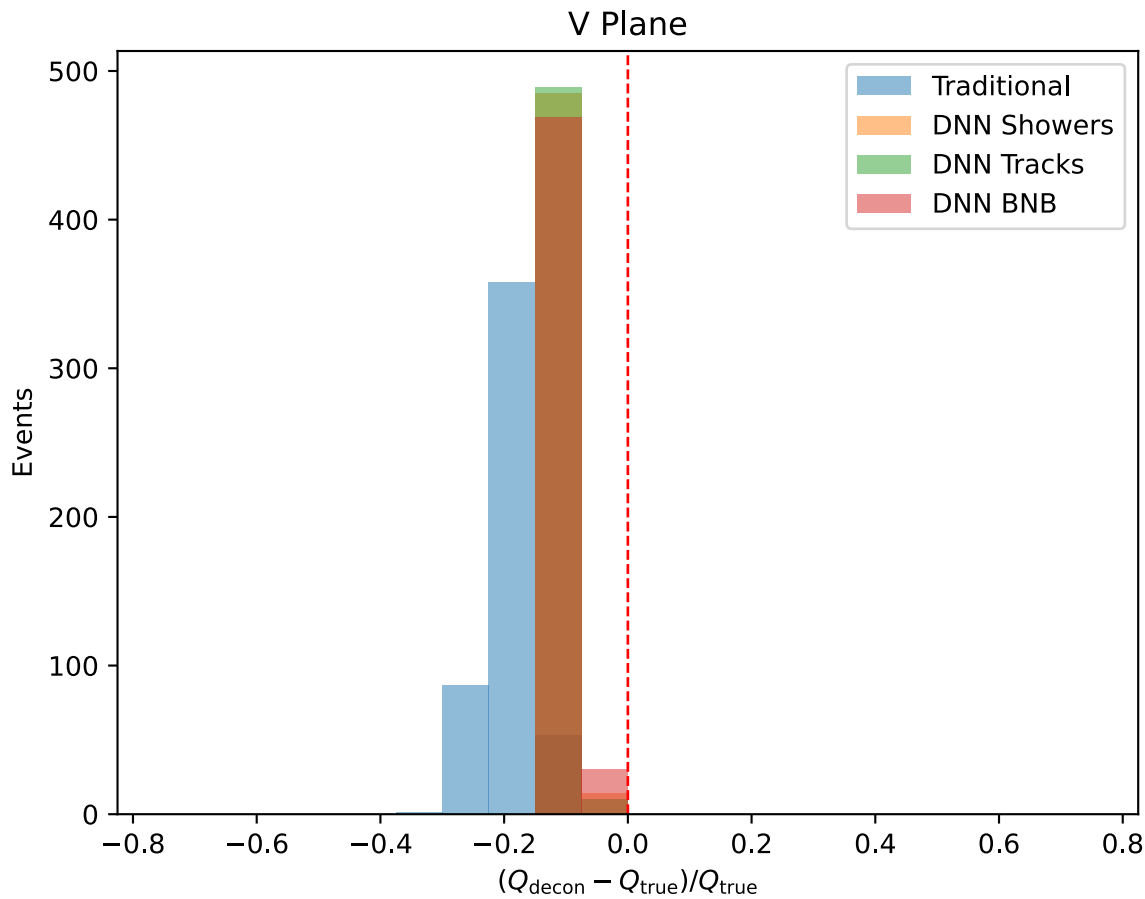
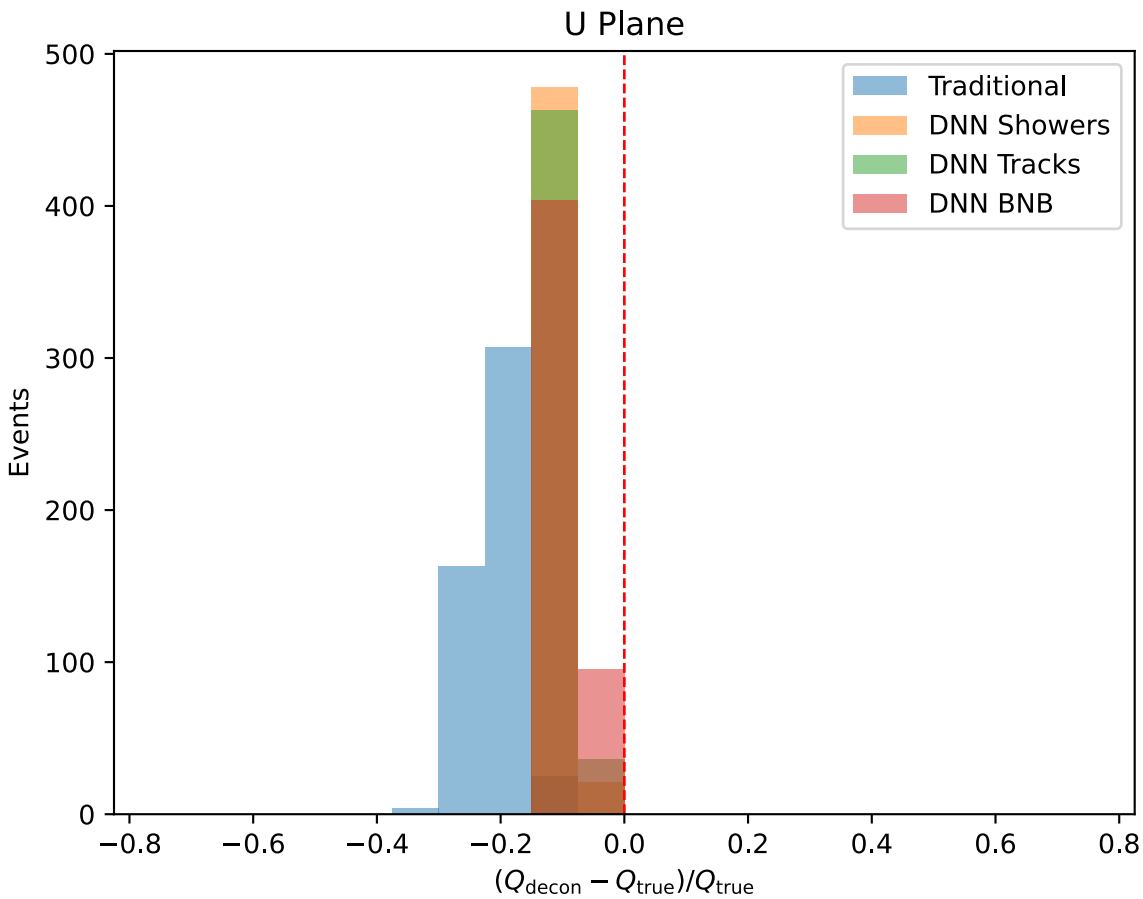
$E_{\text{shower}} \in [1.05 \text{ GeV}, 1.20 \text{ GeV}]$



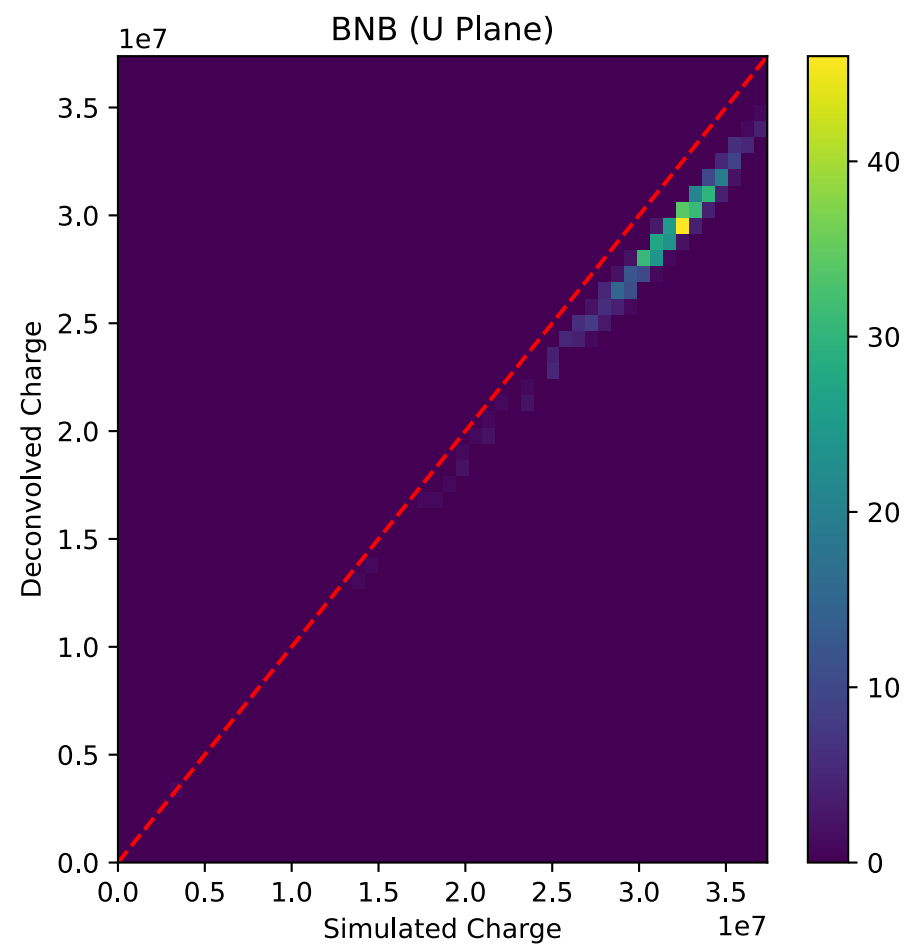
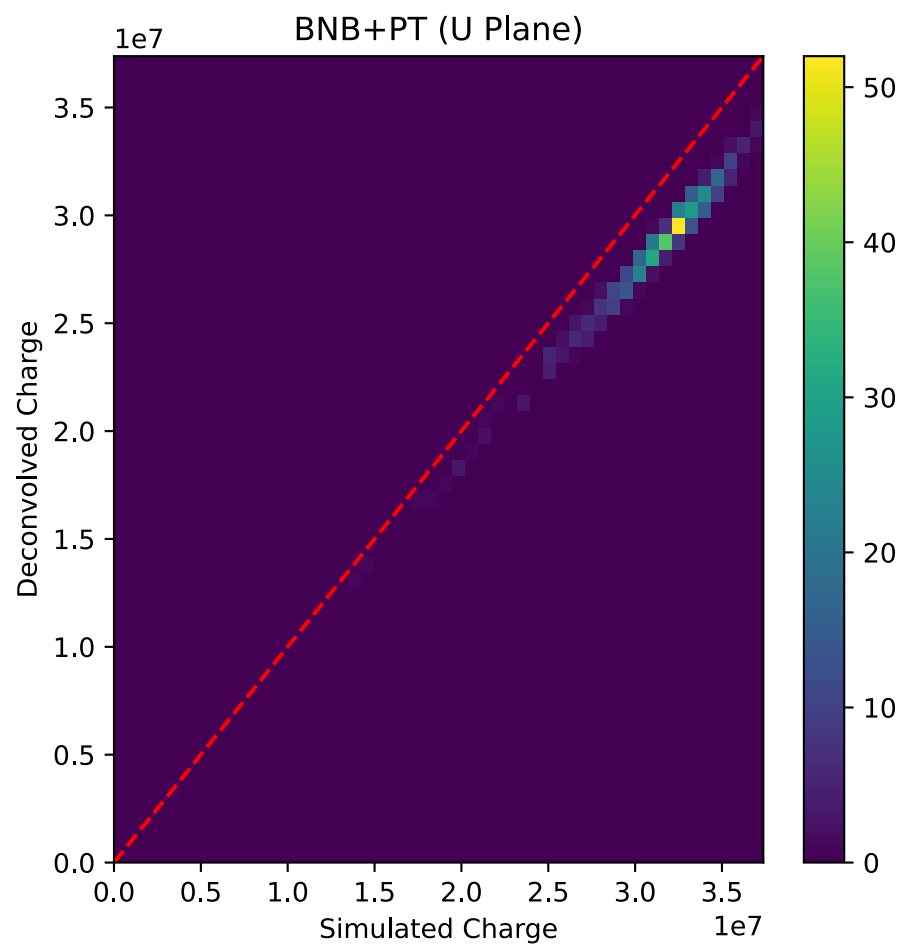
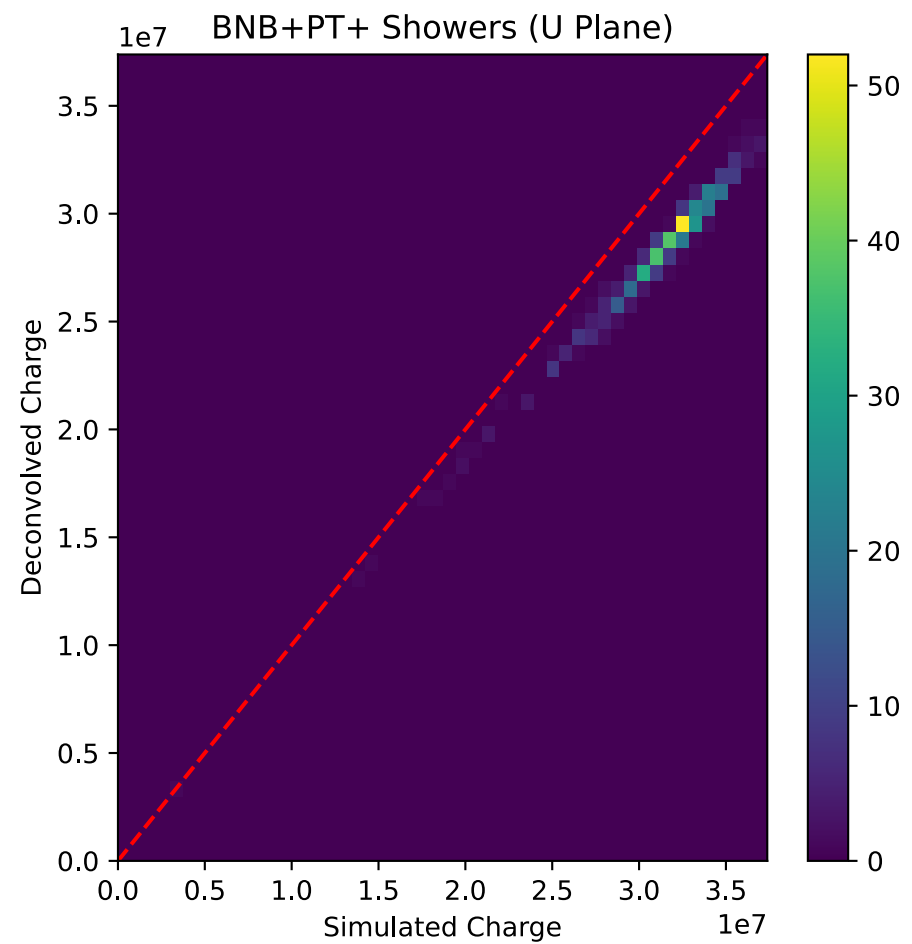
