

# Software Compensation in the AHCAL with Machine Learning

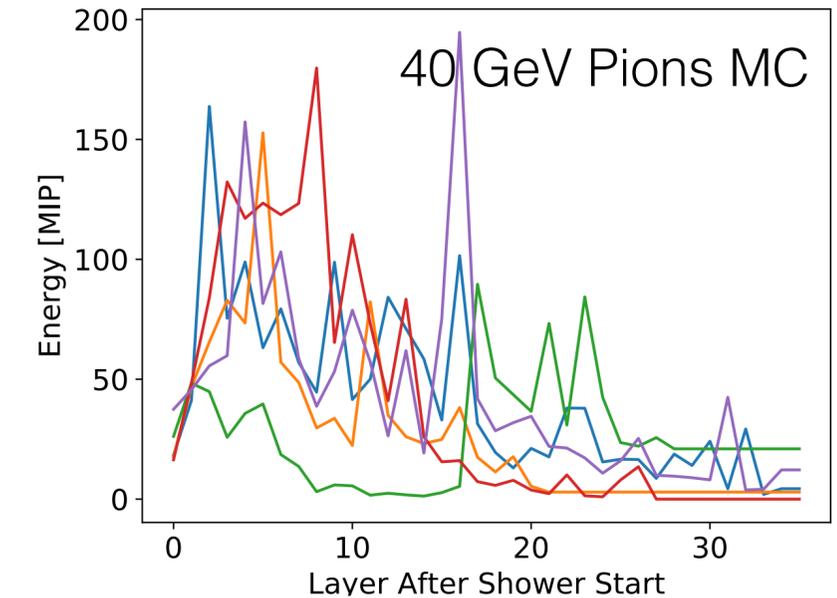
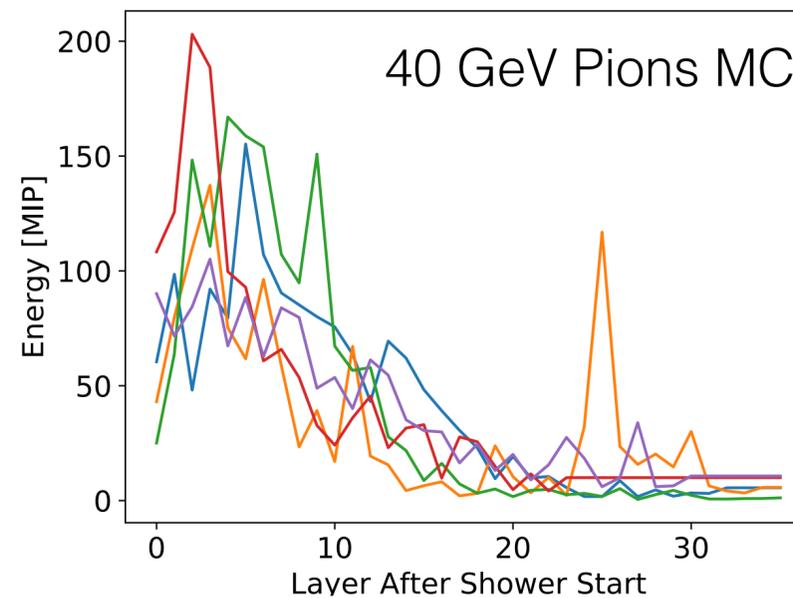
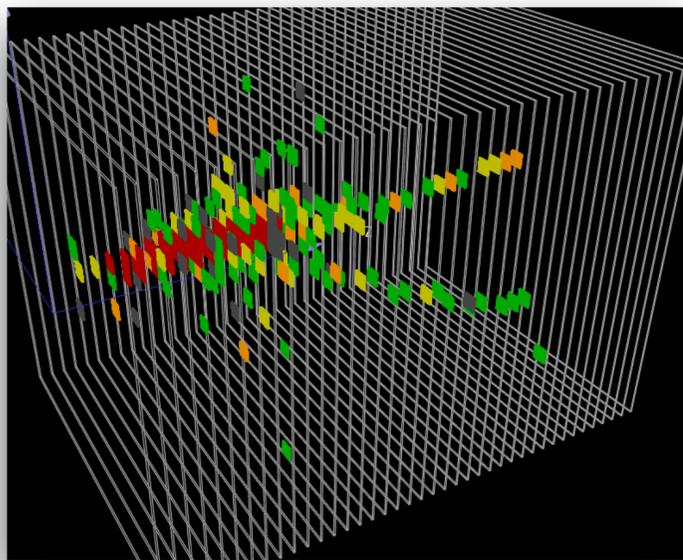
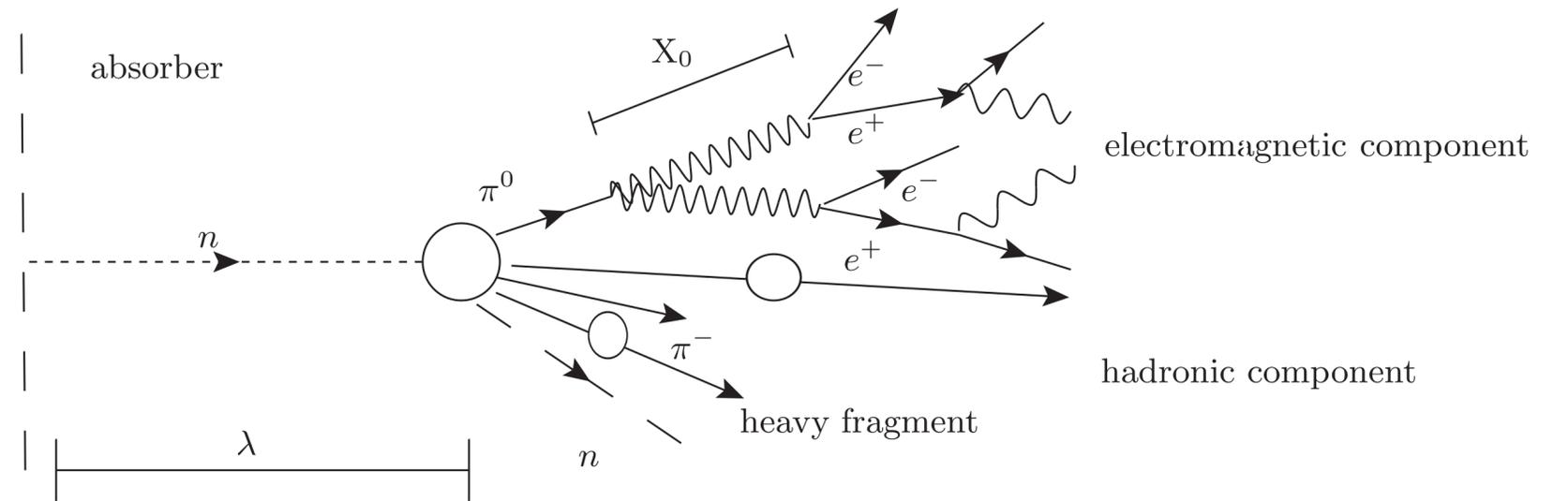
April 12th, 2019

*CALICE Meeting*  
*- Utrecht -*

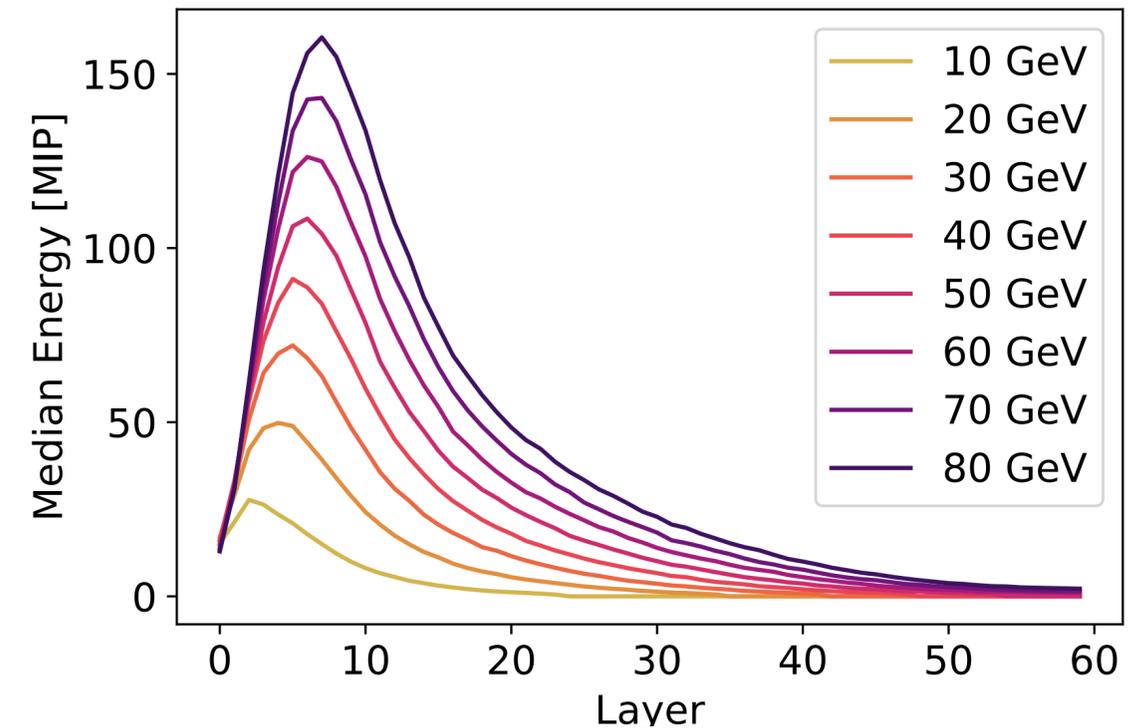
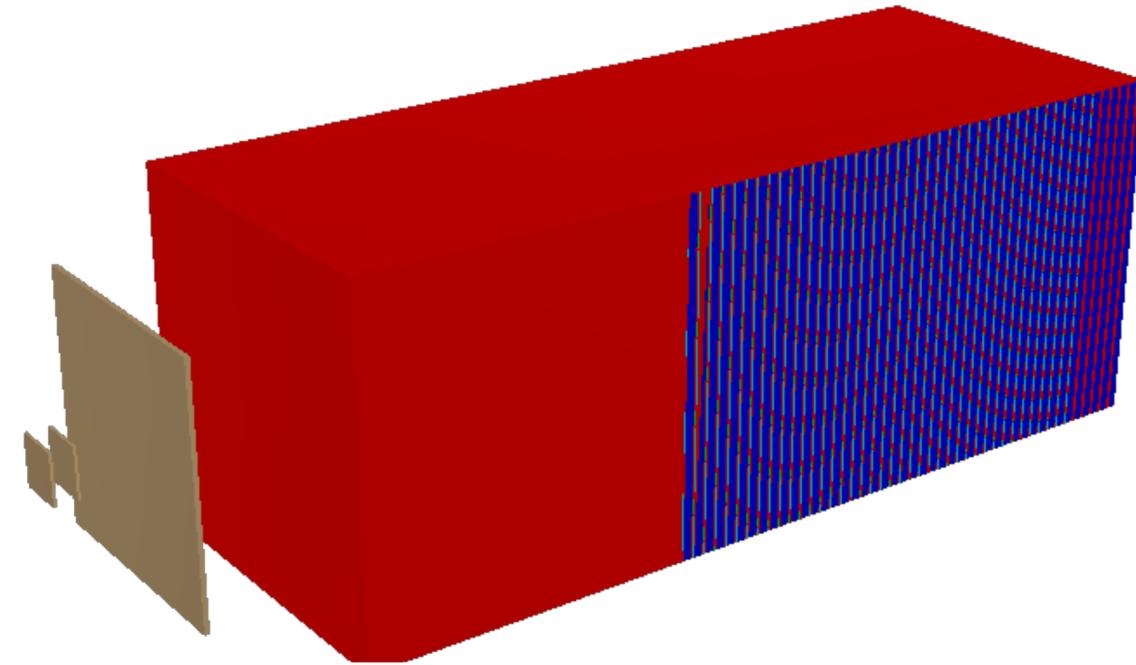
Christian Graf



