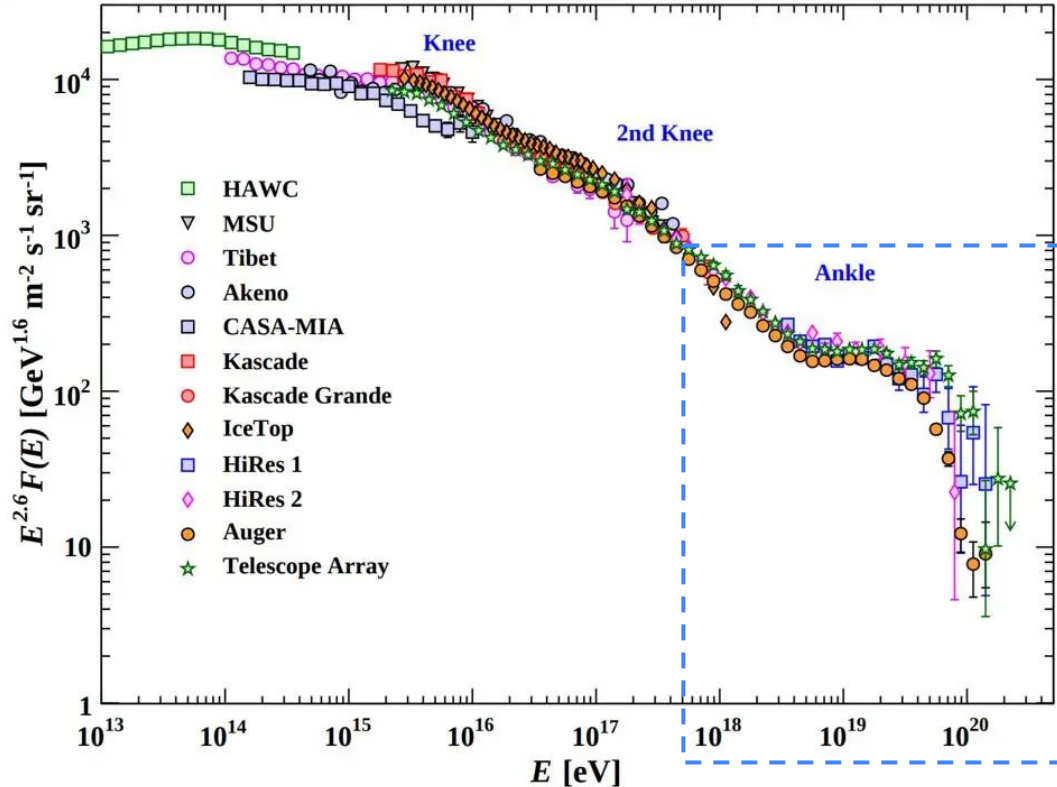


# Energy and mass spectra of TA SD



Novosibirsk, 03.26

# Cosmic Rays Spectrum



Mass and energy spectra of ultra high energy cosmic rays ( $E > 10^{18}$  eV) is subject to active discussions

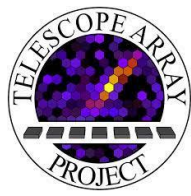
M. Tanabashi et al. (Particle Data Group),  
Phys. Rev. D, 2019

# Outline

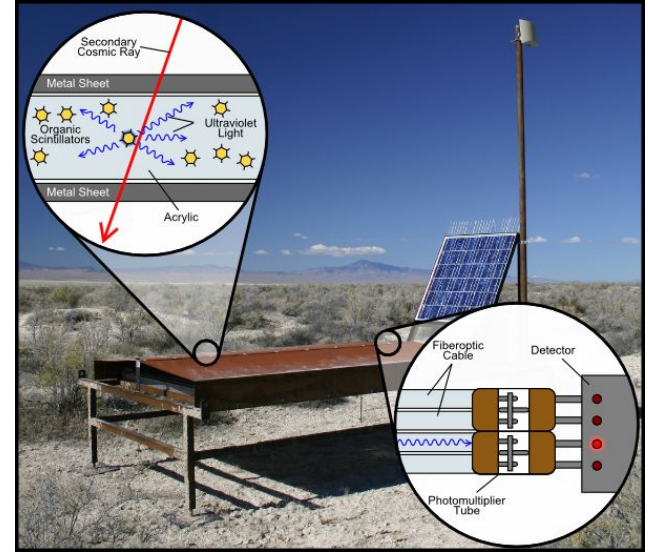
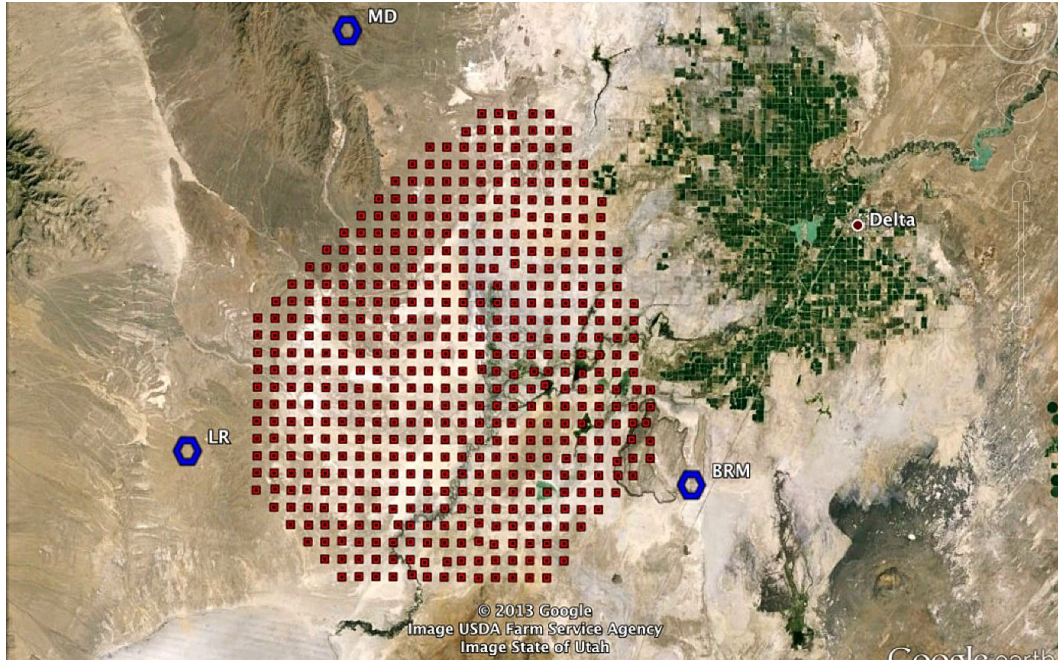
- Telescope Array experiment
- Neural network architecture
- Reliability of the neural network
- Mass-energy spectrum