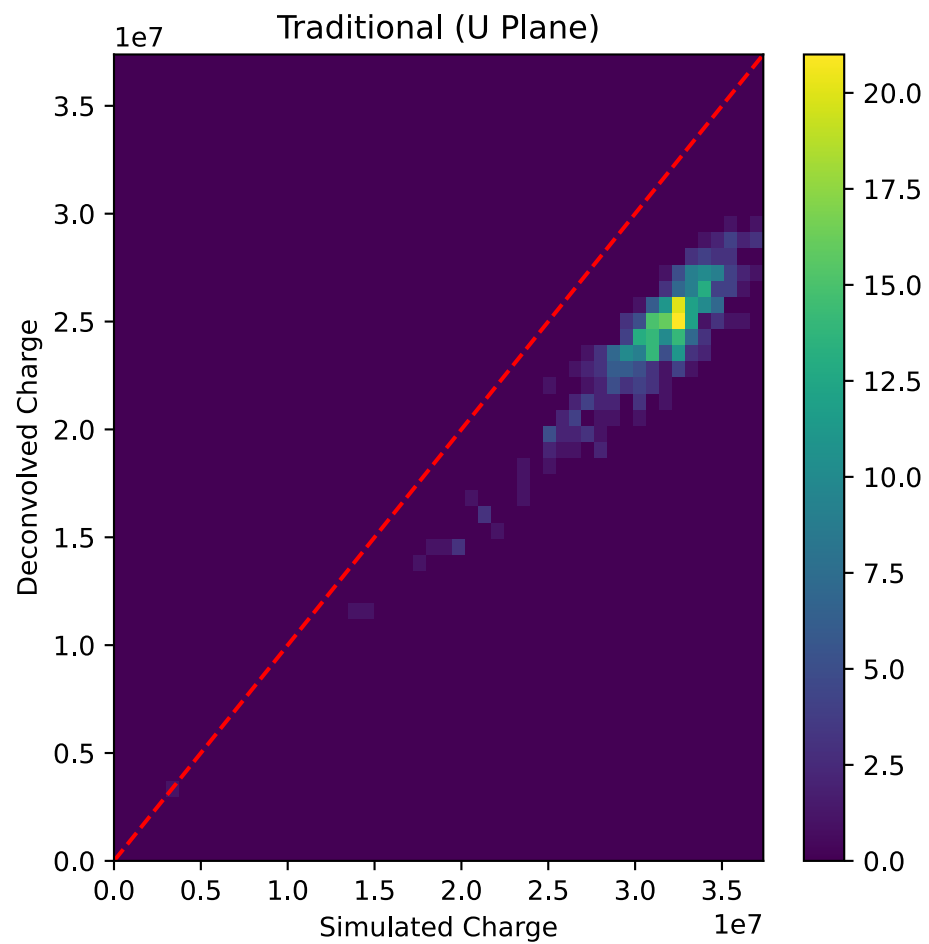
$E_{\text{shower}} \in [1.05 \text{ GeV}, 1.20 \text{ GeV}]$



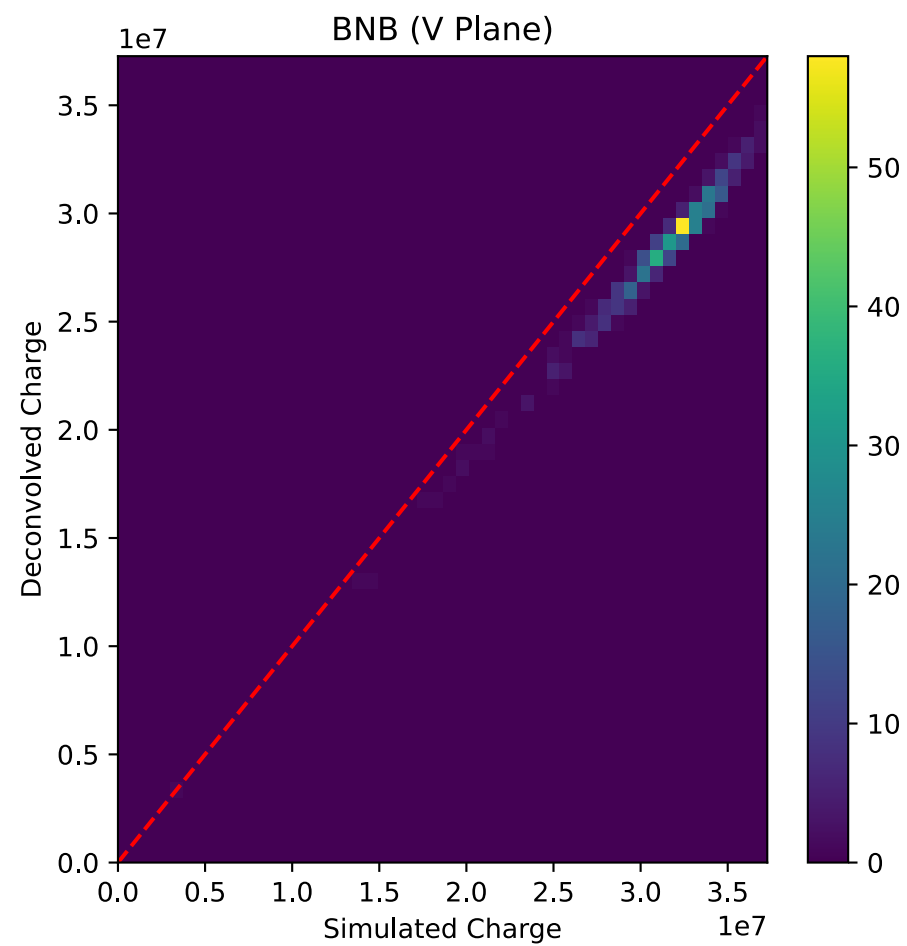
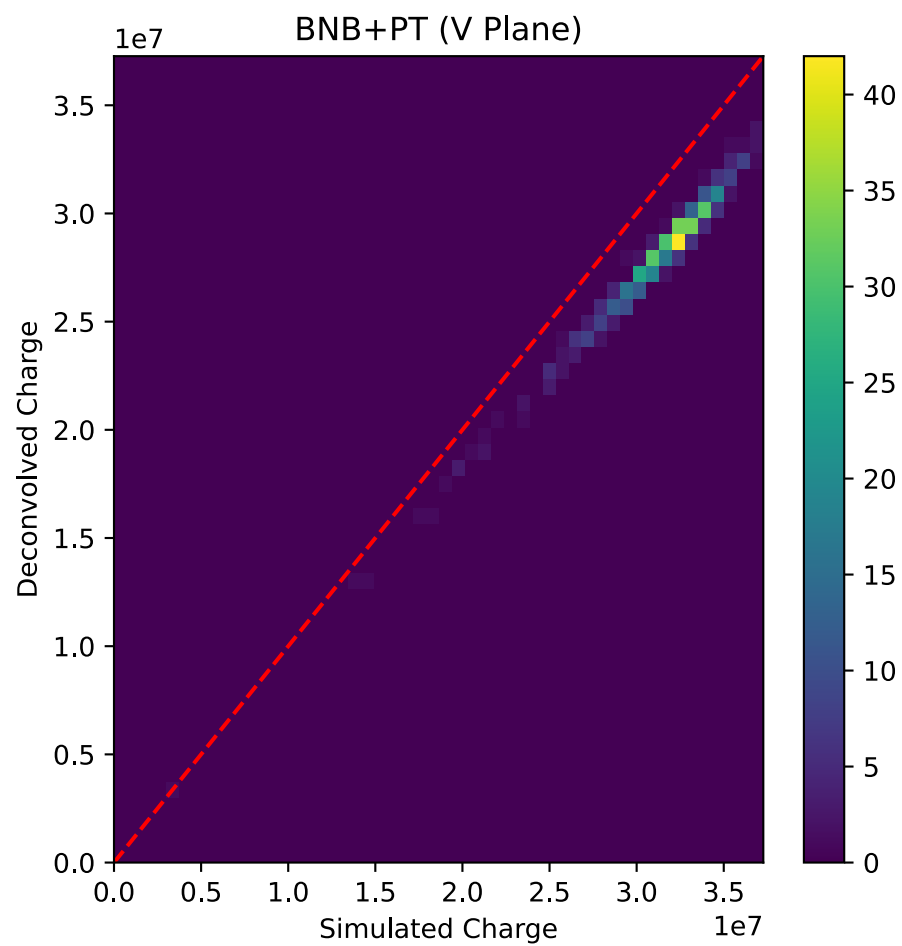
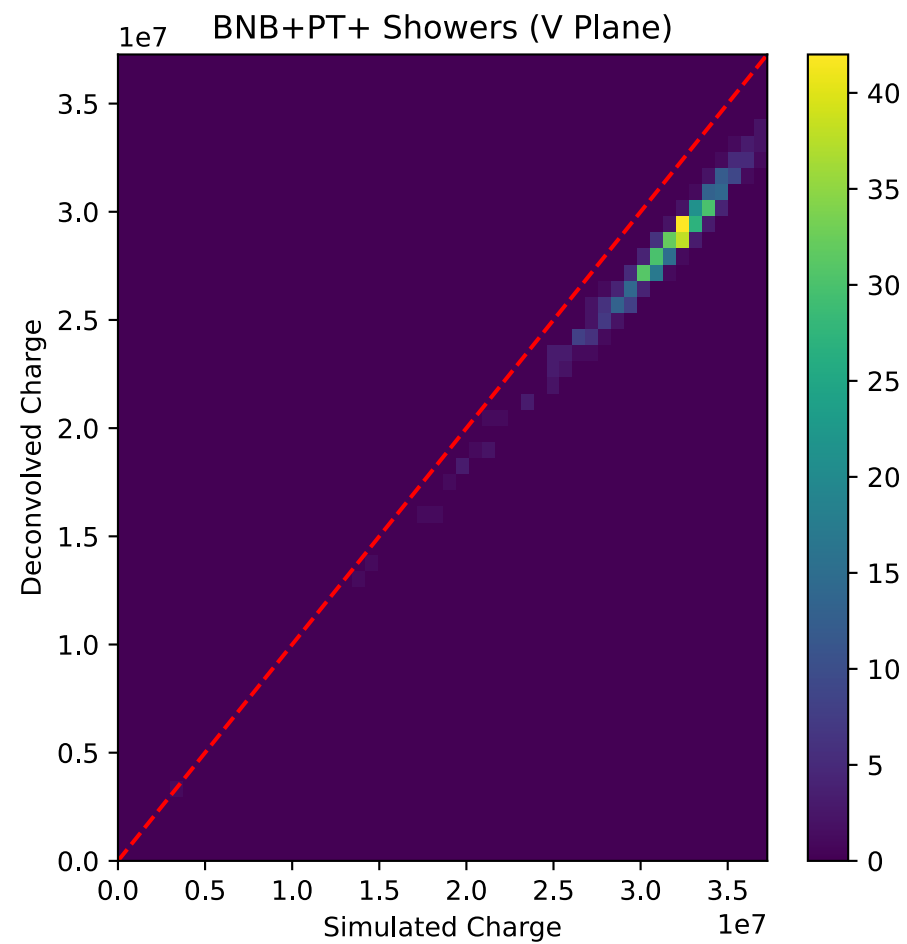
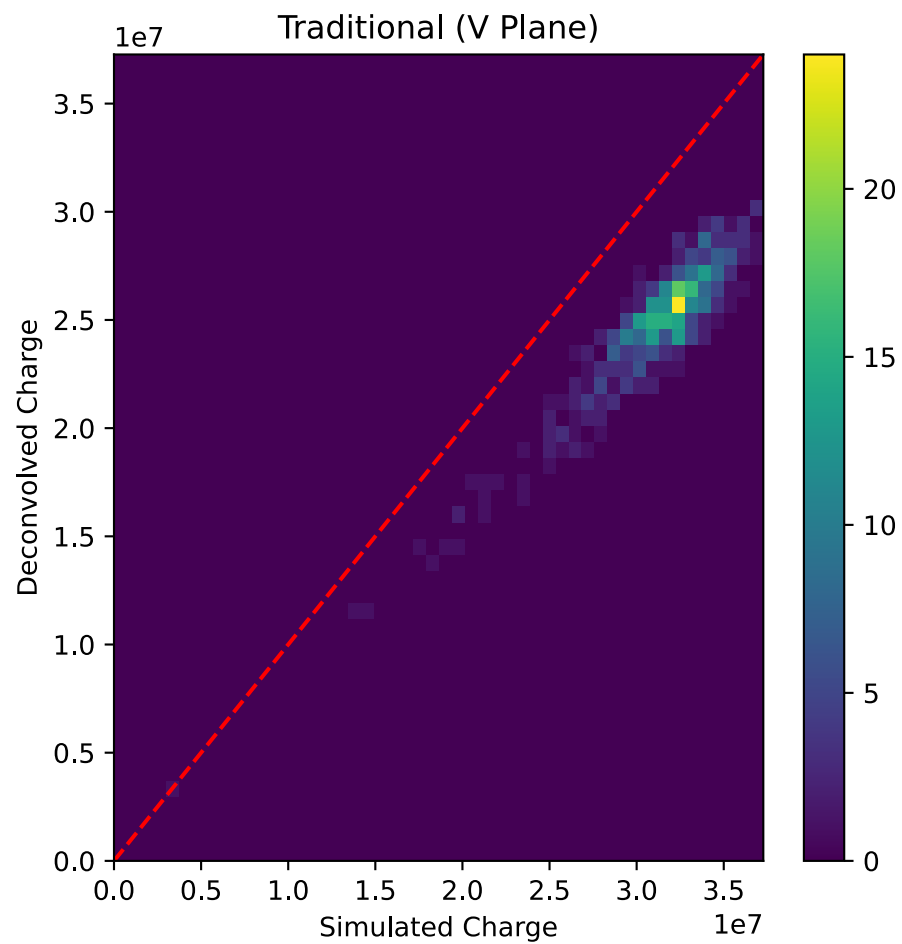