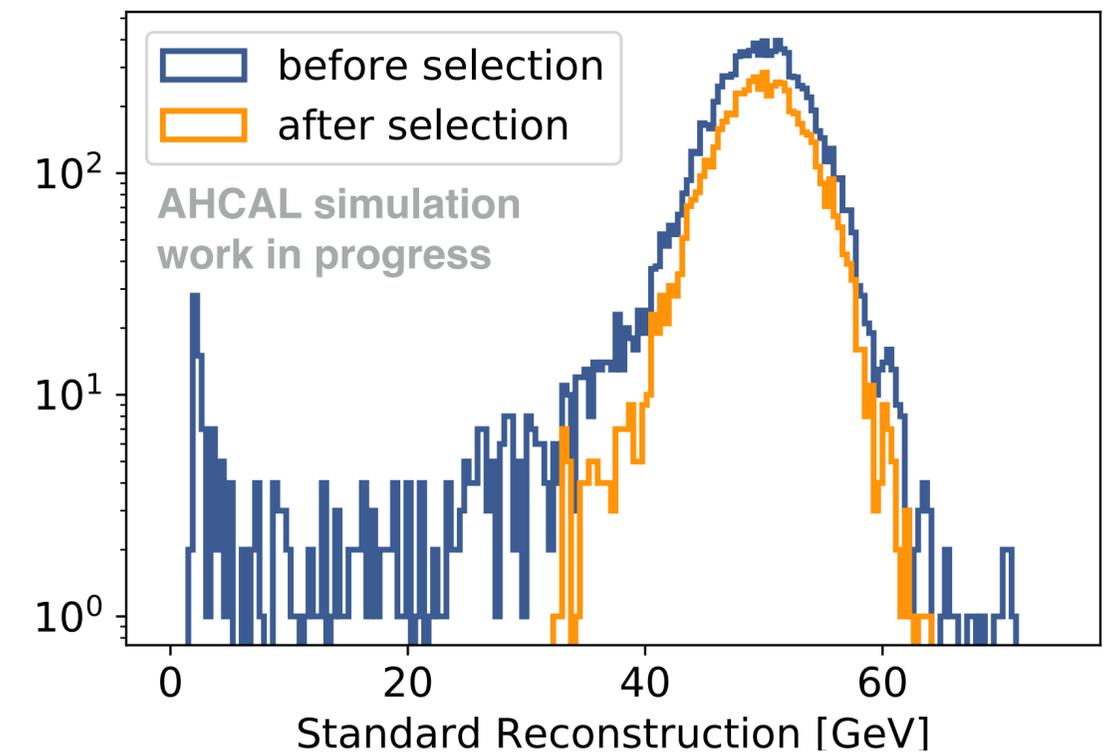
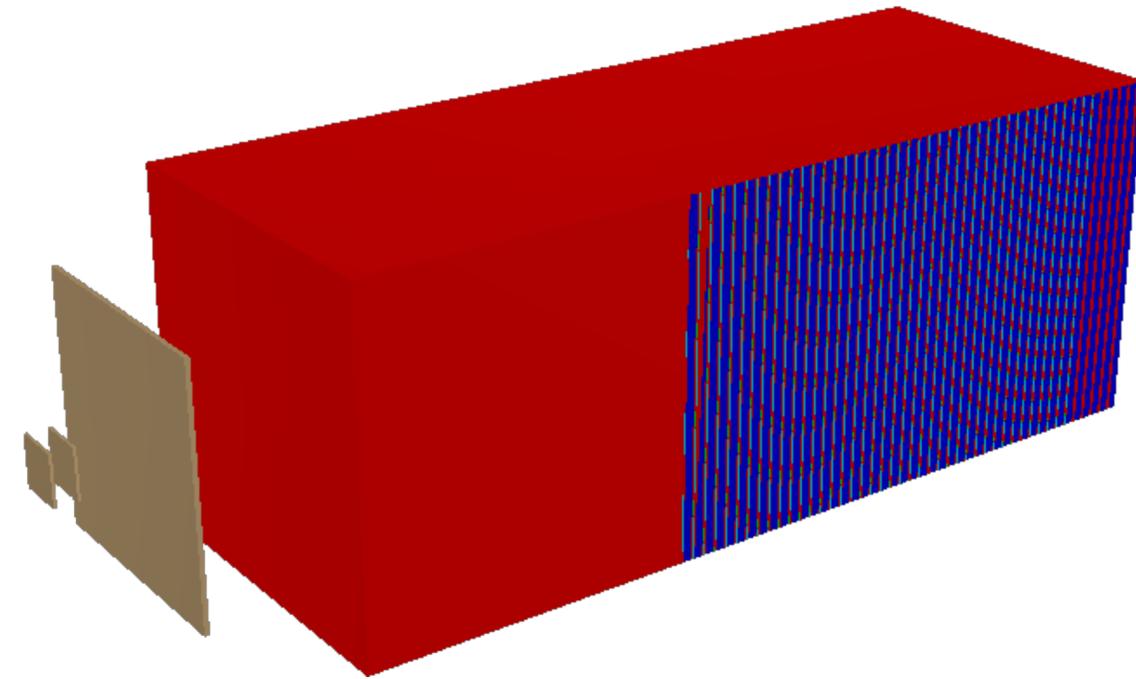
- Fluctuations of a hadronic shower in calorimeter:
  - Measurement effects
  - **EM fraction**
  - Invisible energy
  - Sampling fluctuations
  - Catastrophic events
  - ...



- CALICE AHCAL test beam geometry with 60 layers
- 10 - 85 GeV Pions in 1 GeV steps
- 10,000 events per Energy
- 1ns gaussian time smearing



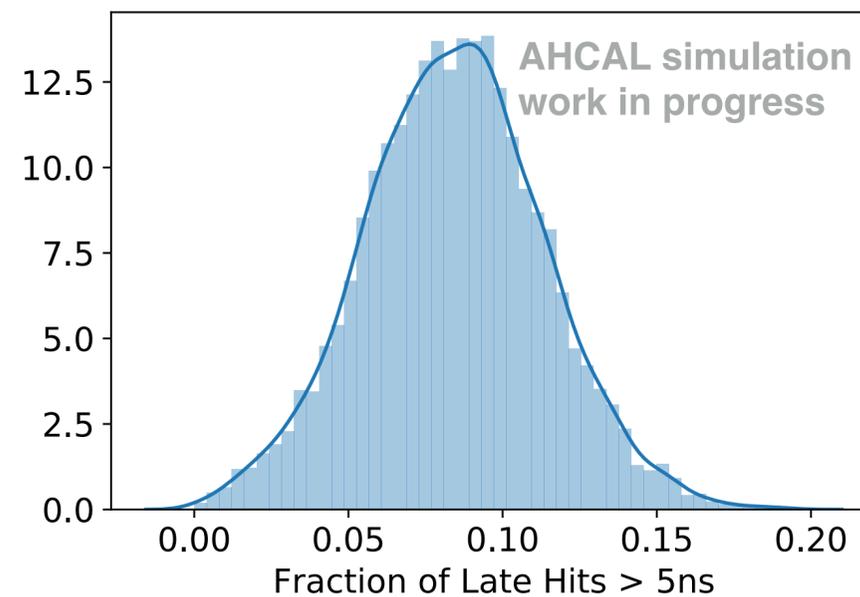
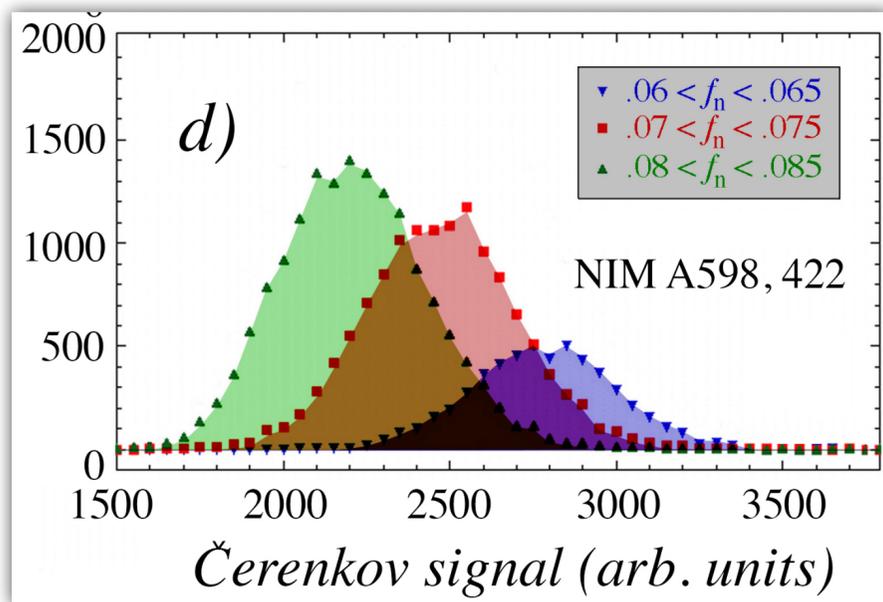
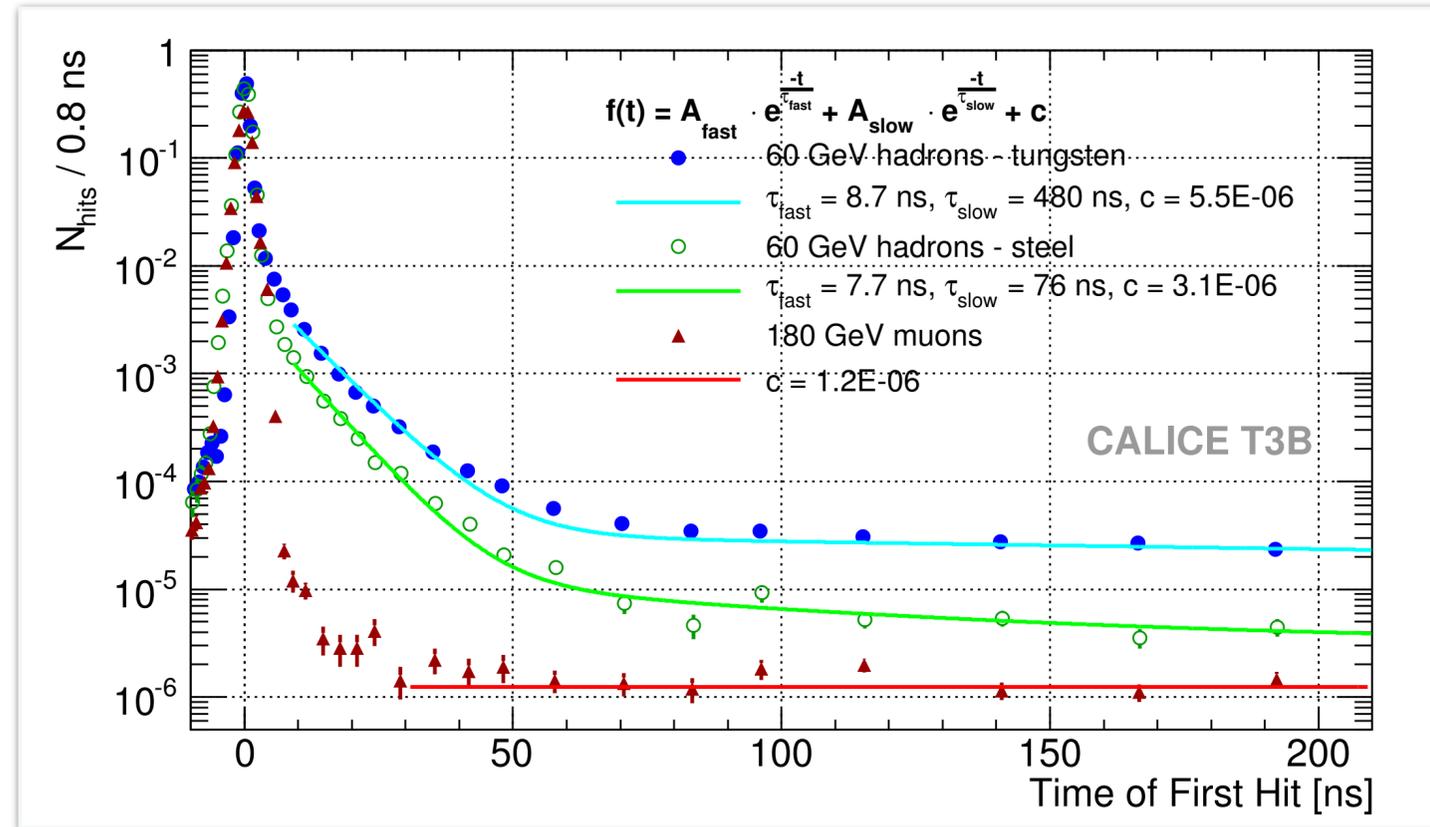
- CALICE AHCAL test beam geometry with 60 layers
- 10 - 85 GeV Pions in 1 GeV steps
- 10,000 events per Energy
- 1 ns gaussian time smearing
- Event selection:
  - Shower start before layer 15
  - Cut on  $\pm 2.5\sigma$