# Telescope Array experiment

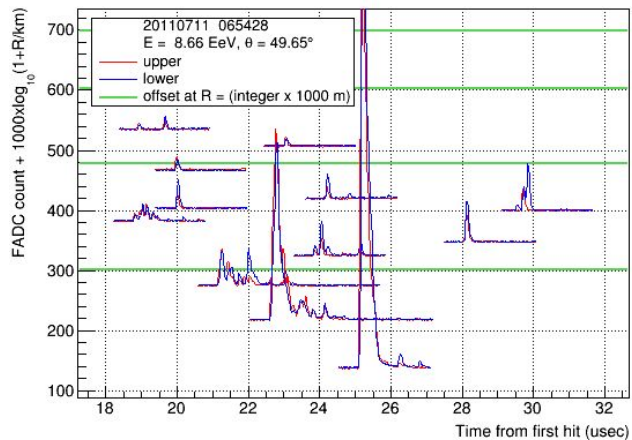
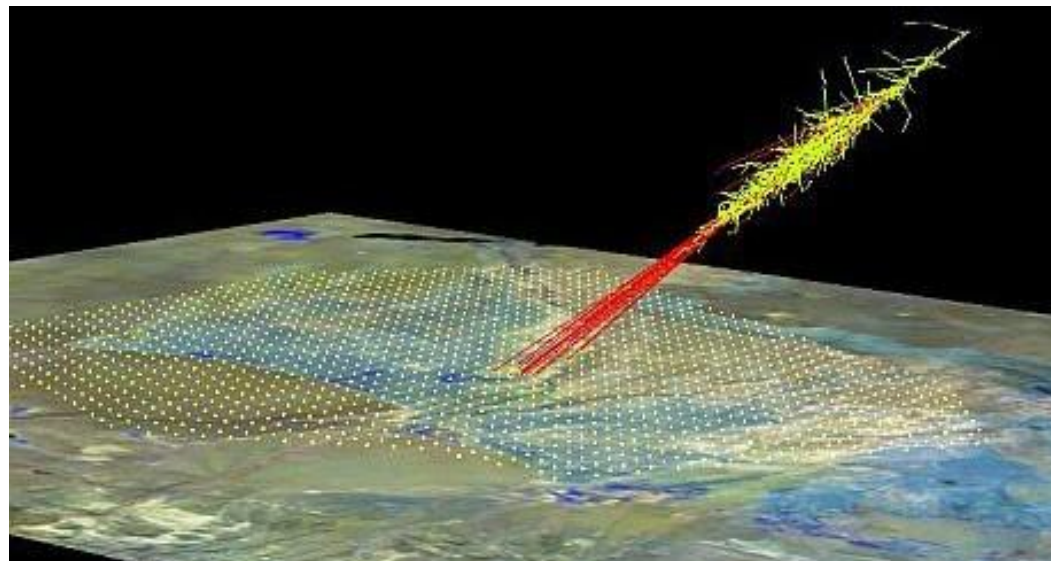
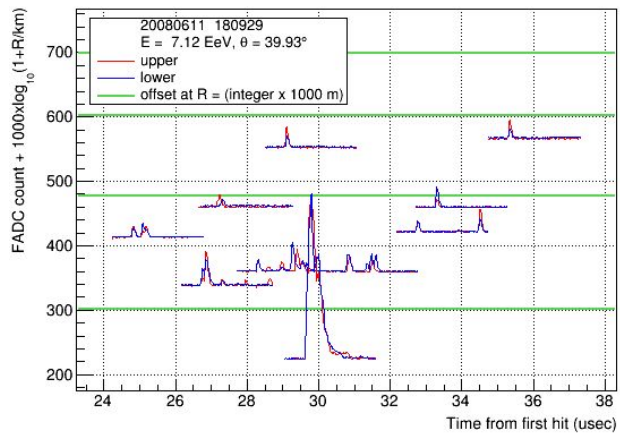
# Telescope Array



Largest cosmic ray observatory in Northern hemisphere  
(700 km<sup>2</sup>, 507 scintillator + 3 fluorescent stations)



# Telescope Array



# Neural network and data representation

# Data representation

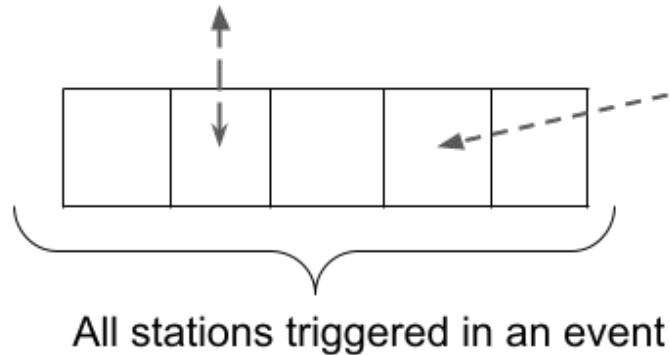
*Each events = set of activated stations*