$E_{\text{shower}} \in [1.05 \text{ GeV}, 1.20 \text{ GeV}]$



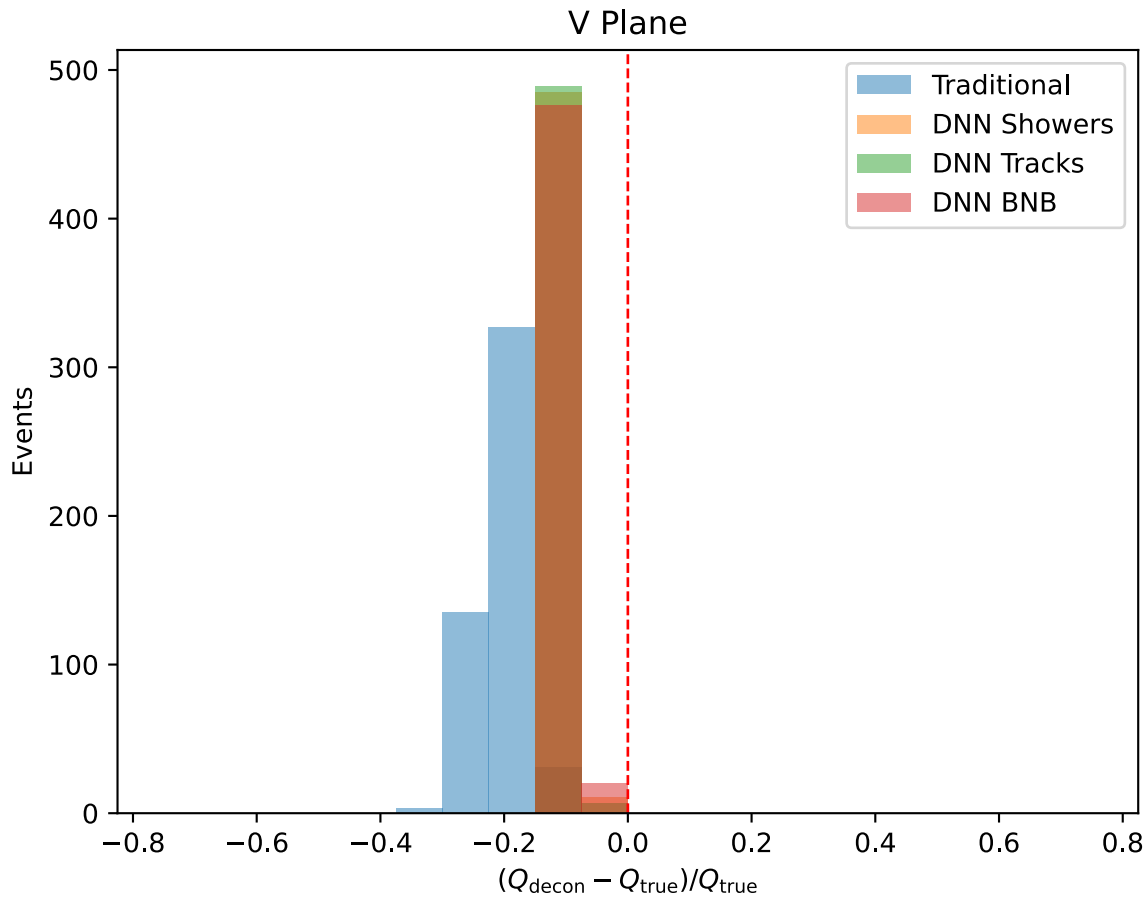
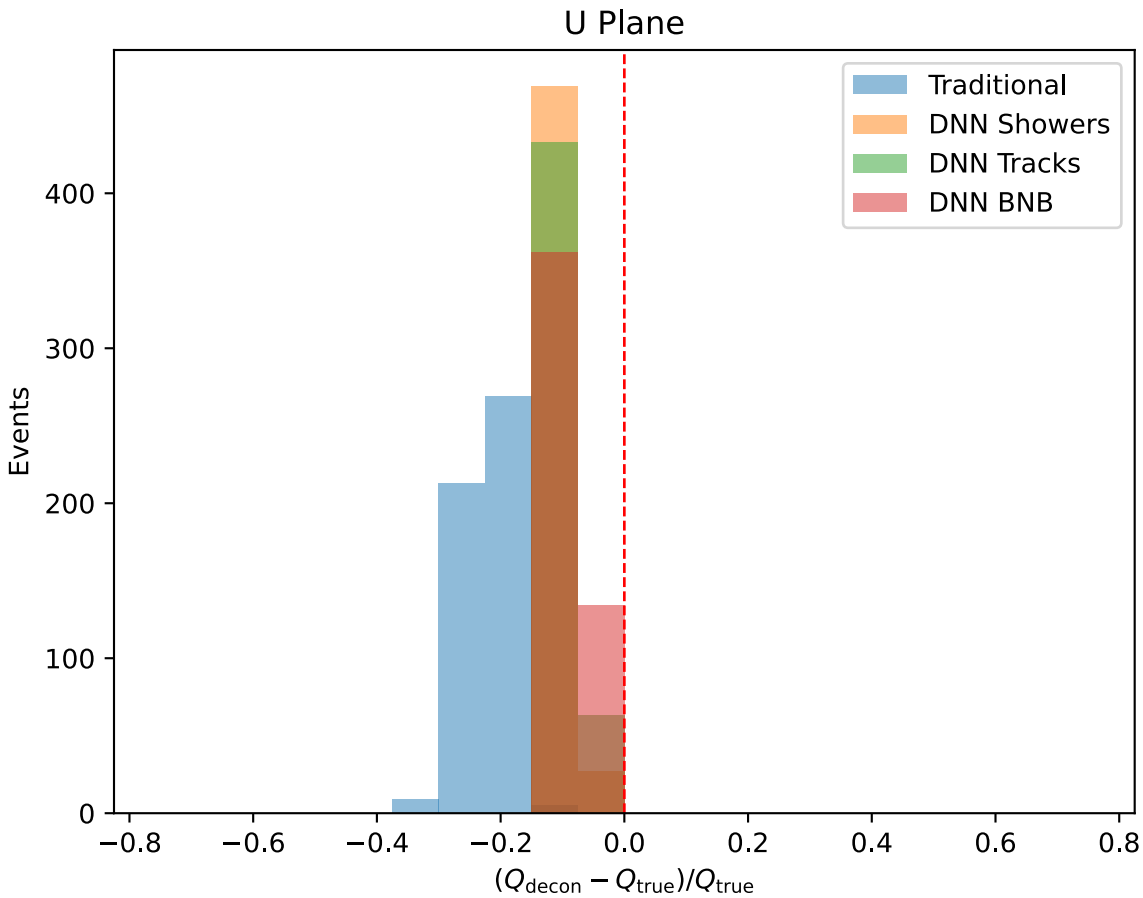
$E_{\text{shower}} \in [1.20 \text{ GeV}, 1.35 \text{ GeV}]$



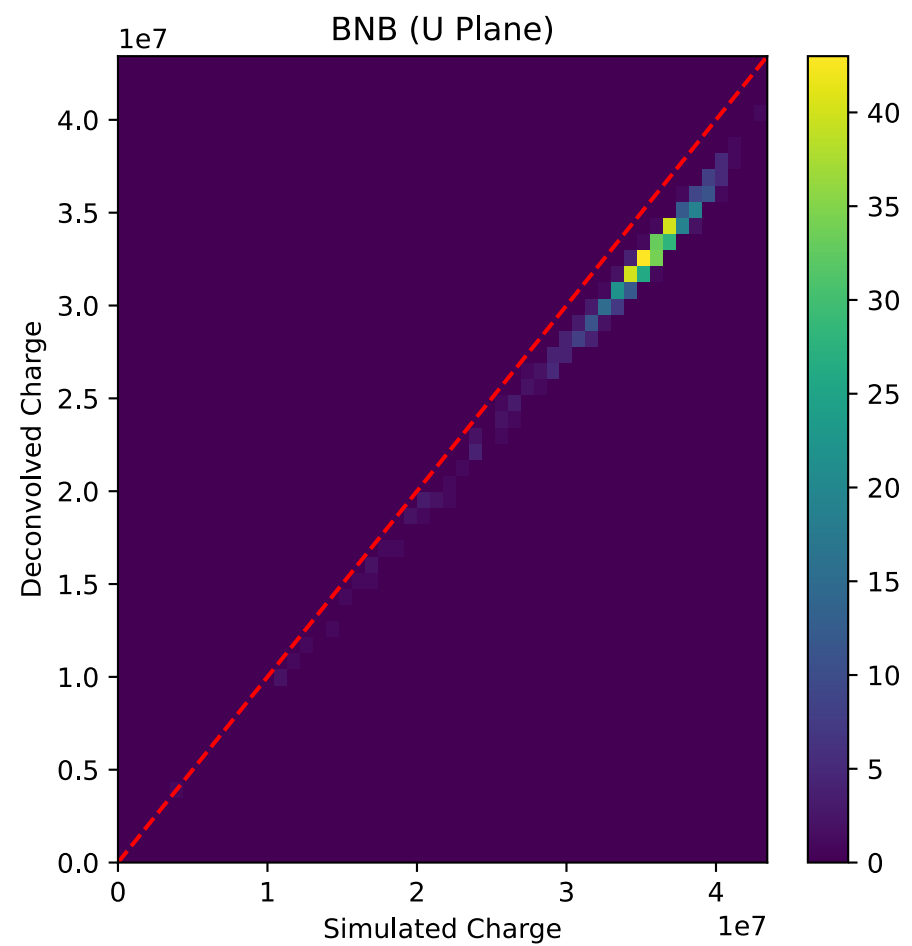