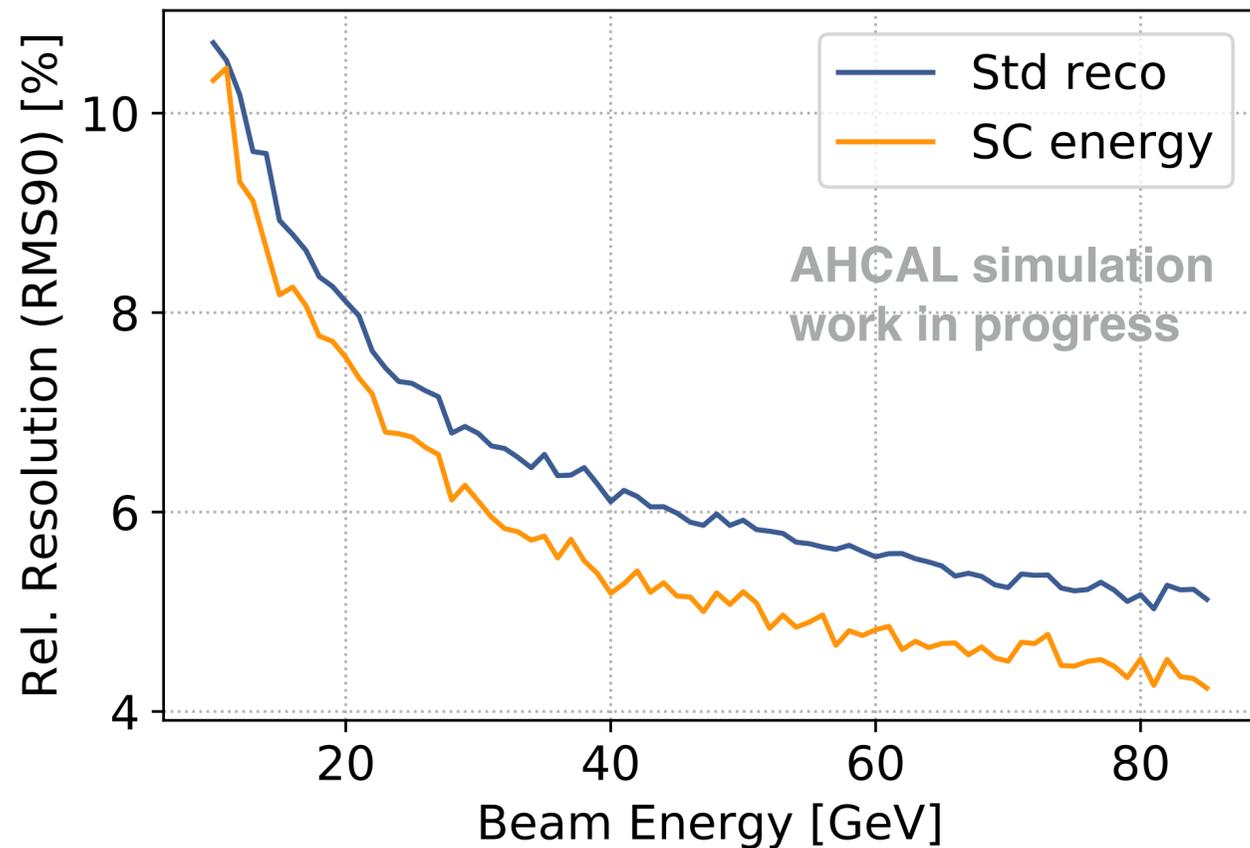
# Timing



- Timing capabilities with technological prototype
- Slow neutrons created in hadronic processes  
—> late hits
- **May help energy reconstruction?**
  - Correlation between neutron content and em-fraction observed in dual read-out calorimeter
- Problem: Not many neutrons in steel absorber!

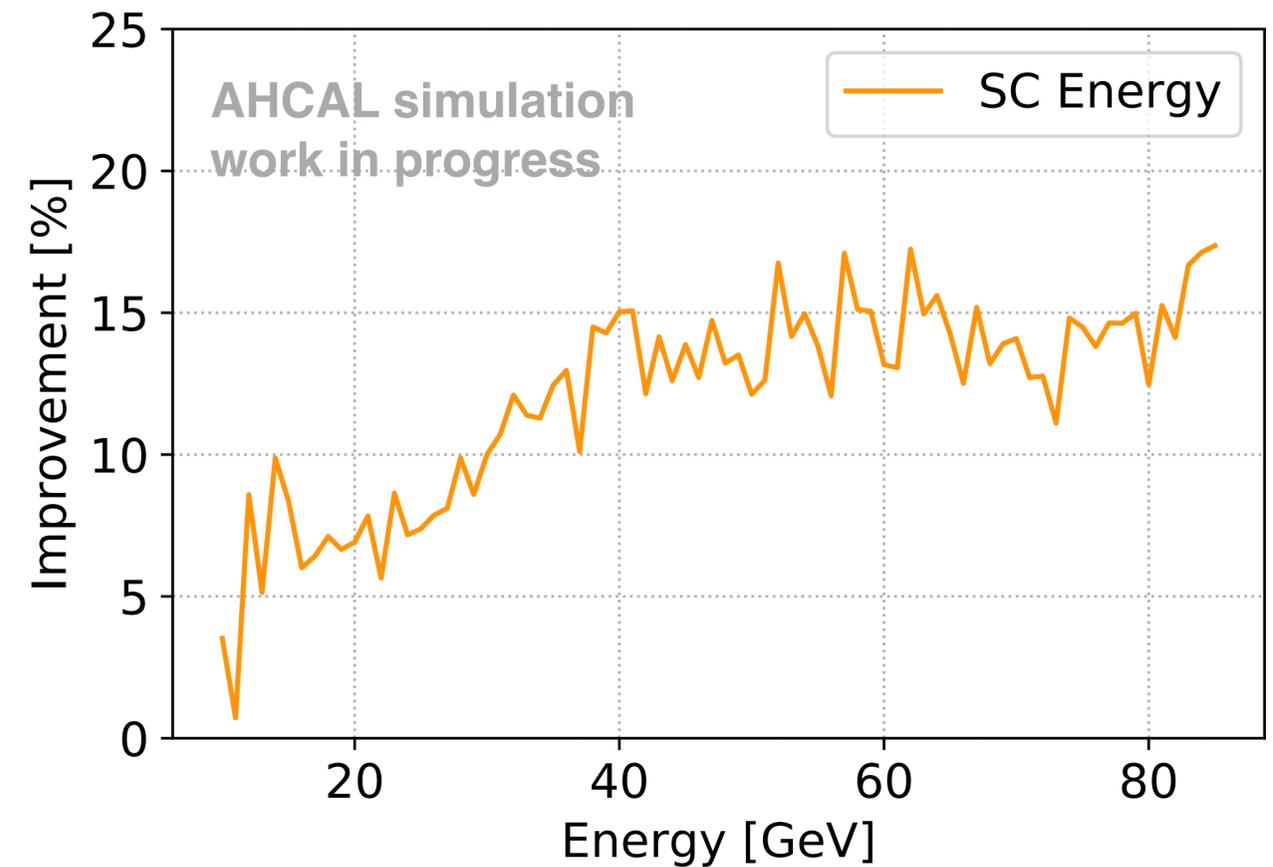


- Global software compensation approach yields **~15%** improvement over std reco
- Using Fraction of hits **> 5 MIP**



## Optimize:

$$E_{\text{reco}} = a + b * f_{\text{HighEnergyHits}} + c * f_{\text{HighEnergyHits}}^2$$



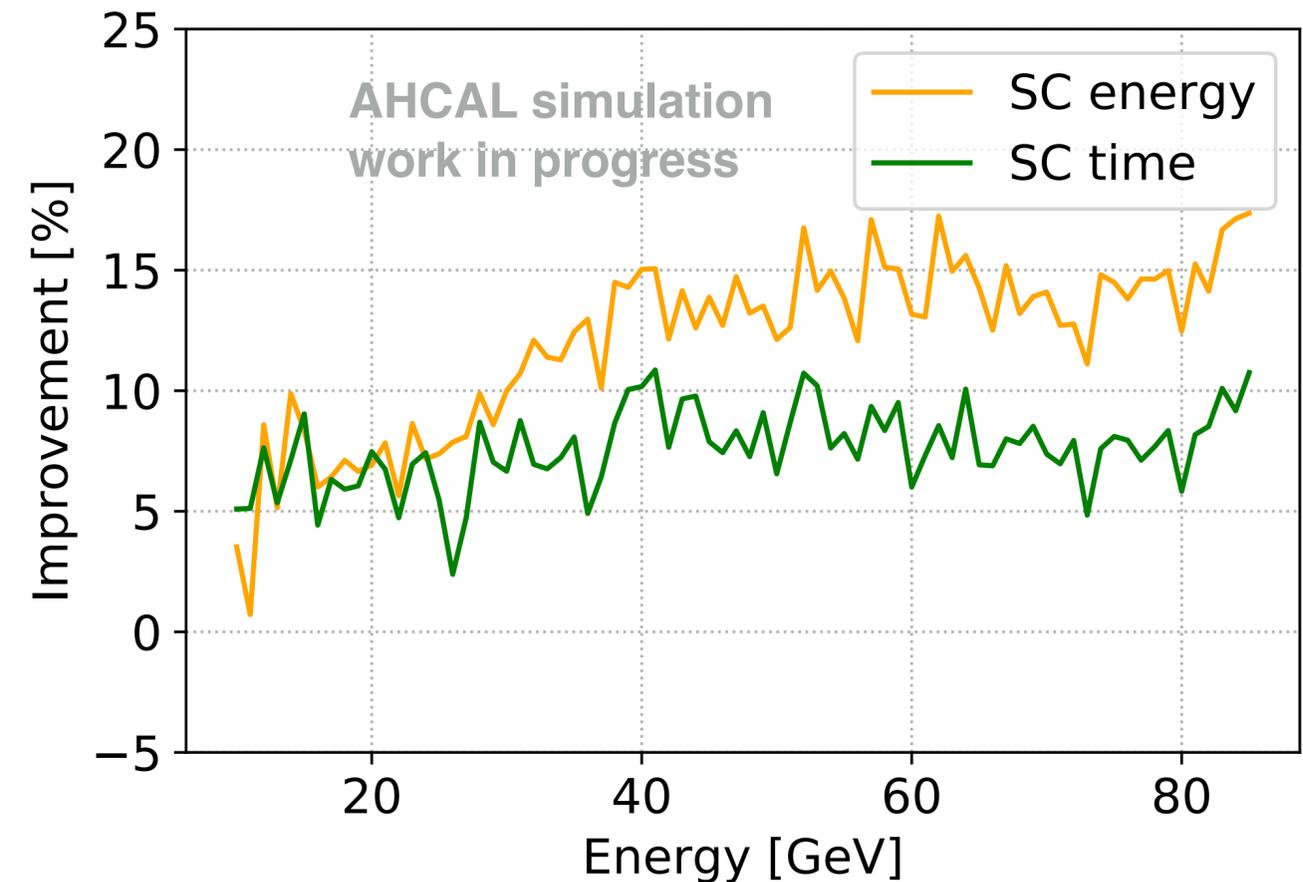
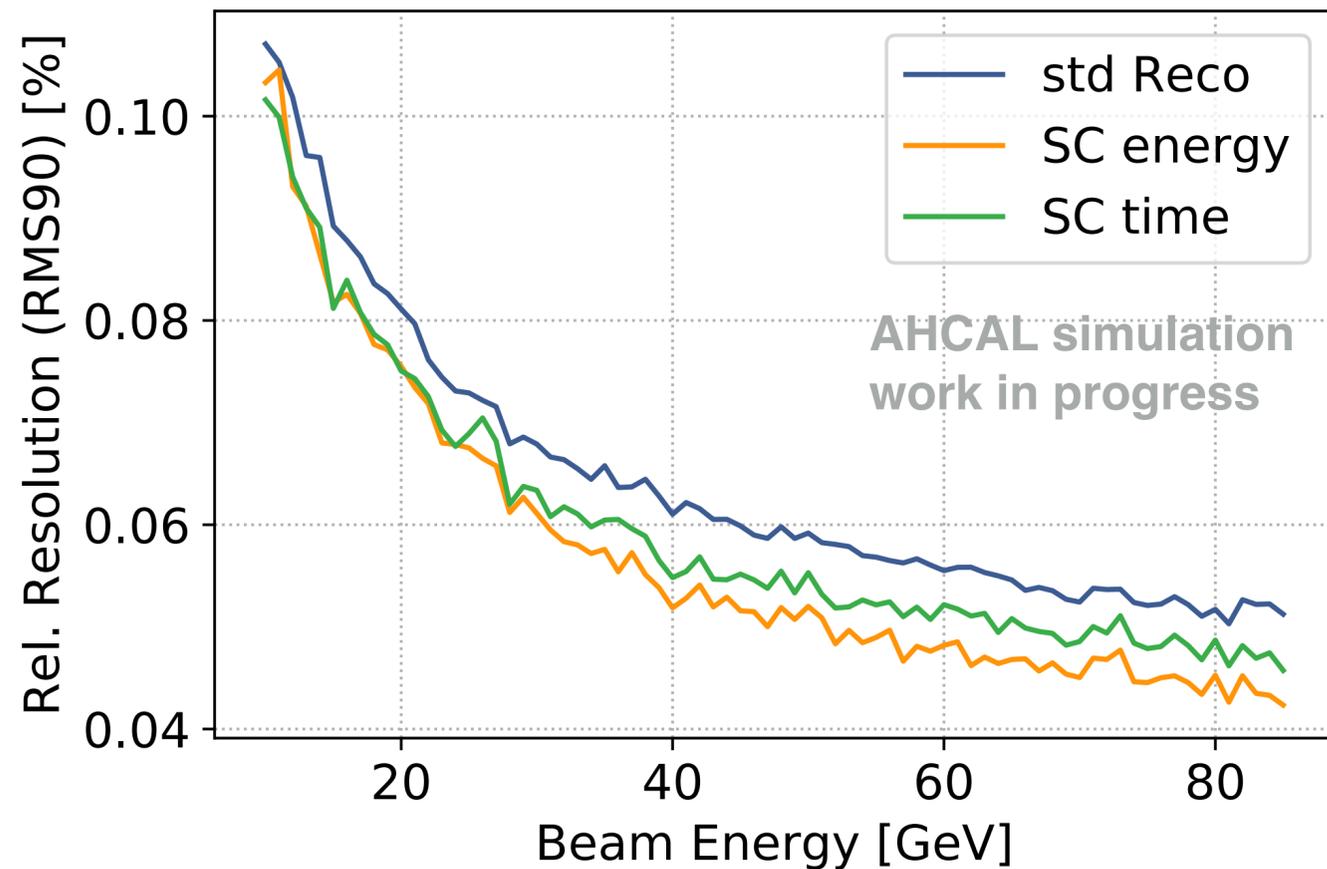
# Software Compensation - Time