Station feature vector:

1-3) Coordinates

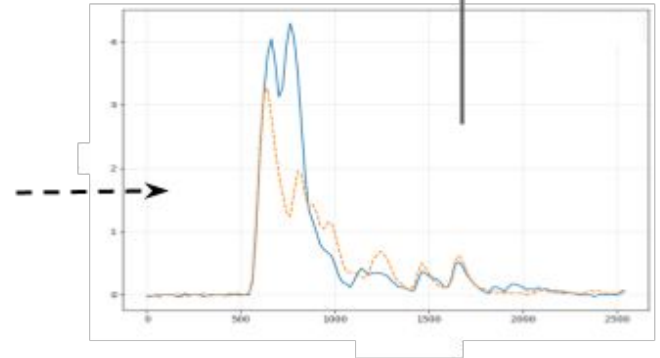
4-5) Activation times  $\oplus$  7-18) Waveform features

6) Registered charge



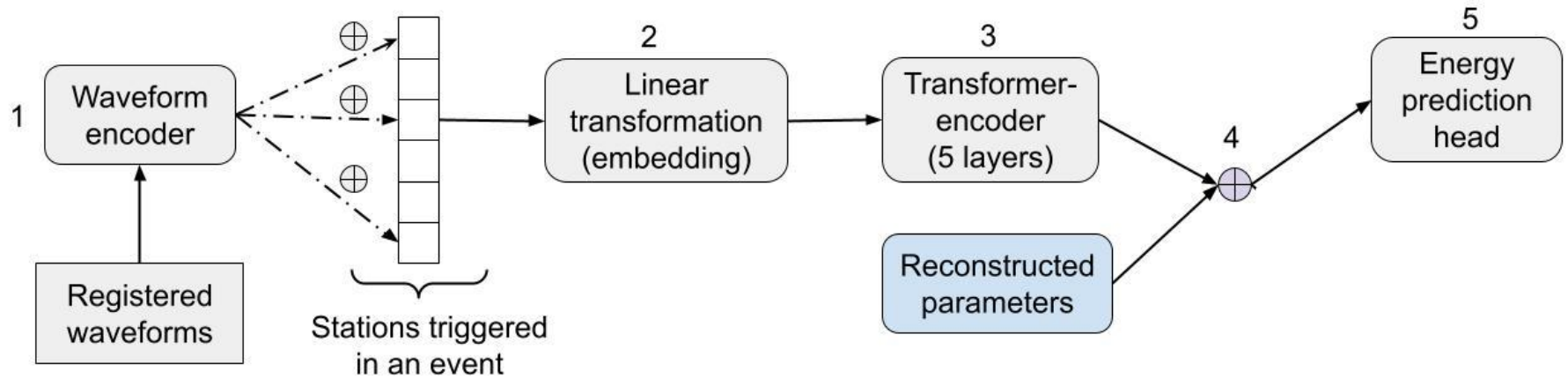
Single station

Corresponding waveform



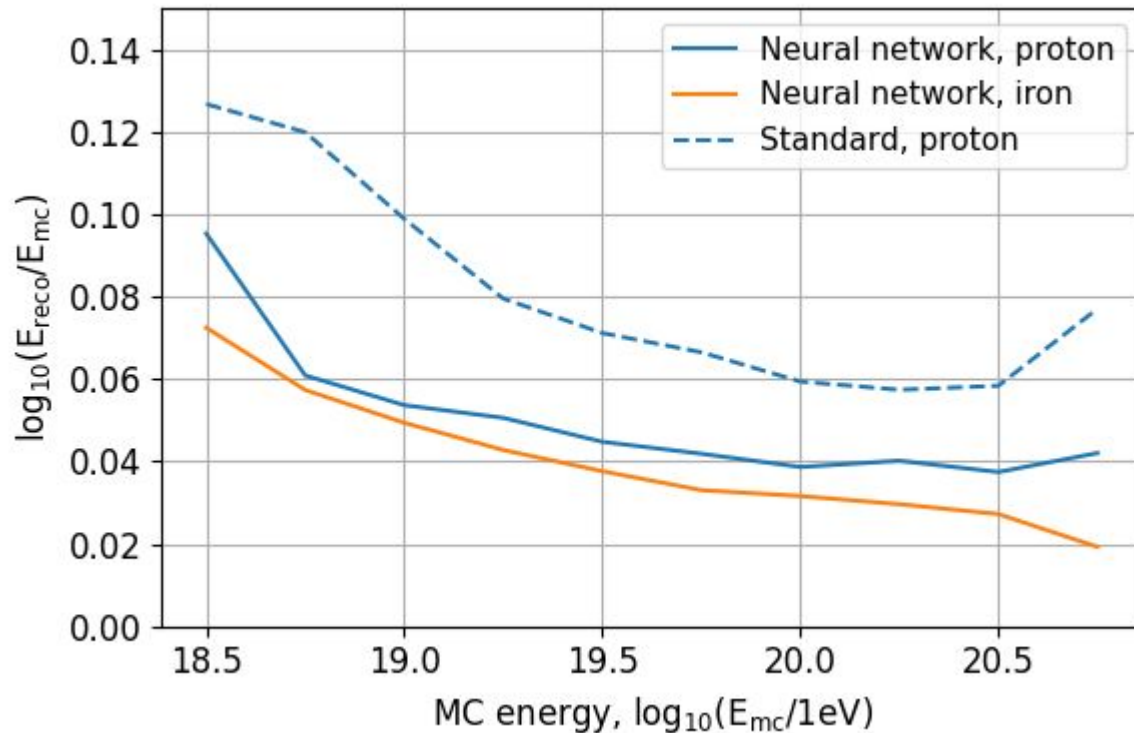
Waveform encoder

# NN architecture



- (1) Encode waveforms, concatenate with station properties (Q, t, r)
- (2) Embed to higher dimensionality
- (3) Pass through Transformer-Encoder
- (4) Concatenate with algorithmically reconstructed parameters
- (5) Predict Energy, Mass, Direction