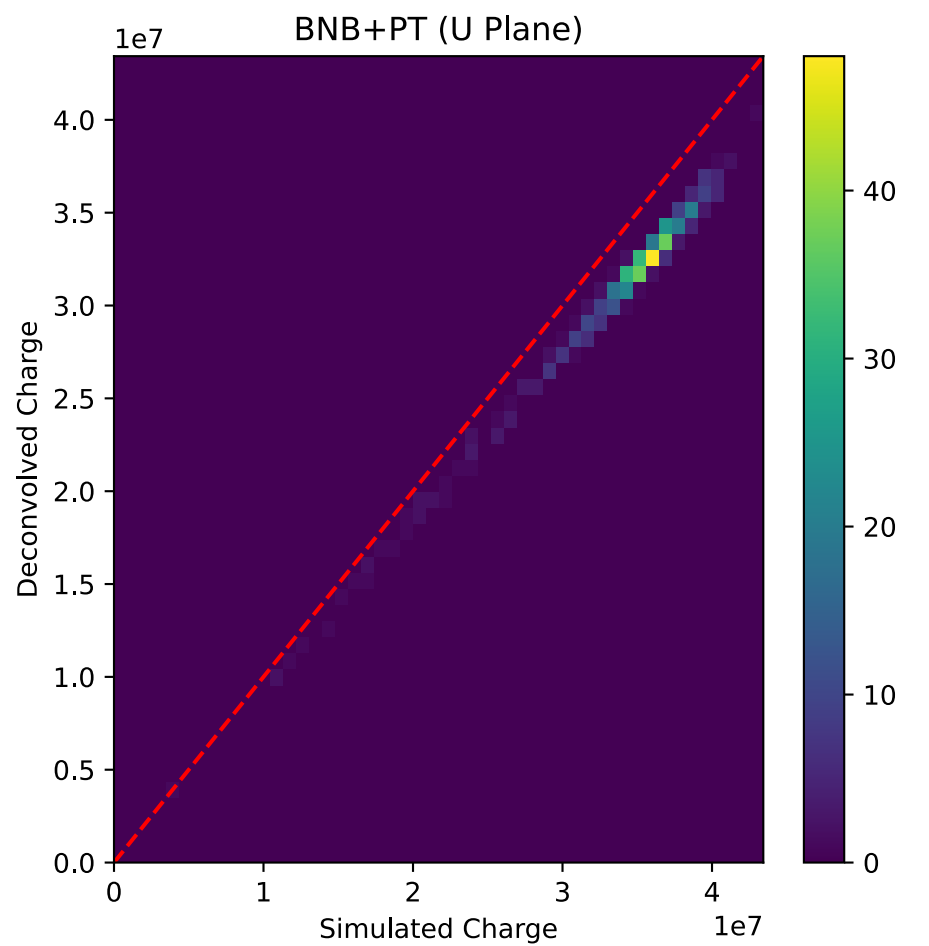
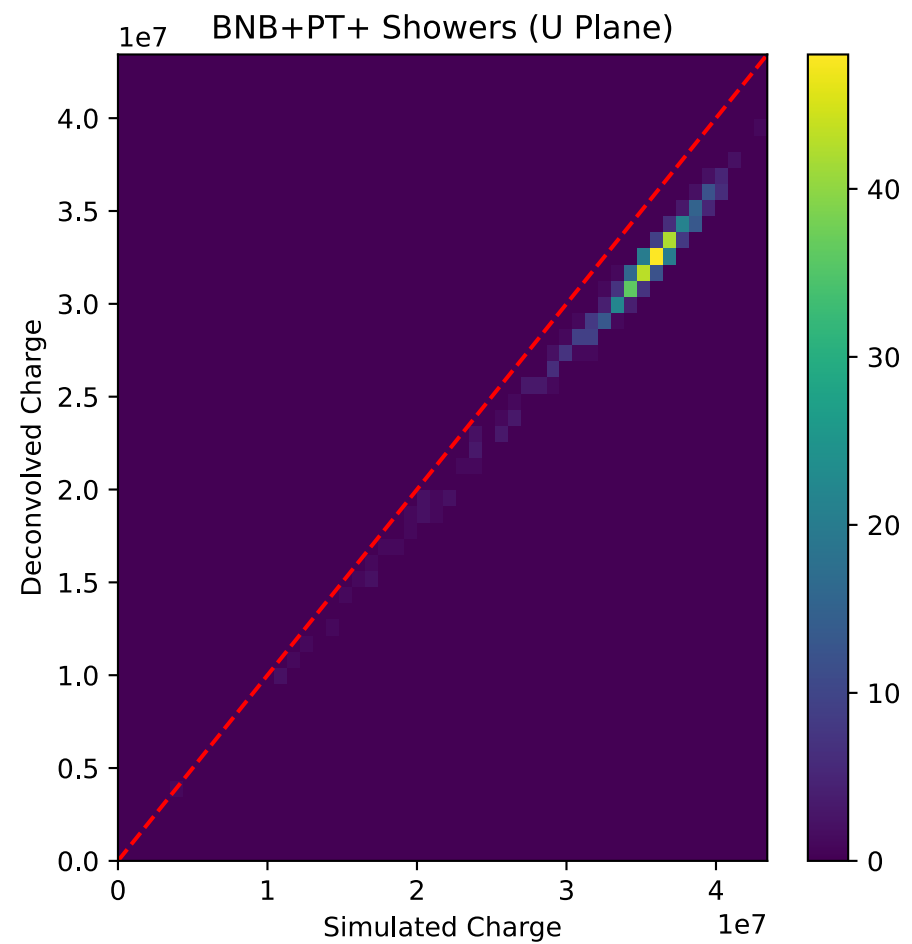
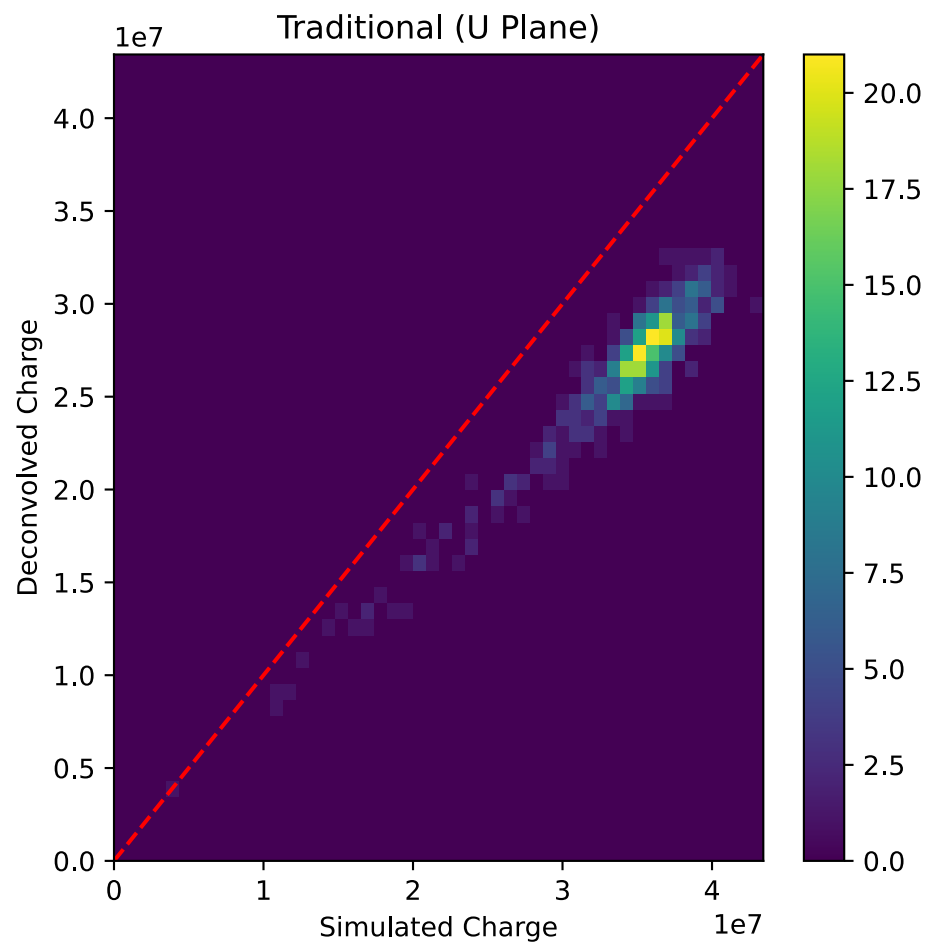
$E_{\text{shower}} \in [1.20 \text{ GeV}, 1.35 \text{ GeV}]$



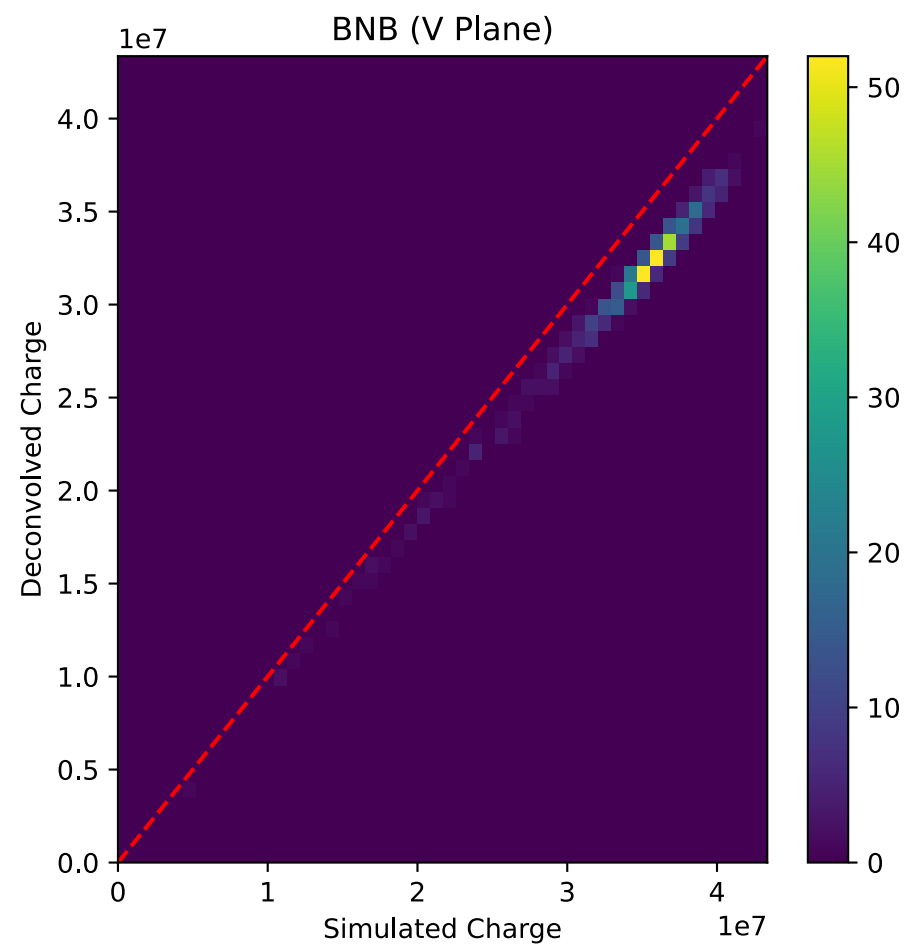
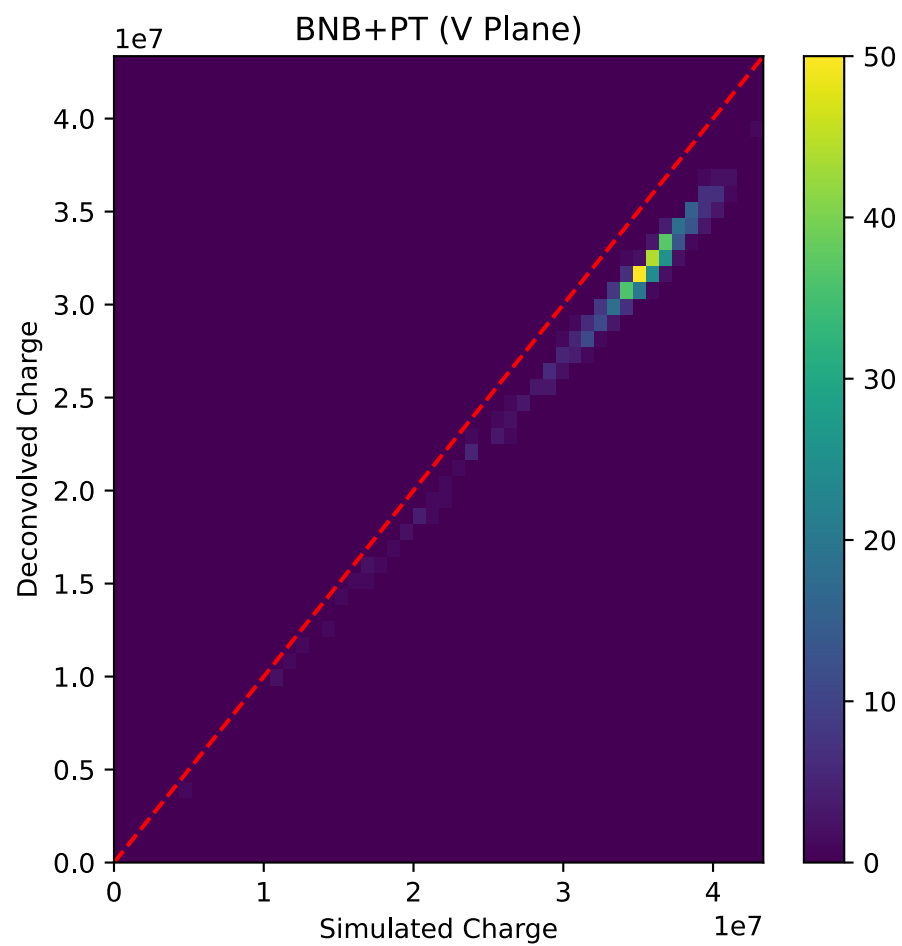