- Global software compensation approach with **time** yields **~8%** improvement over std reco
- Using fraction of late hits **> 5ns**

**Optimize:**

$$E_{\text{reco}} = a + b * f_{\text{LateHits}} + c * f_{\text{LateHits}}^2$$



- Global SC approach with **energy and time** yields **~17%** improvement over std reco
- Using fraction of late hits **> 5ns** and hits **> 5 MIP**

## Optimize:

$$E_{\text{reco}} = a + b * f_{\text{HighEnergyHits}} + c * f_{\text{LateHits}} + d * f_{\text{LateHits}} * f_{\text{HighEnergyHits}}$$

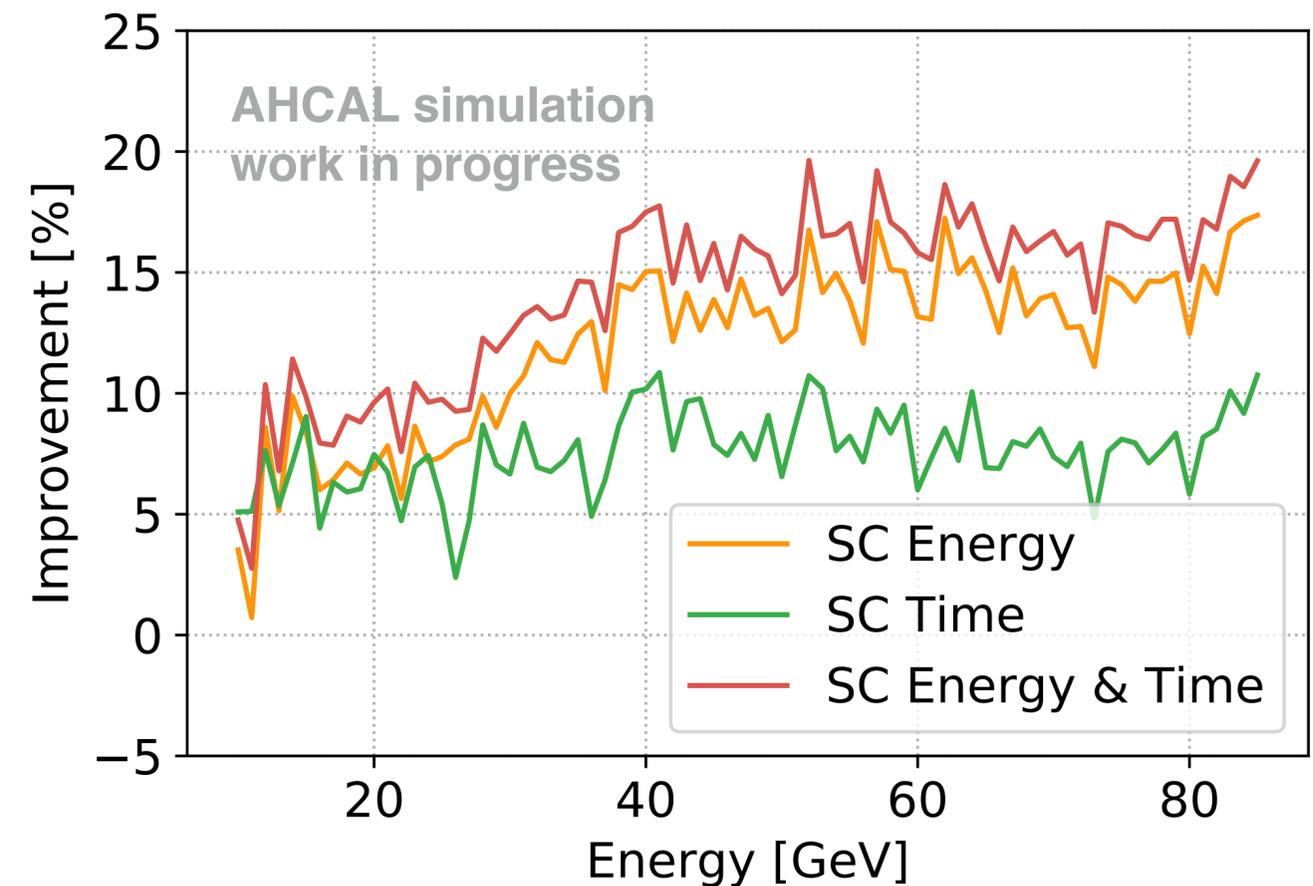
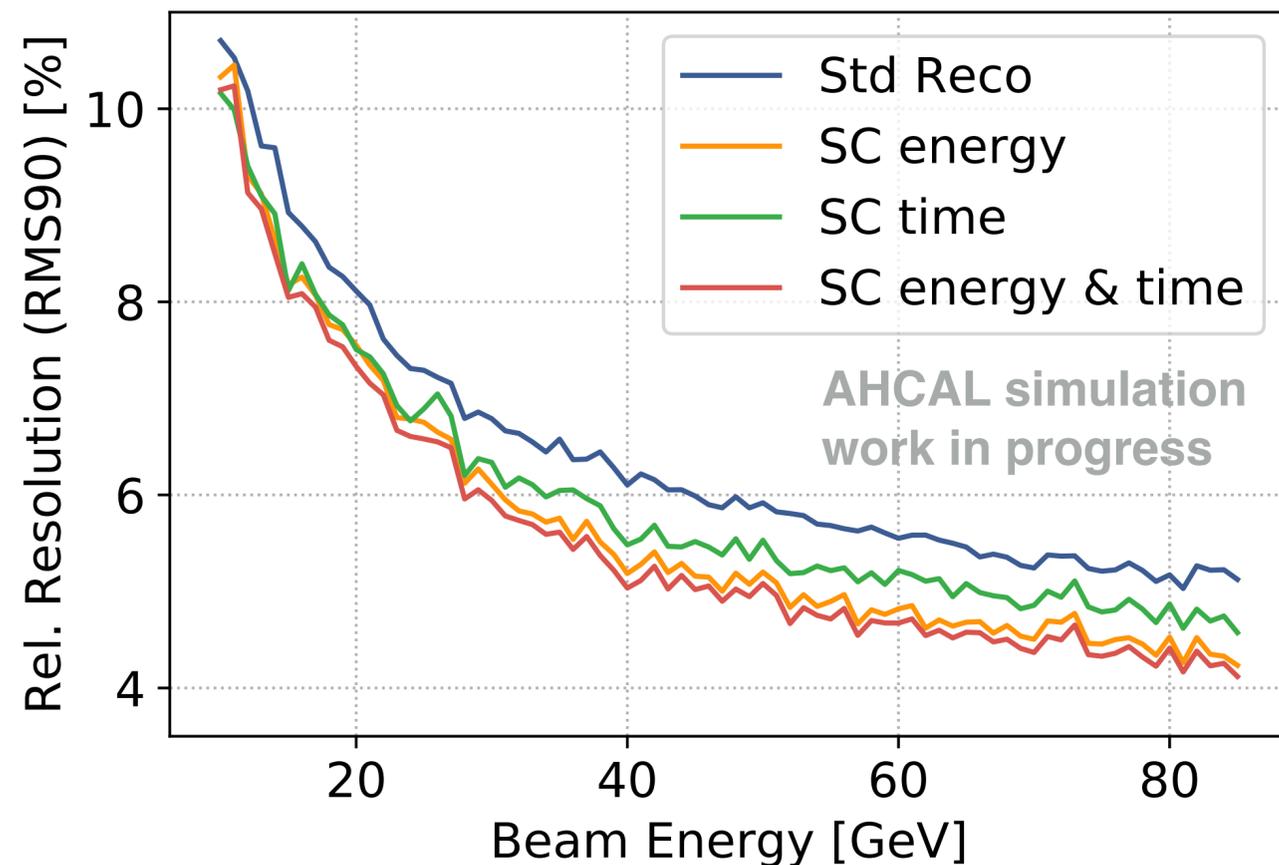
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- Using fraction of late hits **> 5ns** and hits **> 5 MIP**

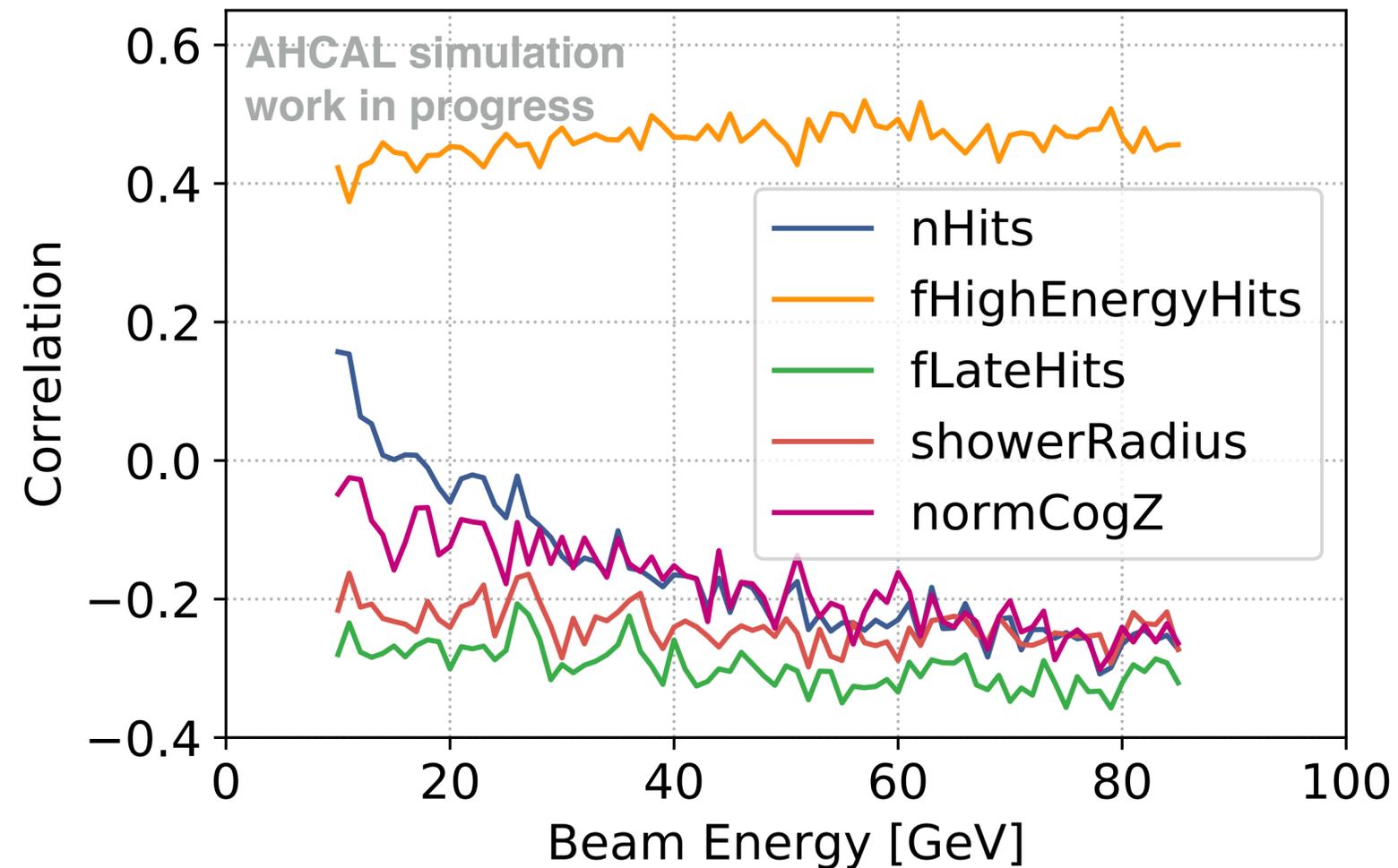
## Optimize:

$$E_{\text{reco}} = a + b * f_{\text{HighEnergyHits}} + c * f_{\text{LateHits}} + d * f_{\text{LateHits}} * f_{\text{HighEnergyHits}}$$