# Energy resolution, 68%

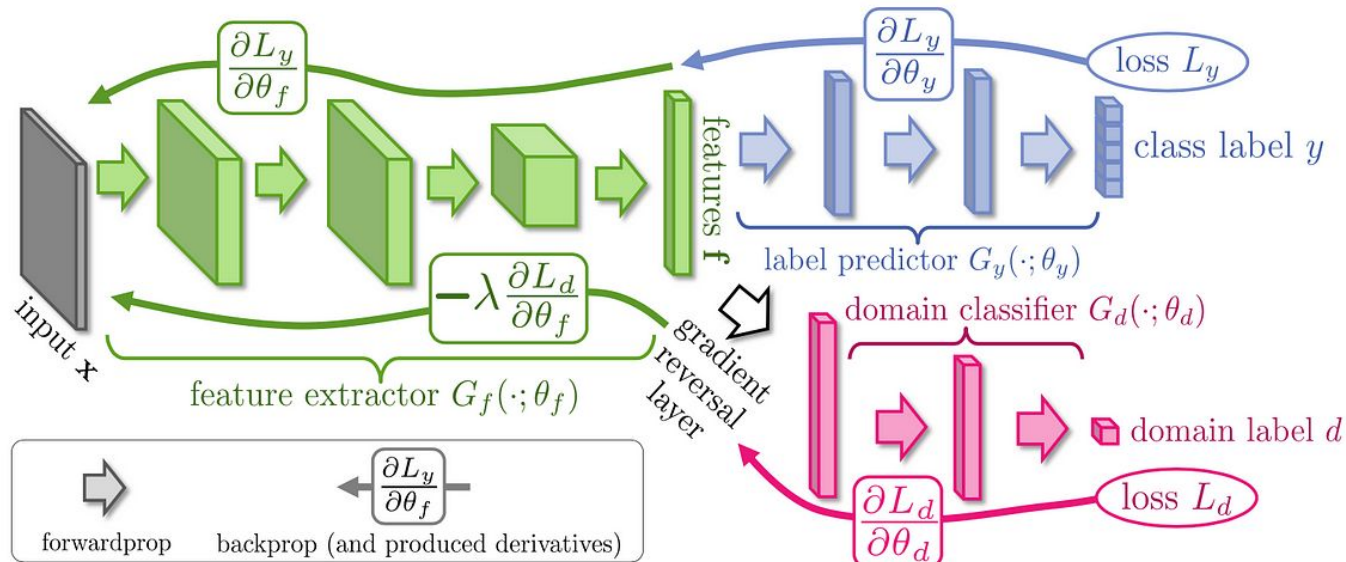


Neural network  
significantly  
improves energy  
resolution!

# Reliability of the neural network mitigating MC and real date differences

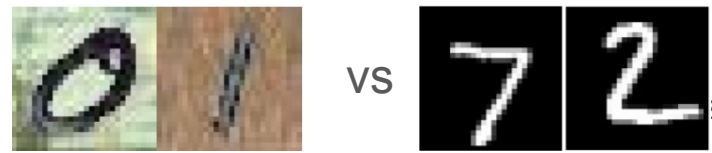
# Domain adaptation

Digit recognition



Domain classifier

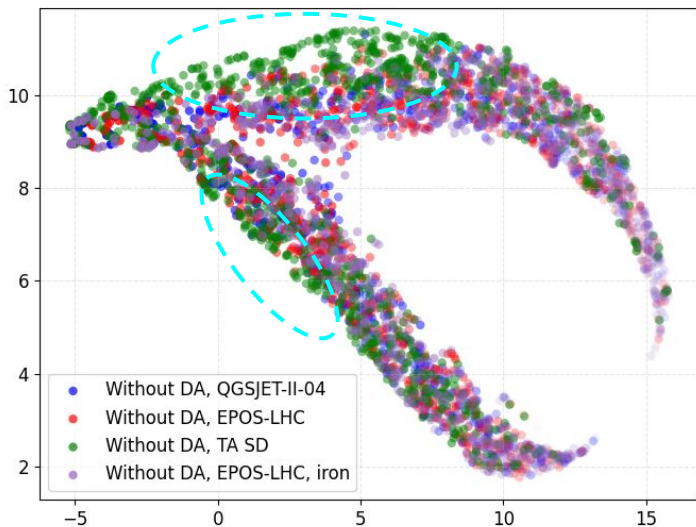
Gradient reversal:  
Exclude domain-specific features