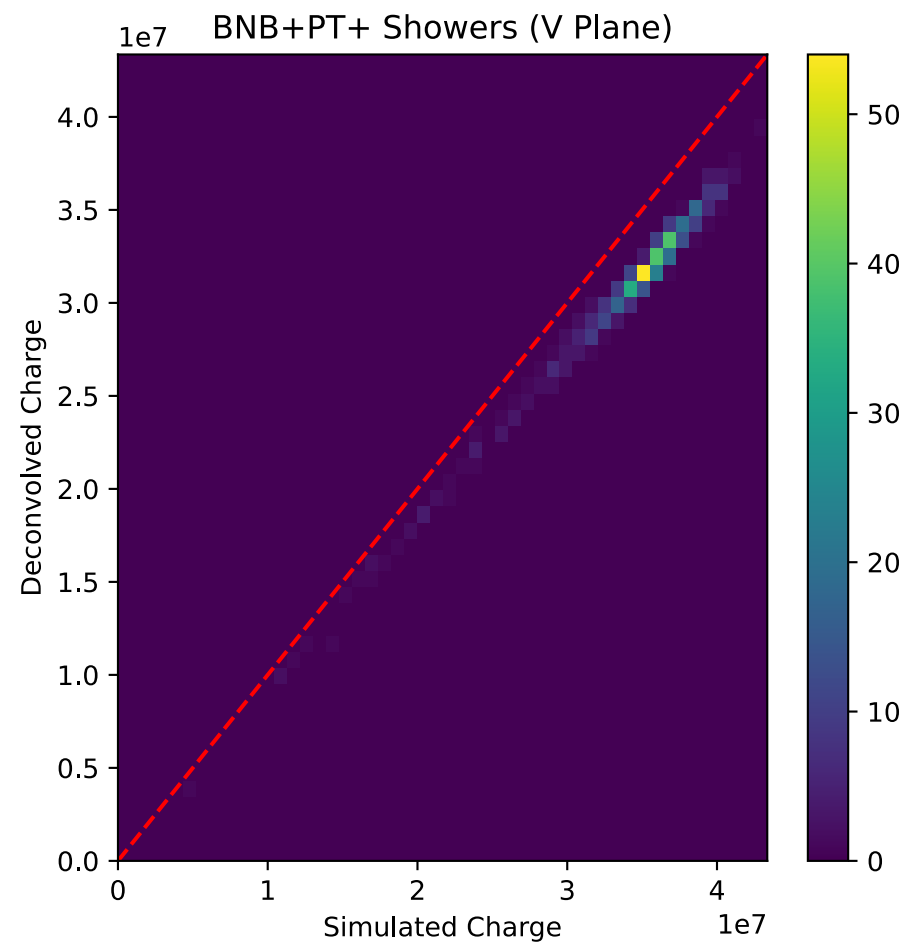
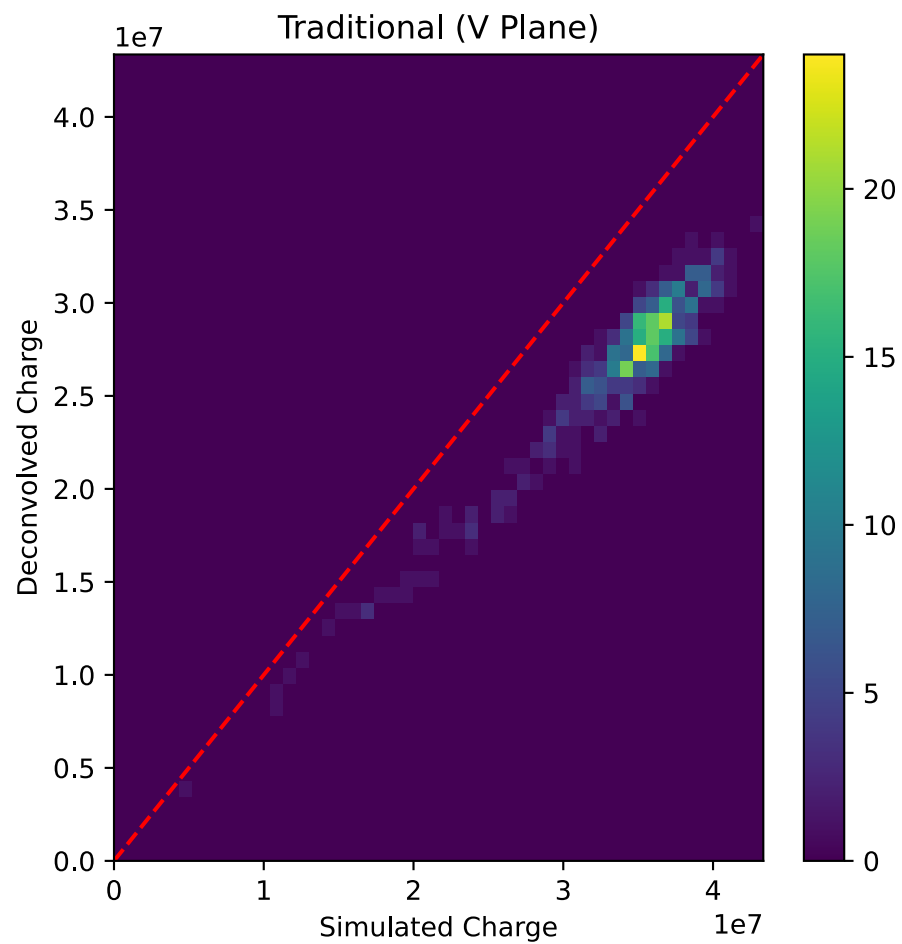
$E_{\text{shower}} \in [1.20 \text{ GeV}, 1.35 \text{ GeV}]$



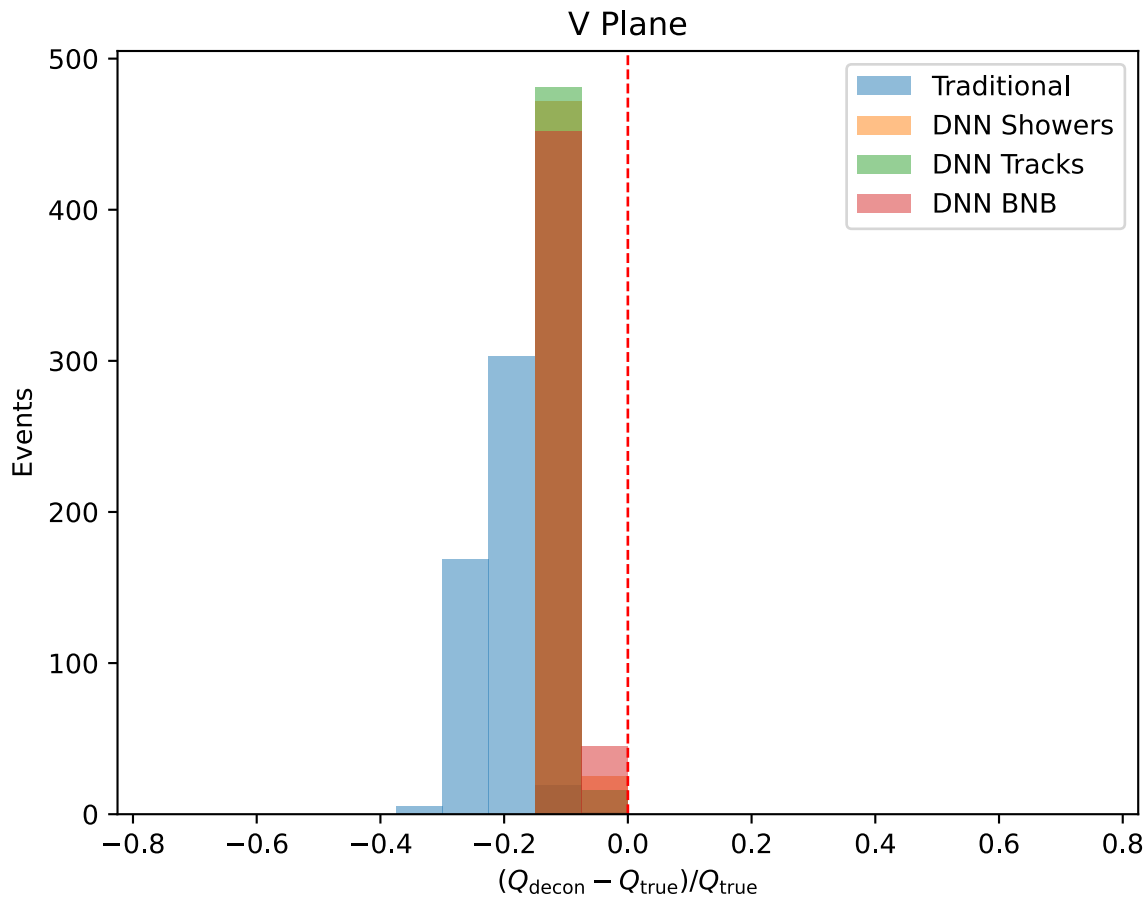
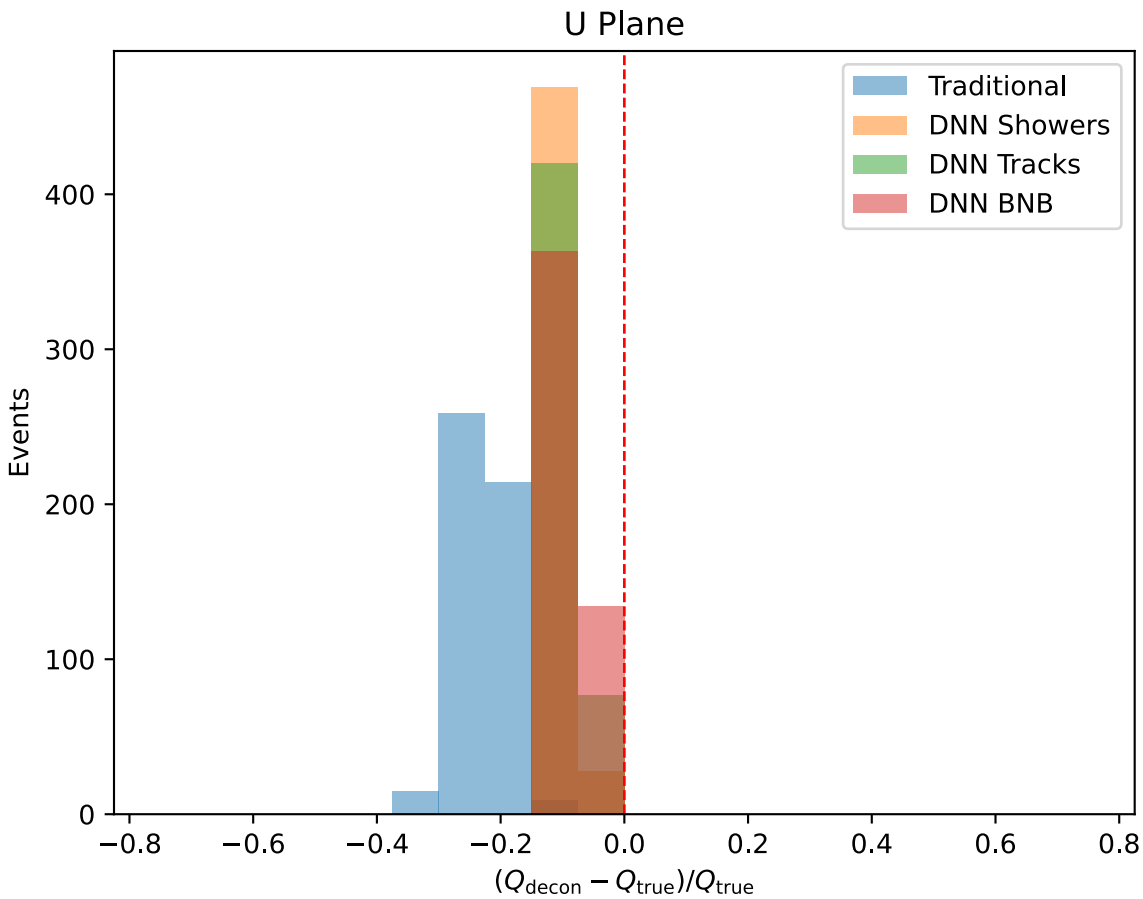
$E_{\text{shower}} \in [1.35 \text{ GeV}, 1.50 \text{ GeV}]$



$E_{\text{shower}} \in [1.35 \text{ GeV}, 1.50 \text{ GeV}]$



$E_{\text{shower}} \in [1.35 \text{ GeV}, 1.50 \text{ GeV}]$



Future Plans

- Evaluate the issues with traditional signal processing output that is jointly produced in the output files with the dnn row output.

WCLSre.....	simtpc2d.....	dnnspp.....	std::vector<recob::Wire>.....	11264
WCLSre.....	simtpc2d.....	simpleSC.....	std::vector<sim::SimChannel>.....	11264
WCLSre.....	simtpc2d.....	wiener.....	std::vector<recob::Wire>.....	11264
WCLSre.....	TriggerResults	art::TriggerResults.....1
WCLSre.....	simtpc2d.....	gauss.....	std::vector<recob::Wire>.....	11264

- Generate showers with BNB energy distribution and expected BNB angle distribution.
- Perform charge extraction again.