- Correlation coefficients of features with reconstructed energy

## Before SC

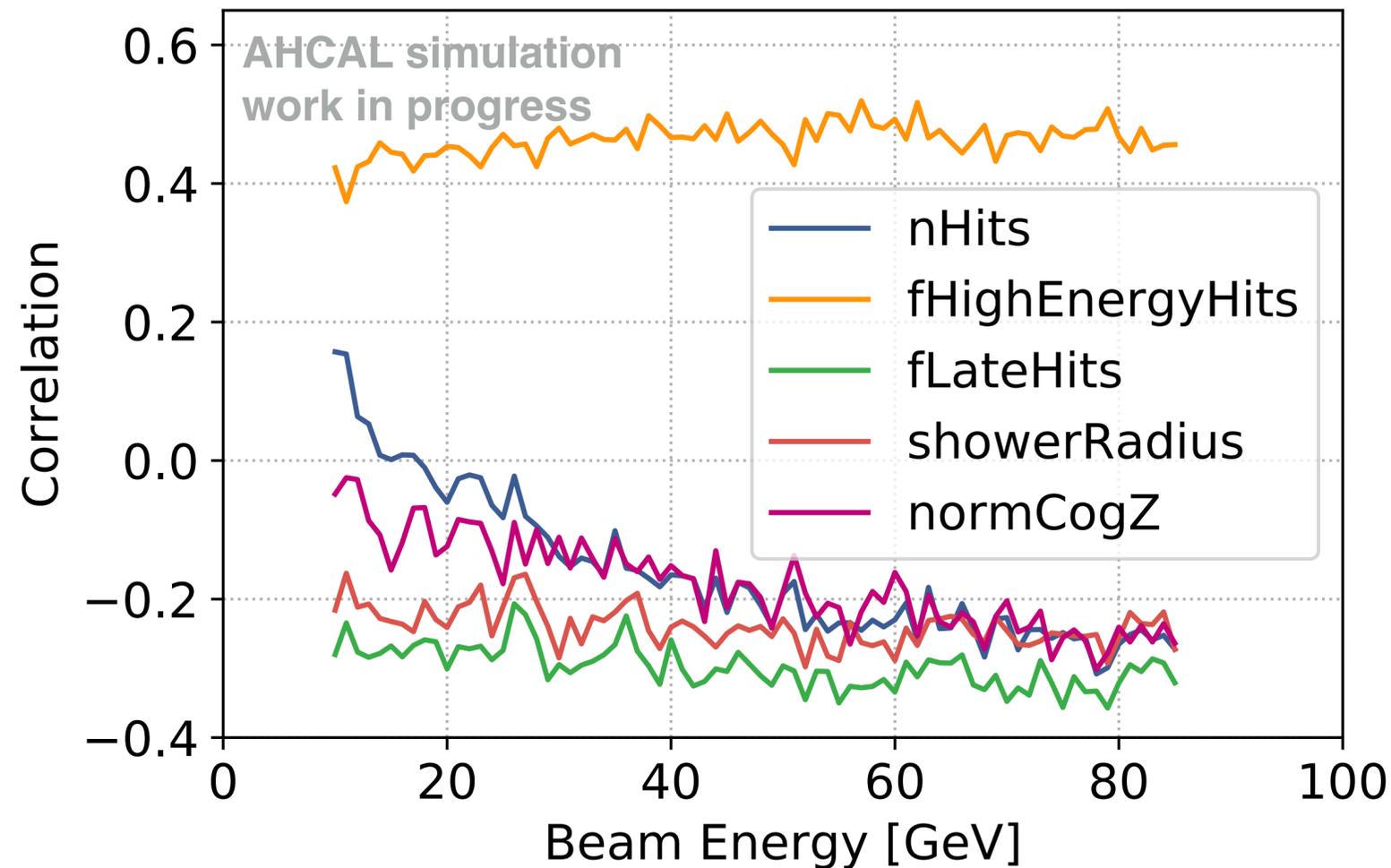


# Correlations

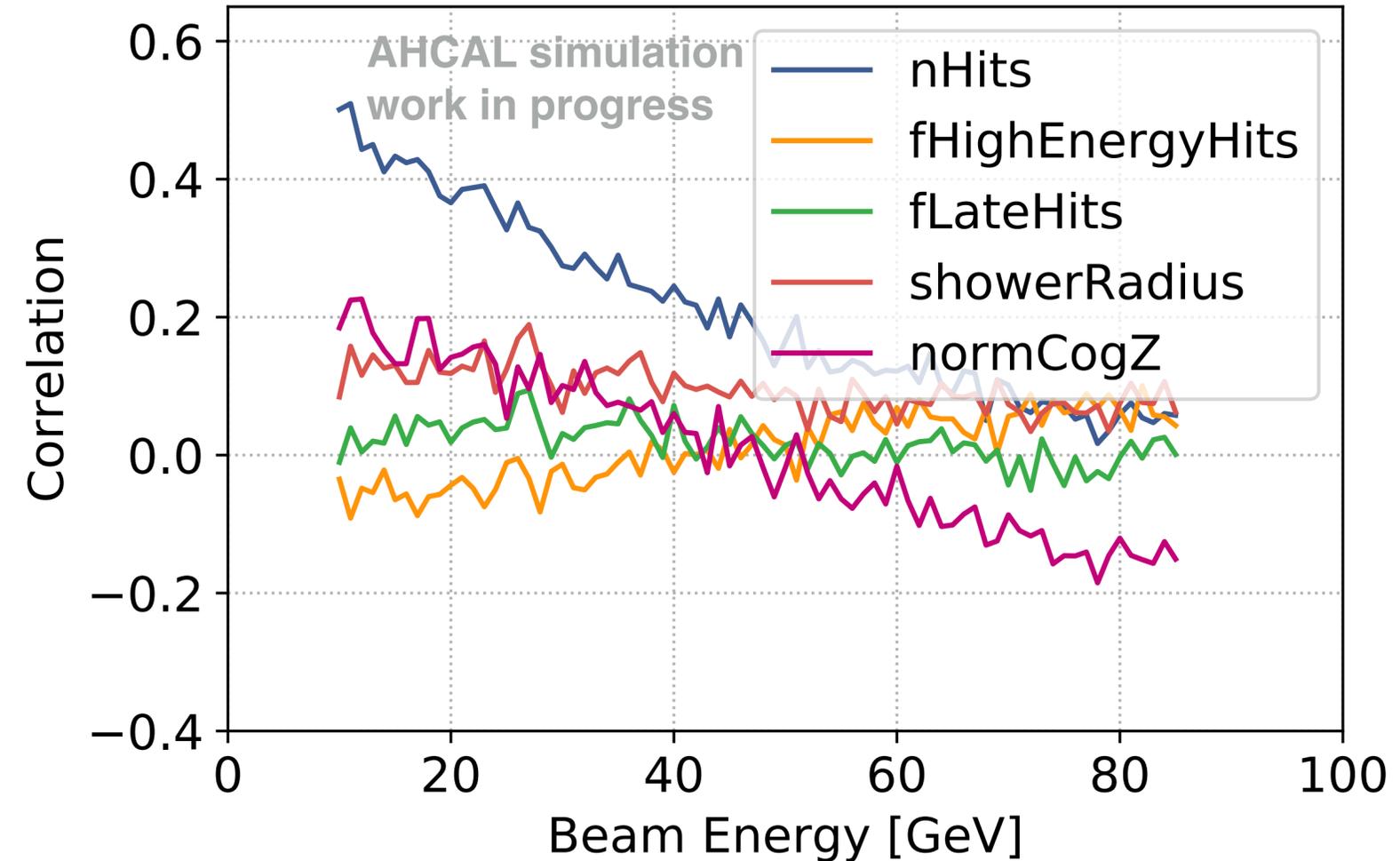


- Correlation coefficients of features with reconstructed energy
- After software compensation: still some correlation left