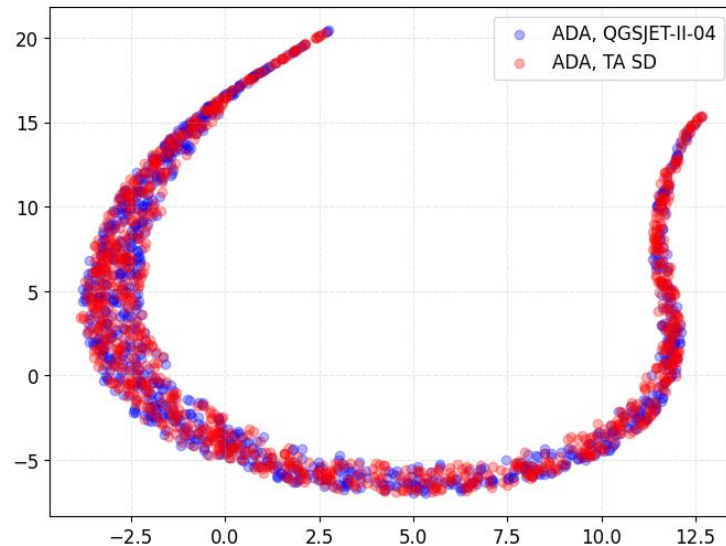
# Domain adaptation effect

## Comparison of feature distribution after the encoder

UMAP without DA



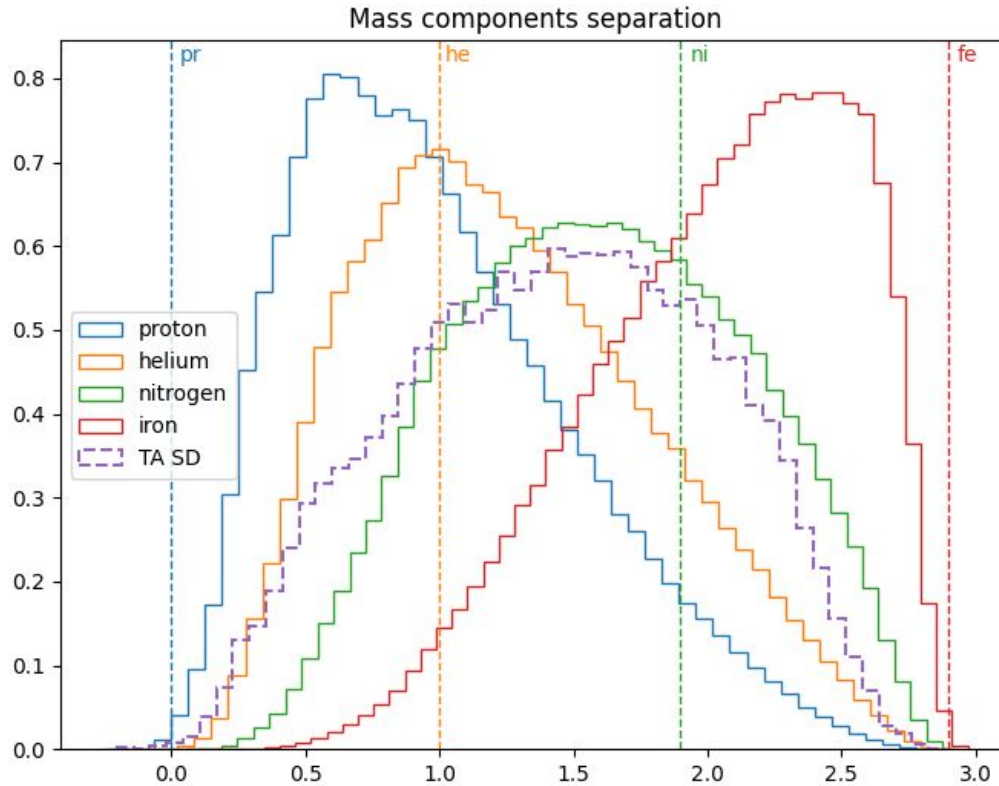
UMAP with DA



Domain Adaptation makes feature distributions close

# Mass-energy spectra

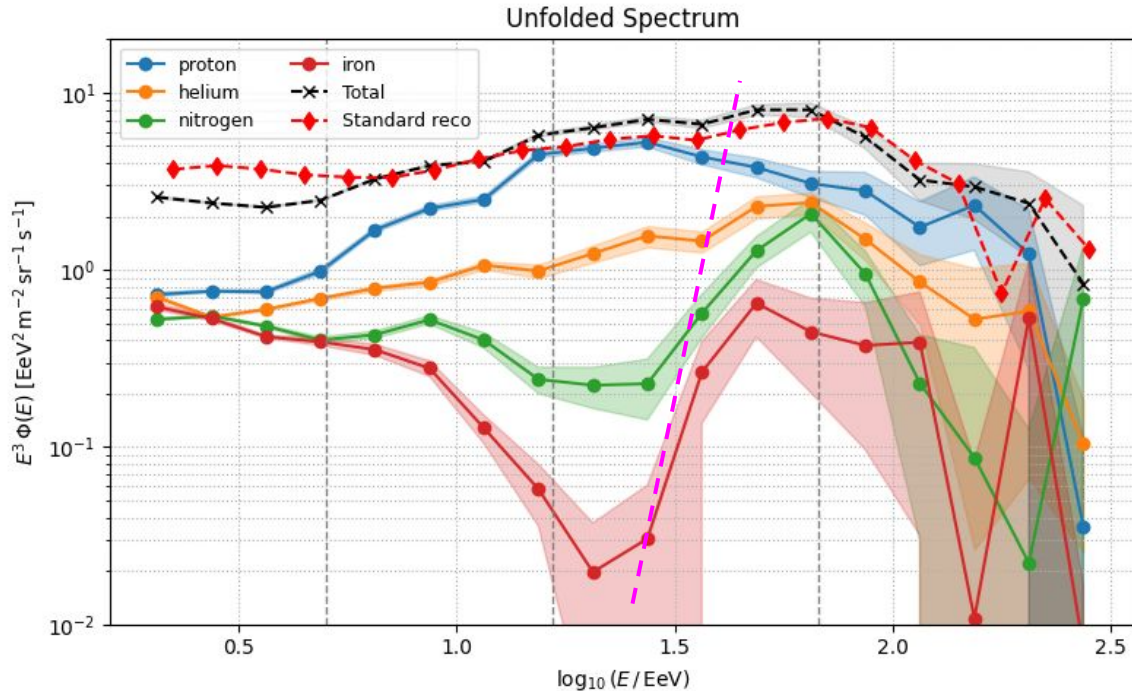
# Mass resolution



Good separation between  
proton and iron.

This would have been  
impossible without NNs.

# 2D iterative mass-energy unfolding



Iterative unfolding:

1. Unfold energy spectrum
2. Unfold mass spectrum

Rapid increase of heavy nuclei at high energies

This might be a sign of local source

# Conclusion

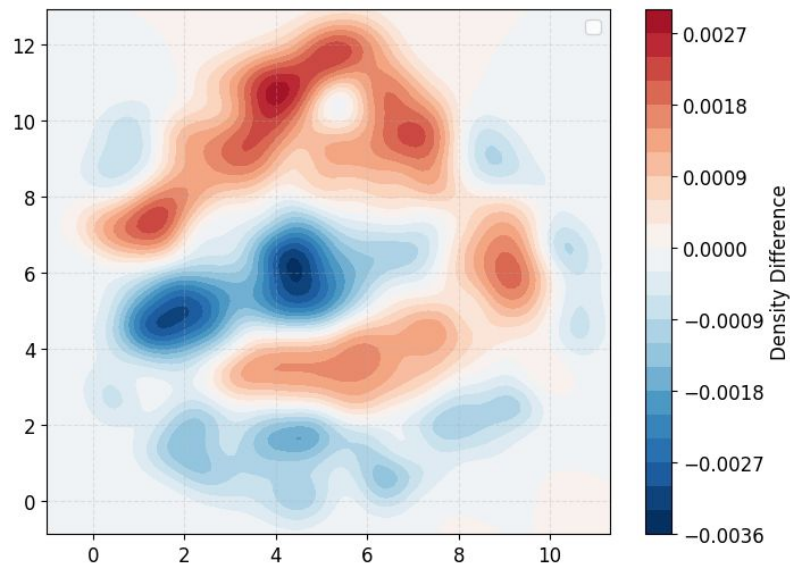
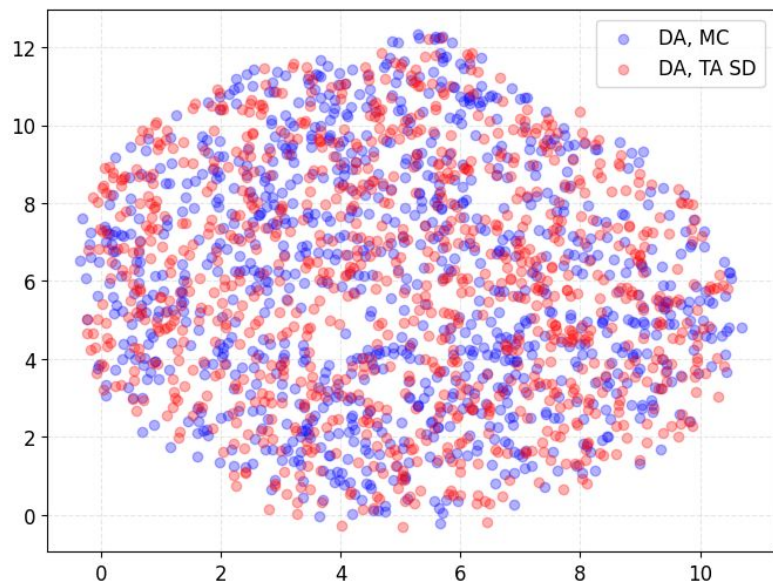
- Neural network improve energy resolution and allow for mass reconstruction
- Domain Adaptation is important for mitigating domain shift
- Mass-Energy spectra favours local source with intermediate mass spectrum

# Appendix

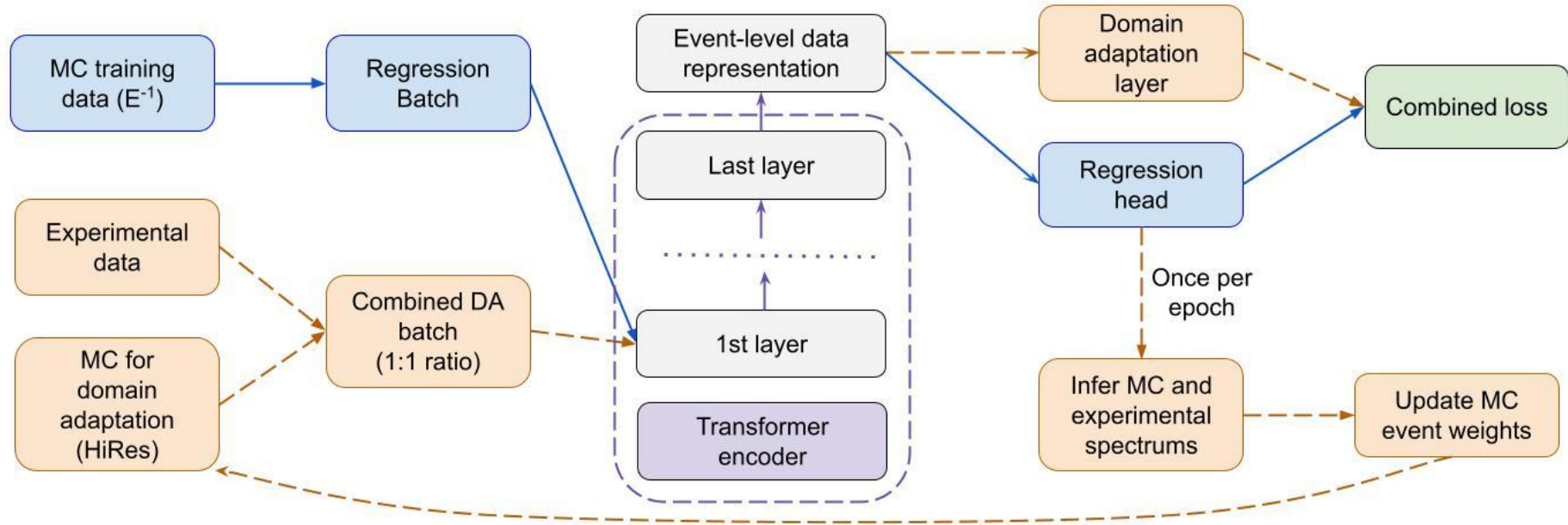
# Full Domain Adaptation

- 4 cosmic ray primaries: proton, helium, nitrogen, iron
- Domain adaptation for mass, energy and arrival direction distributions.