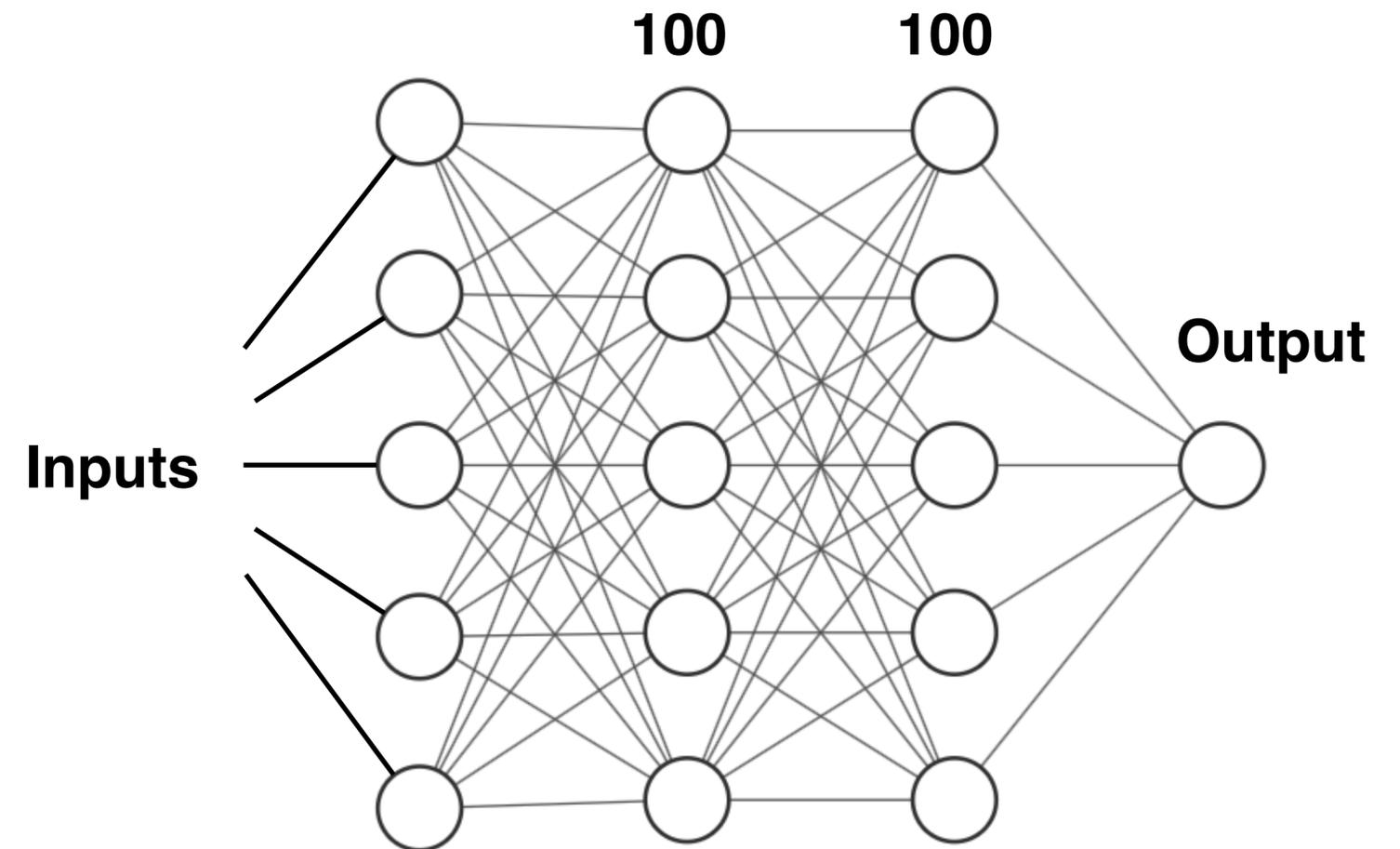
Before SC



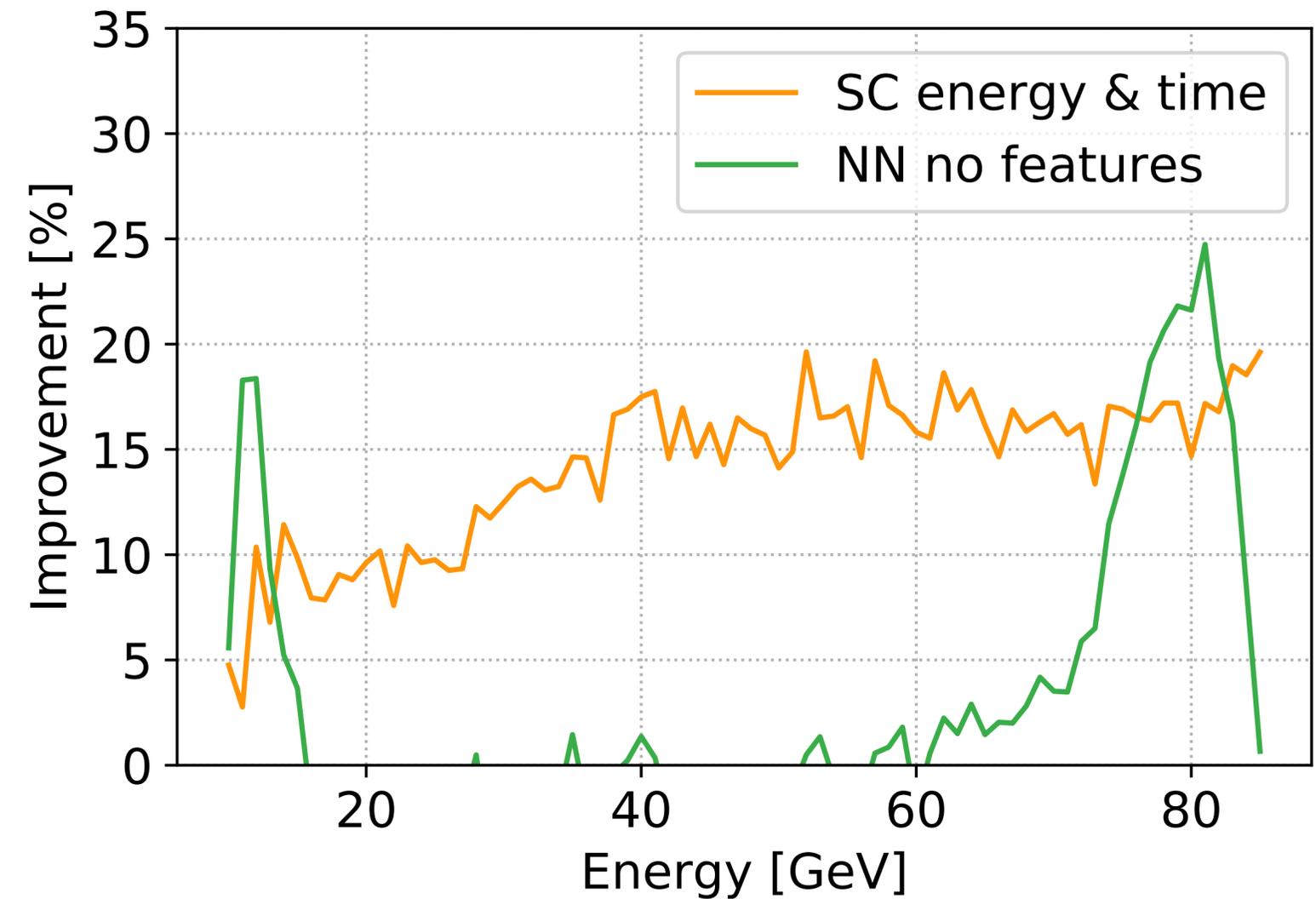
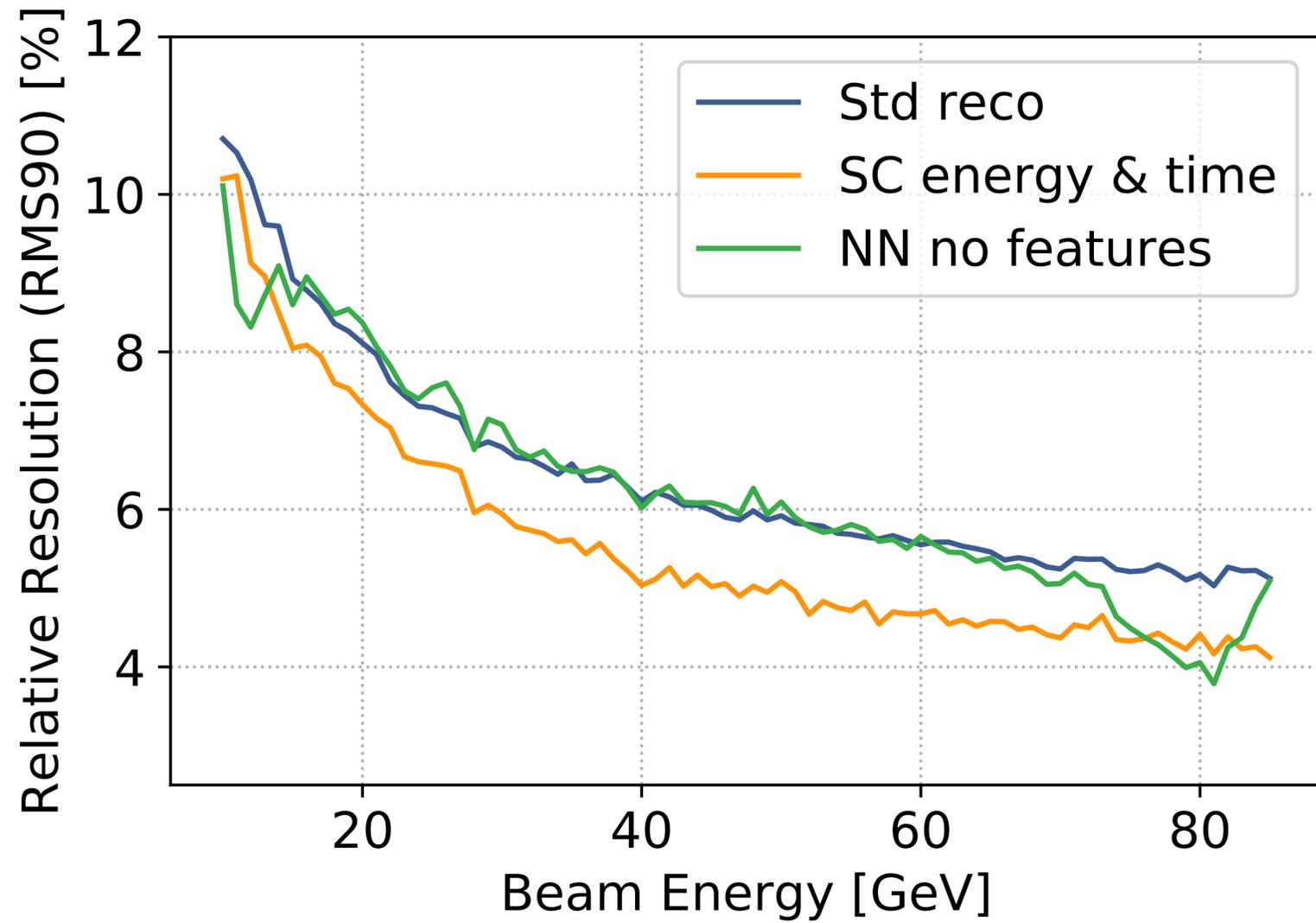
After SC



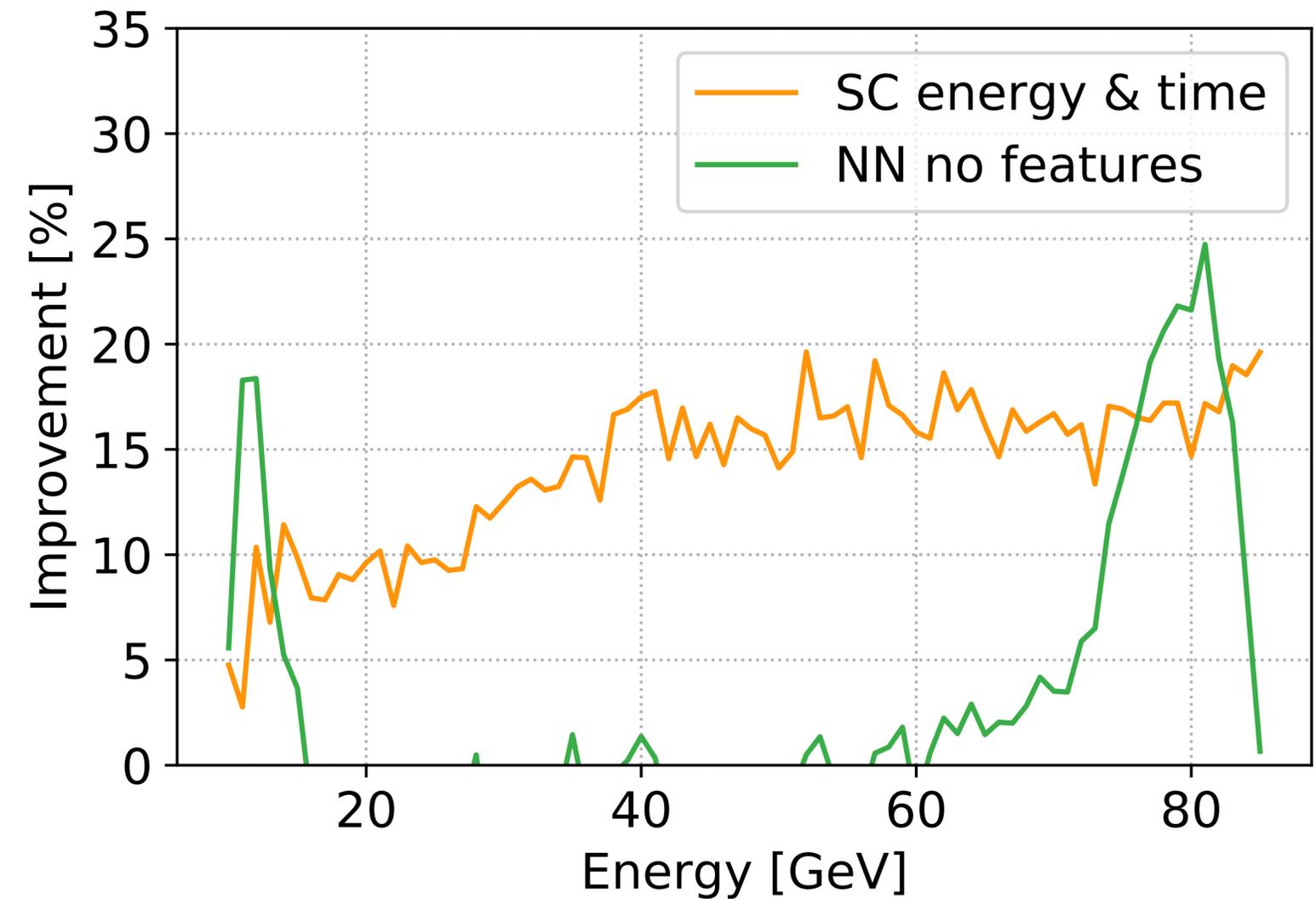
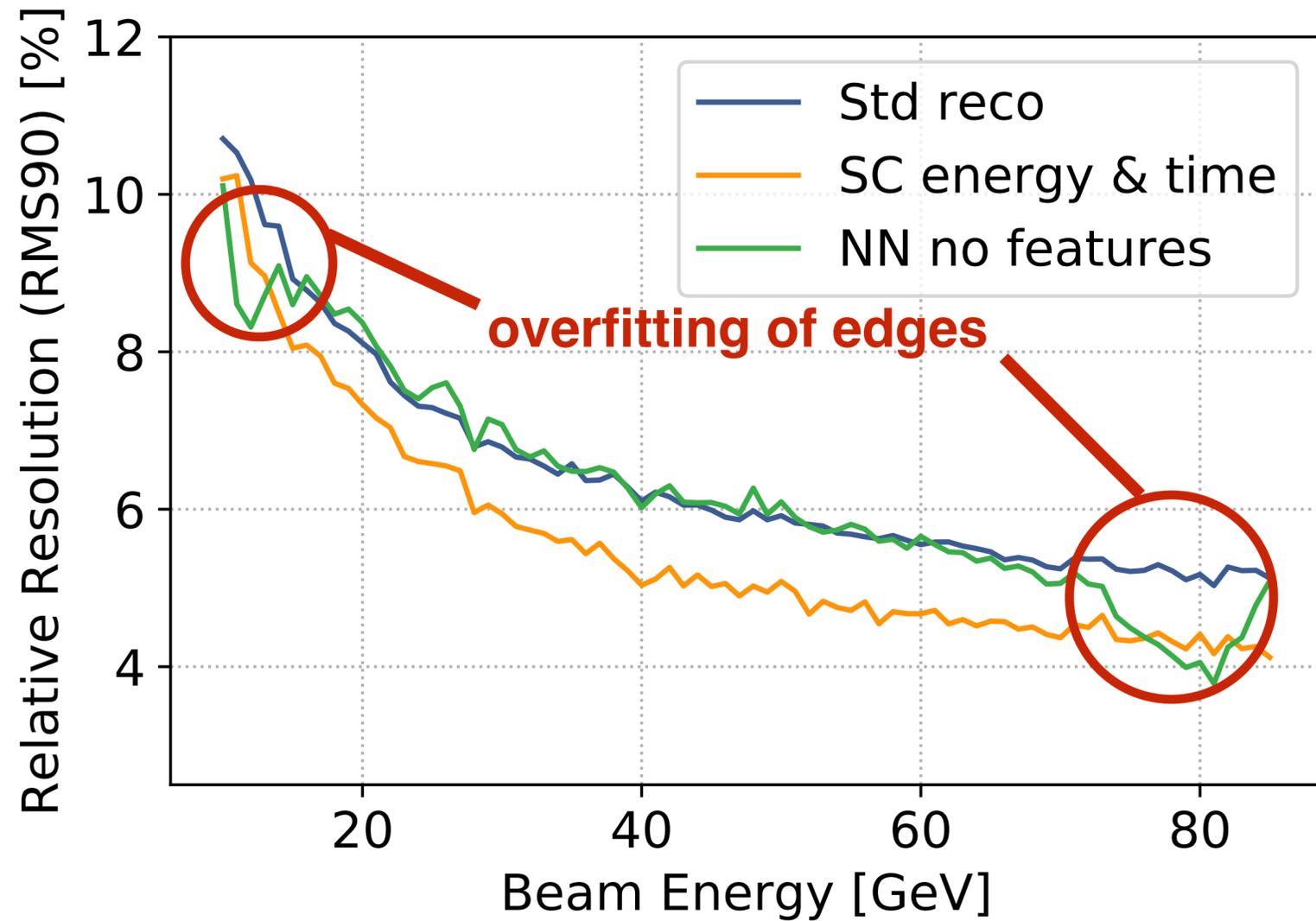
- Continuous training energies (1 GeV steps)
- Splitting data in train and test data set (70/30)
- Event selection must not depend on true energy
- Scaling of features (mean = 0, var = 1)
- Learning on  $dE = E_{\text{stdReco}} / E_{\text{true}}$
- Simple architecture:
  - 2 fully connected layers
  - 100 neurons each
- Prevent overfitting  
(small network size, dropout, l2 regularization)
- Loss function: Mean squared error



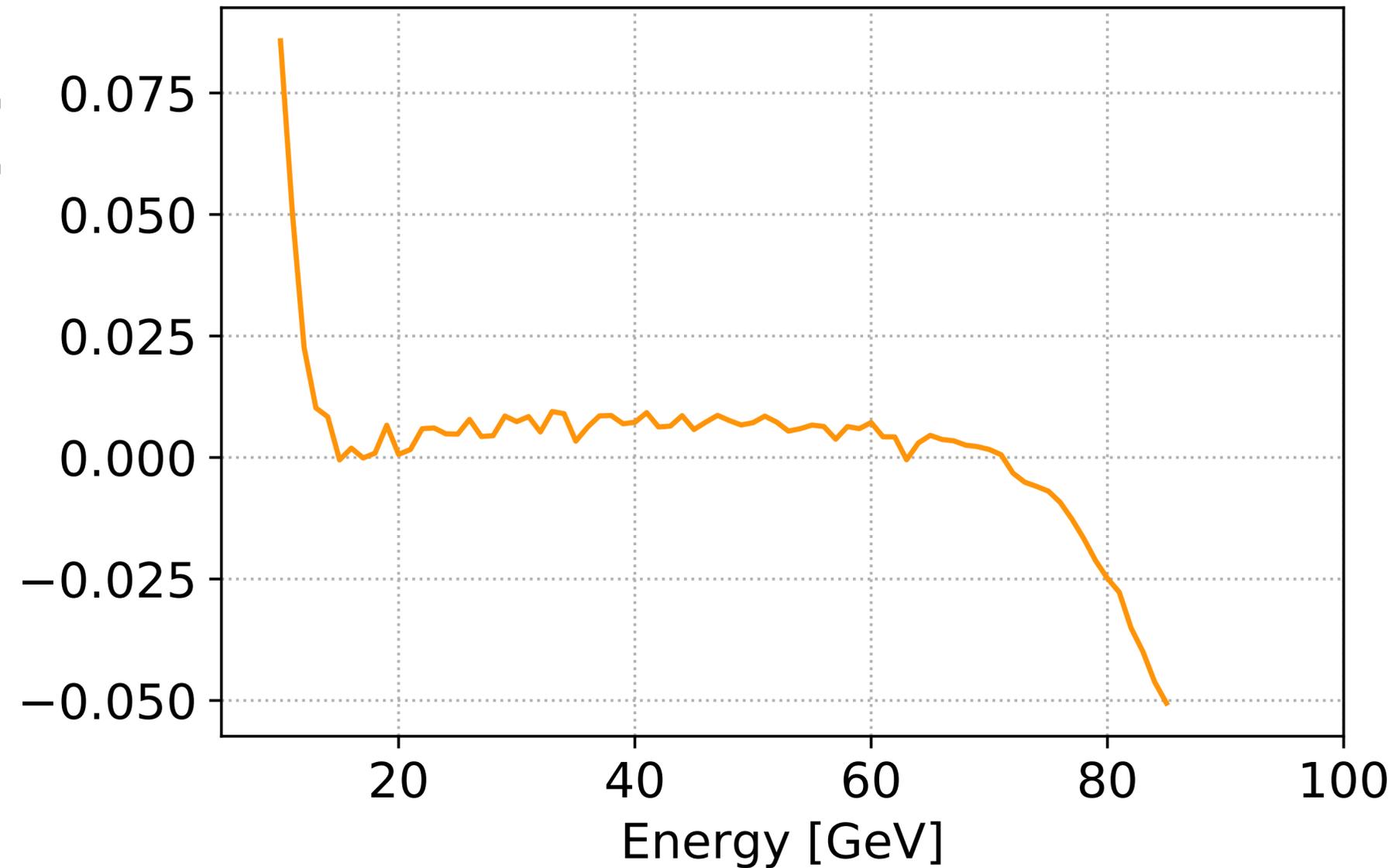
- **Input:** Only energy sum



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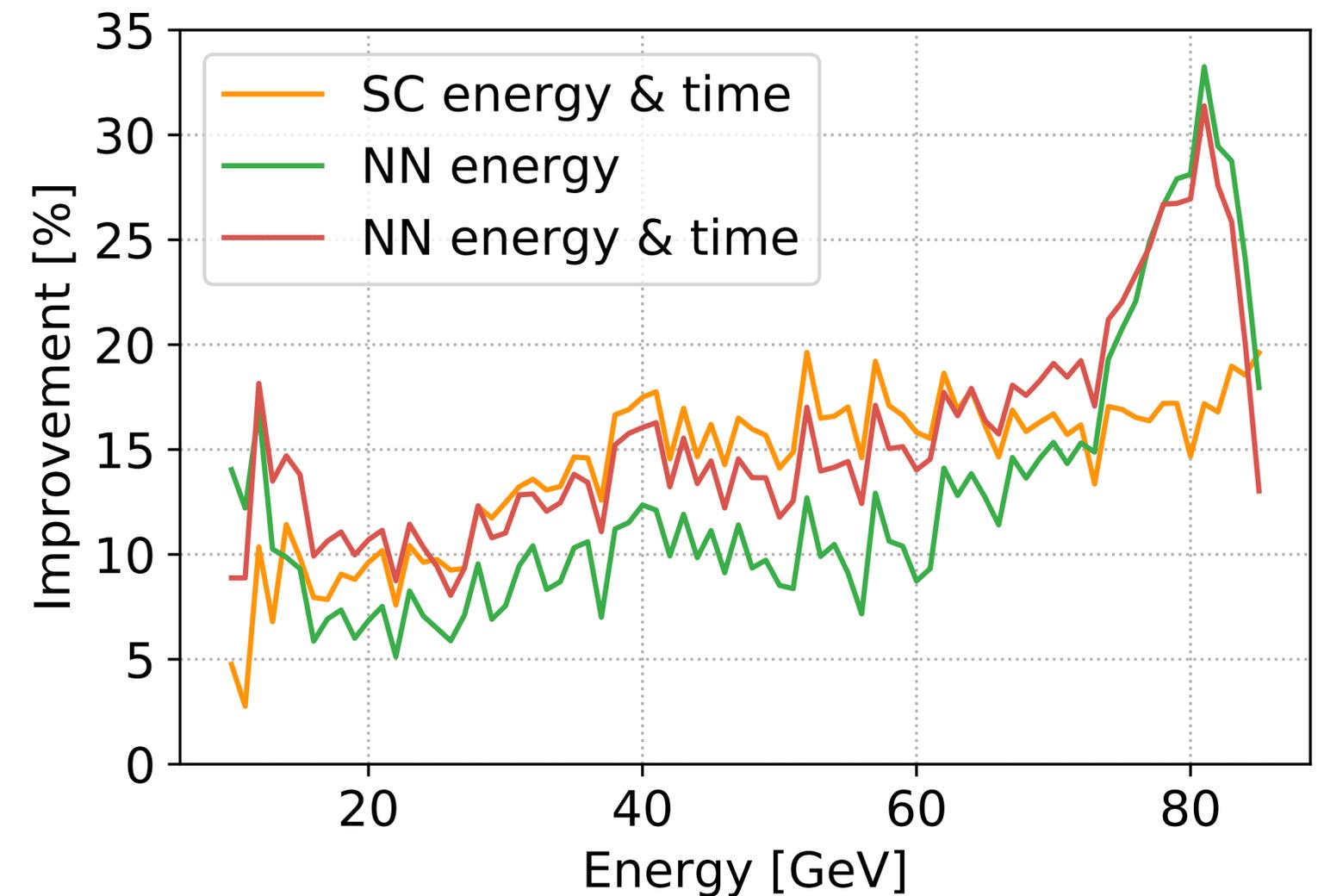
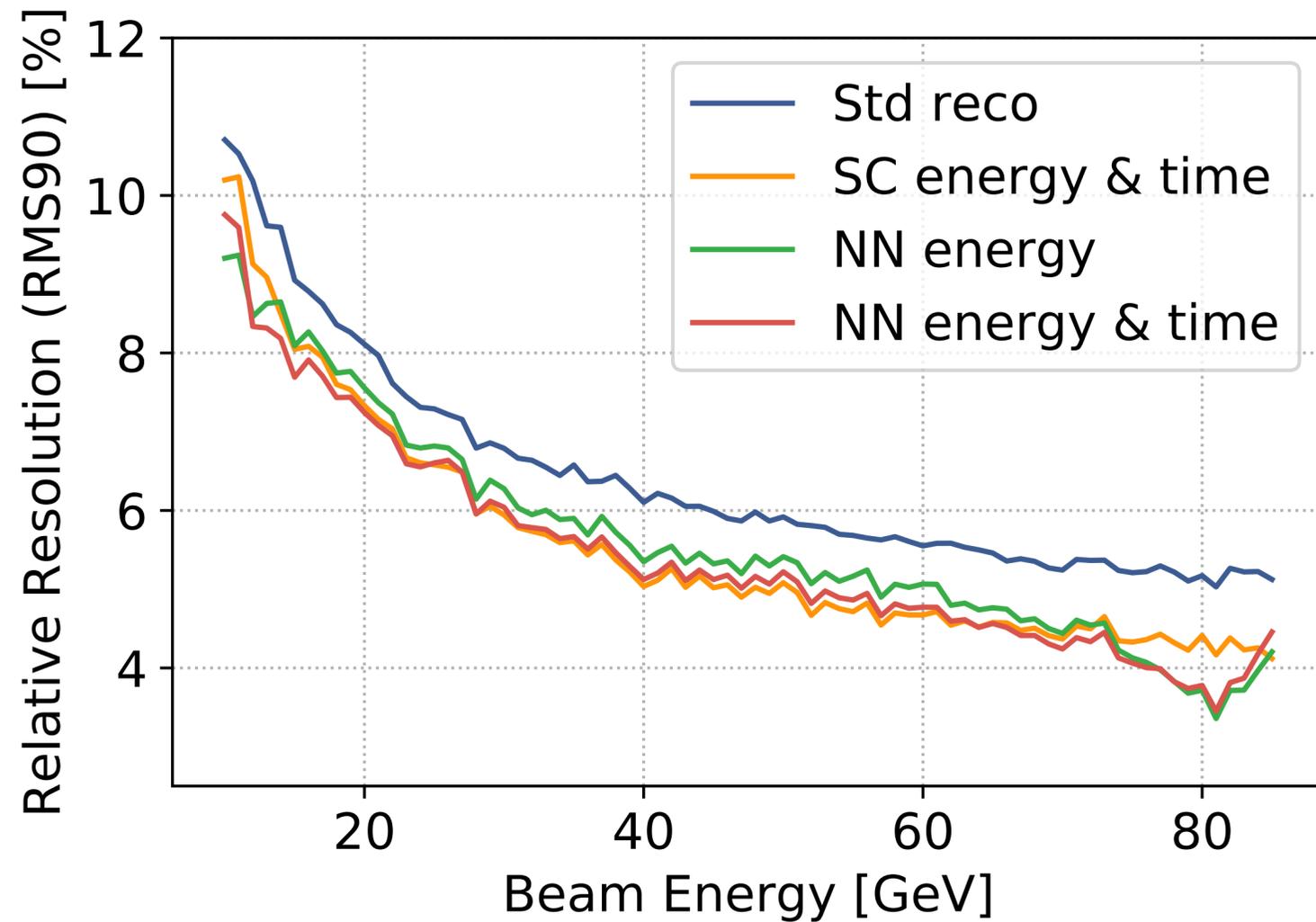
- Non-linearity at the edges