UMAP for encoder features

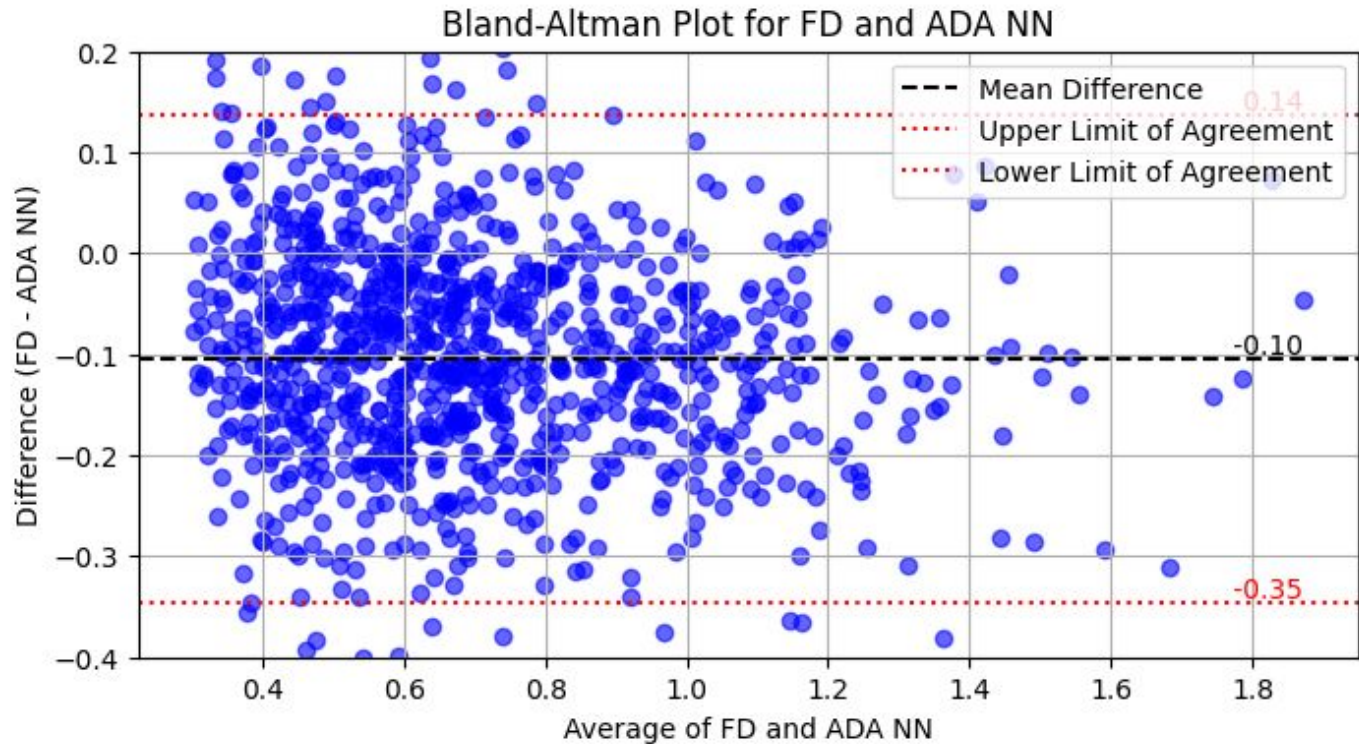


# Data flow with Adaptive Domain Adaptation



find equilibrium point - iteratively reweight MC events so that during training the estimated MC and TA SD data spectra coincide.

# SD/FD comparison



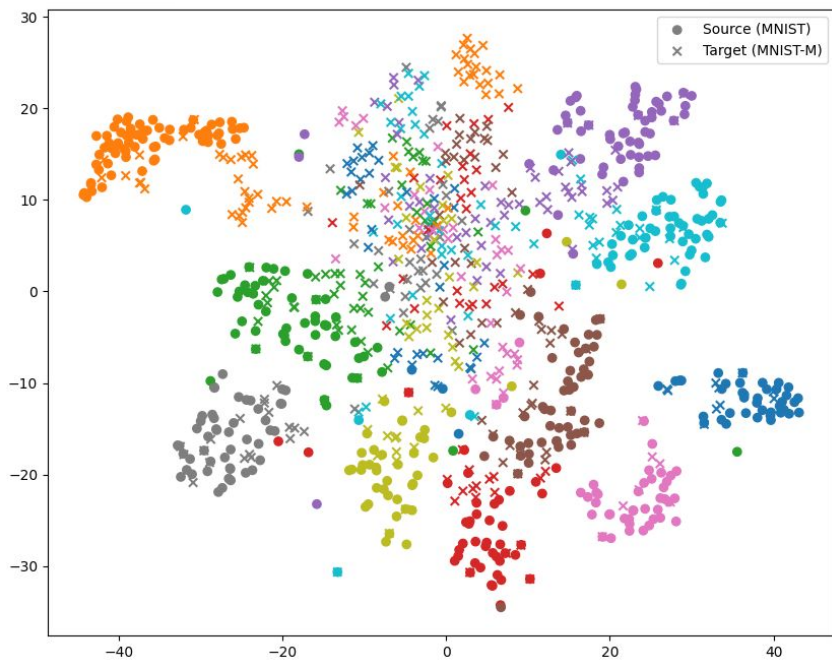
Good agreement with FD: residual variance 0.0075 (0.01 for standard reco)

# Illustration of Domain Adaptation

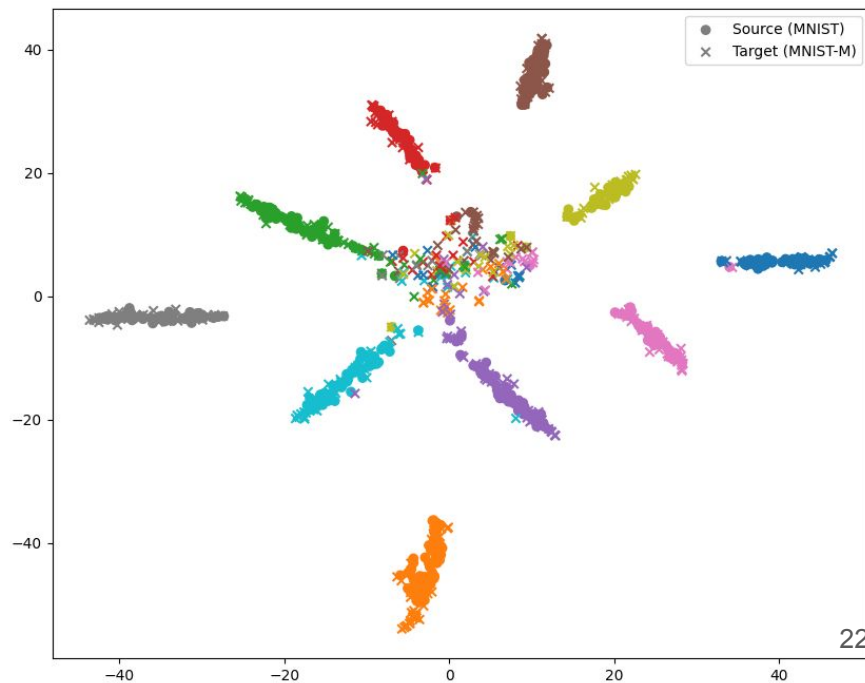
Accuracy of predictions on MNIST-M:

without DA: 57.5%, with DA:81.5, direct learning: 98.9%

2D t-SNE without DA

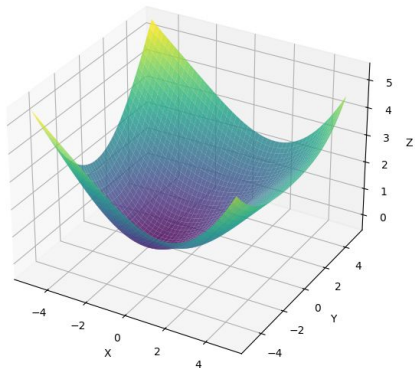


2D t-SNE with DA

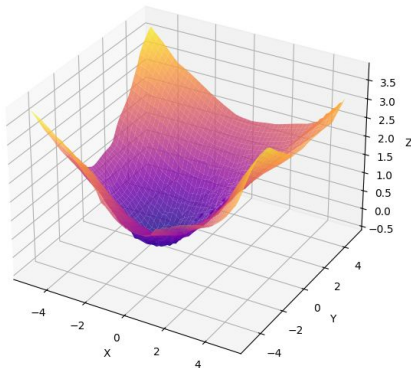


# Why doing domain adaptation

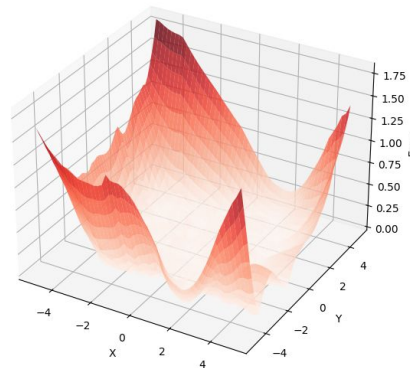
True Function  
(Full Range)



Neural Network Predictions  
(Full Range)

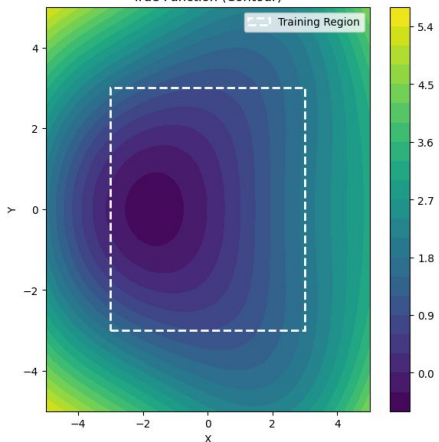


Absolute Error Surface  
(Red = Higher Error)

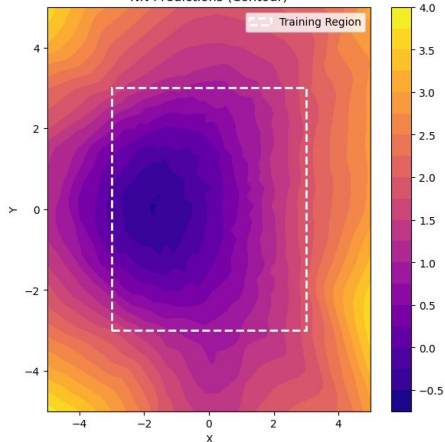


NNs are bad at extrapolating

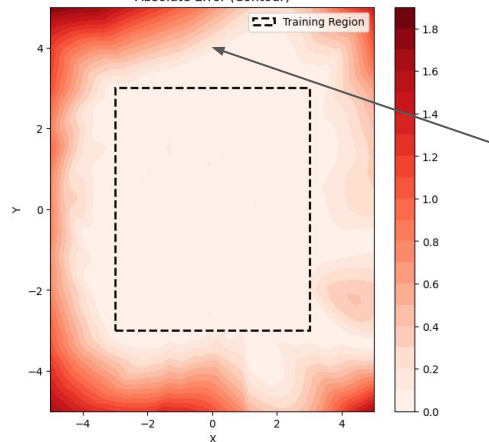
True Function (Contour)



NN Predictions (Contour)



Absolute Error (Contour)



High uncontrollable errors outside the training phase space