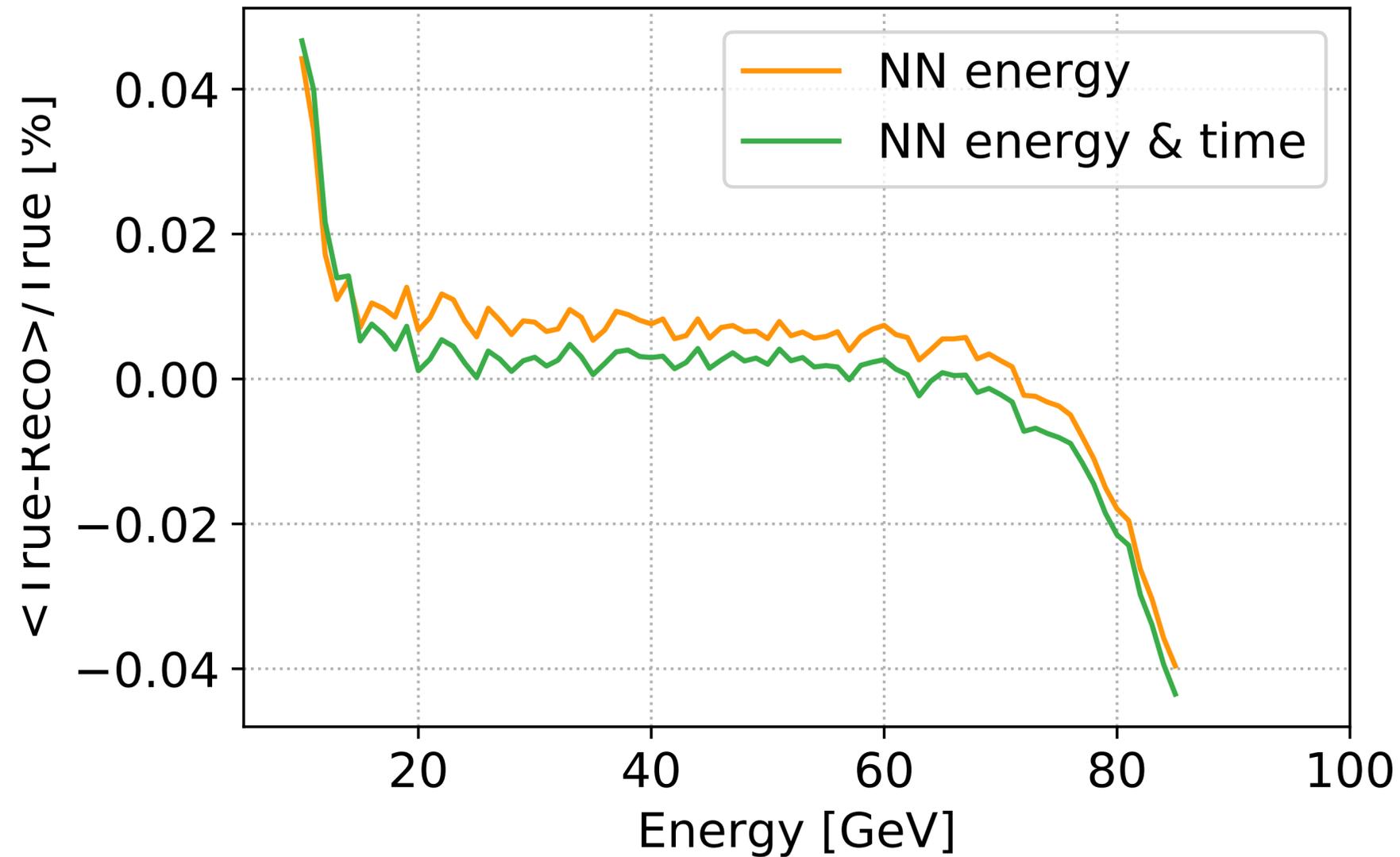
# NN Energy & Time



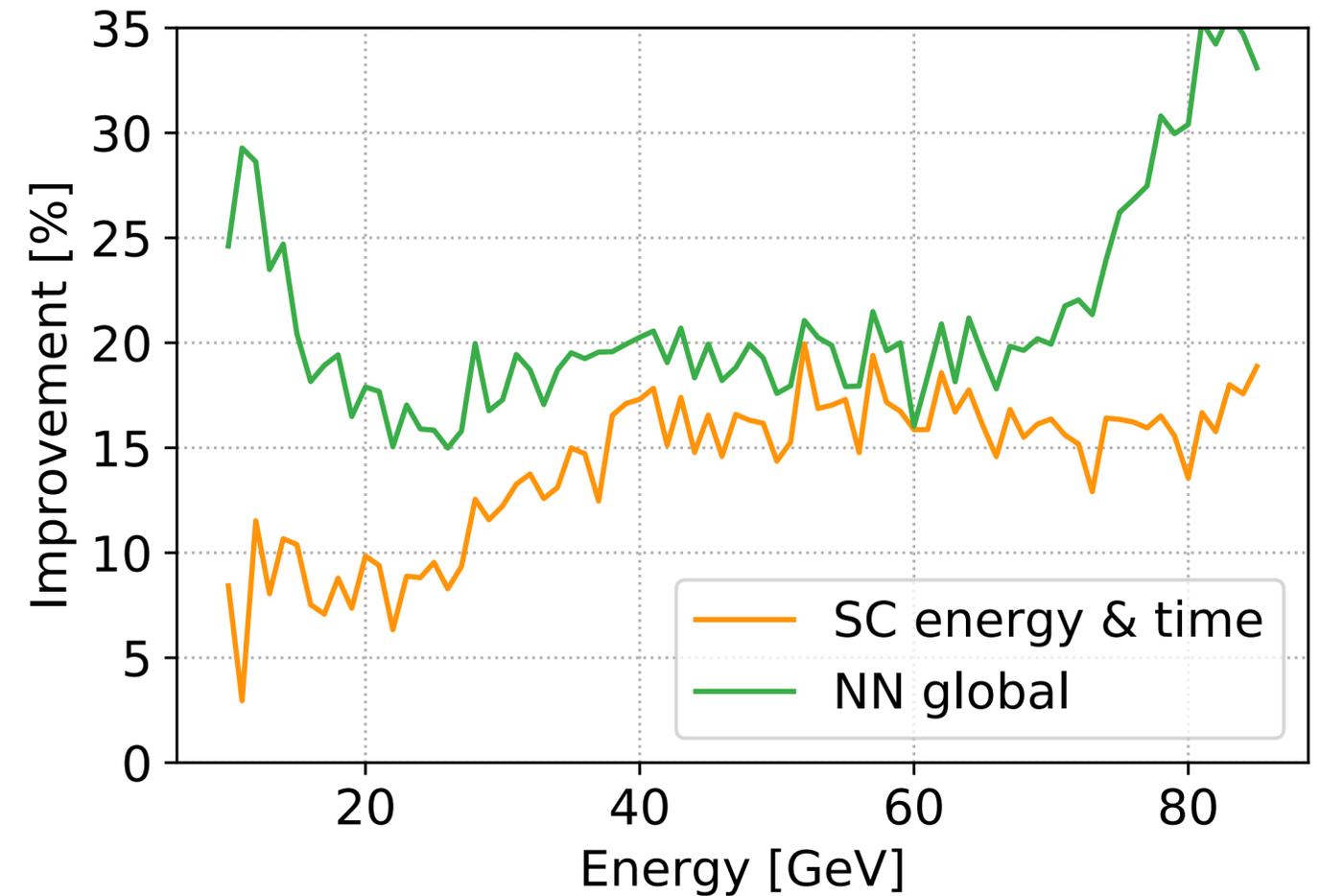
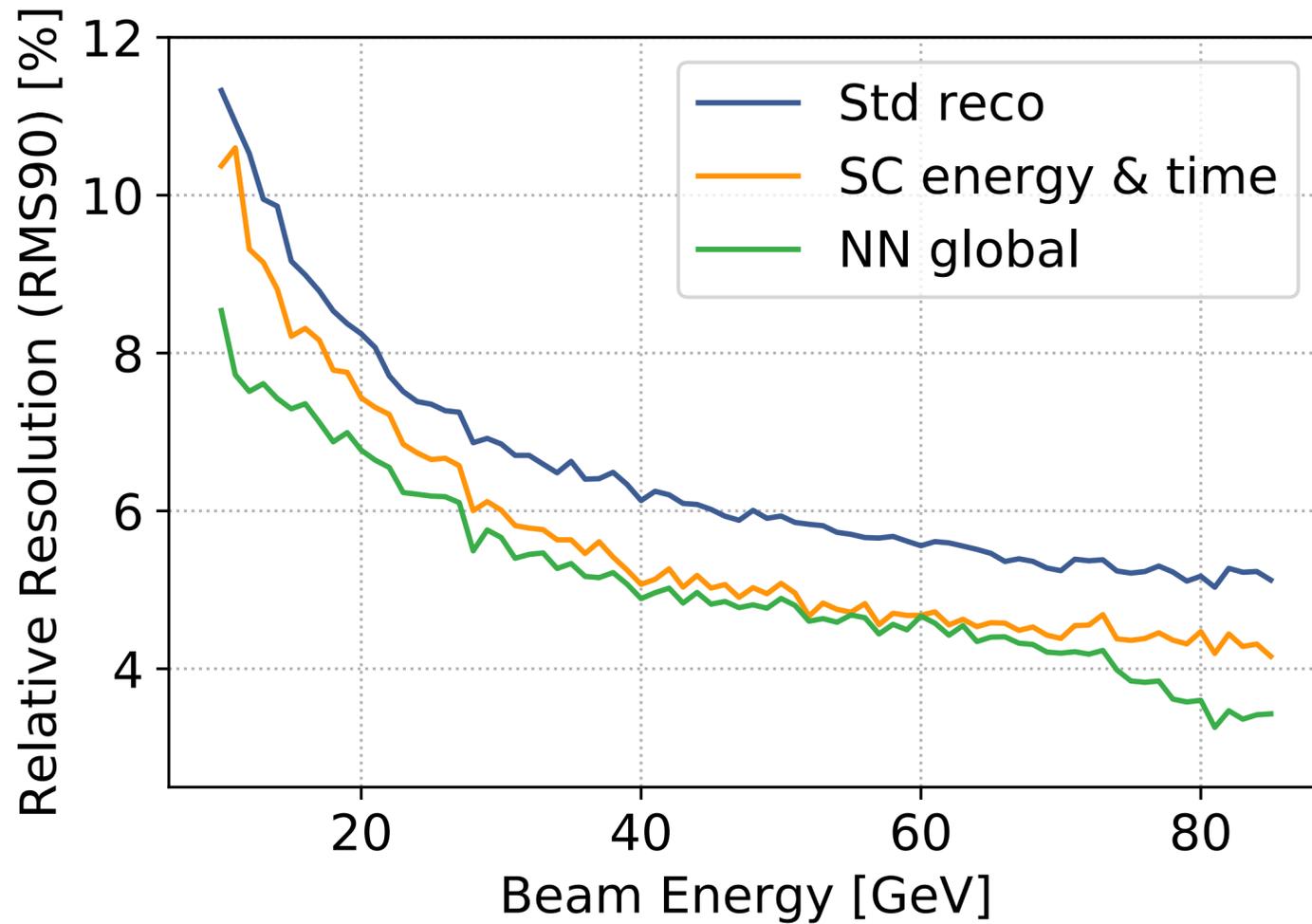
- **Input:** energy (#hits > 5mip), time (#hits > 5ns)



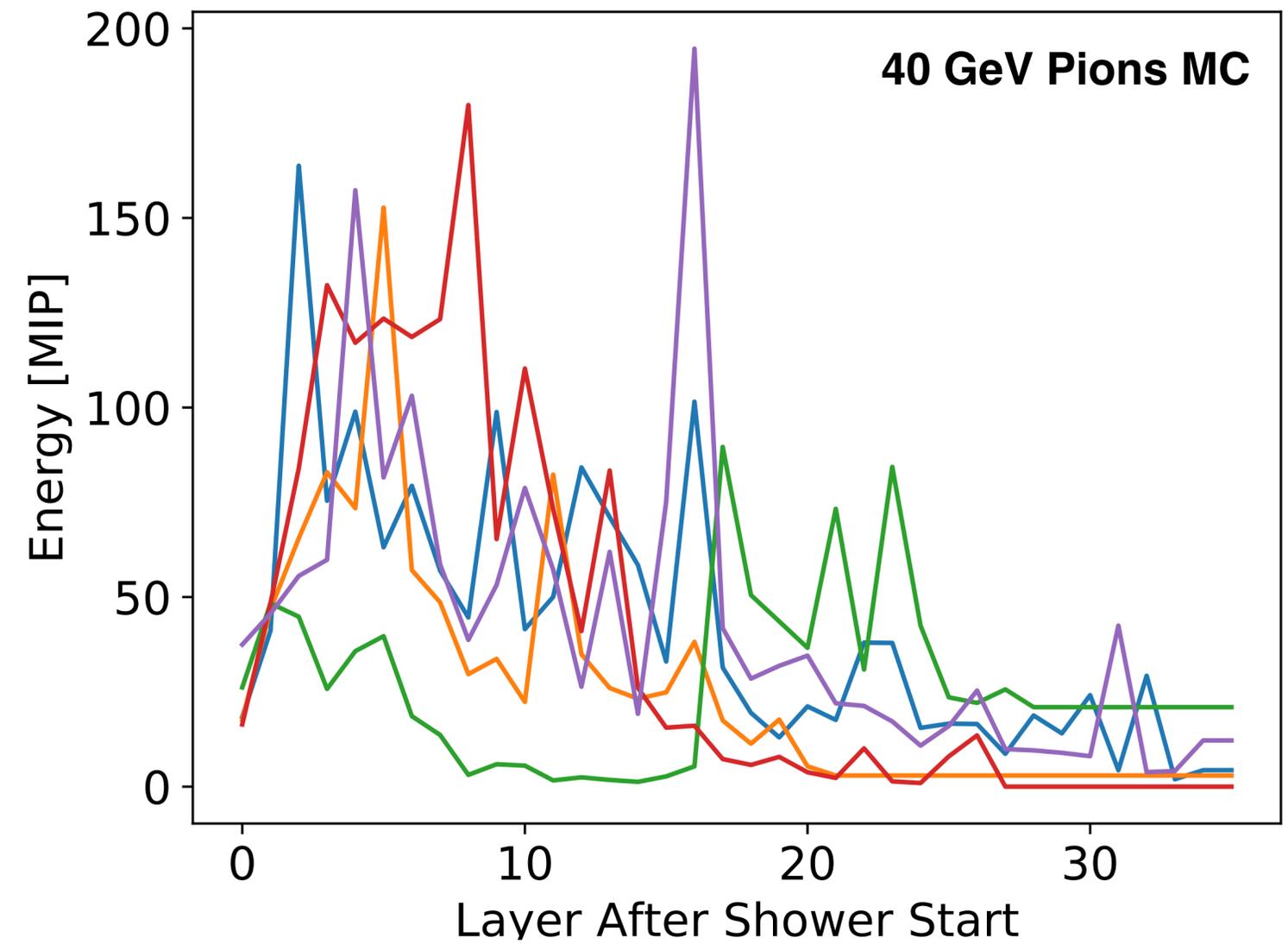
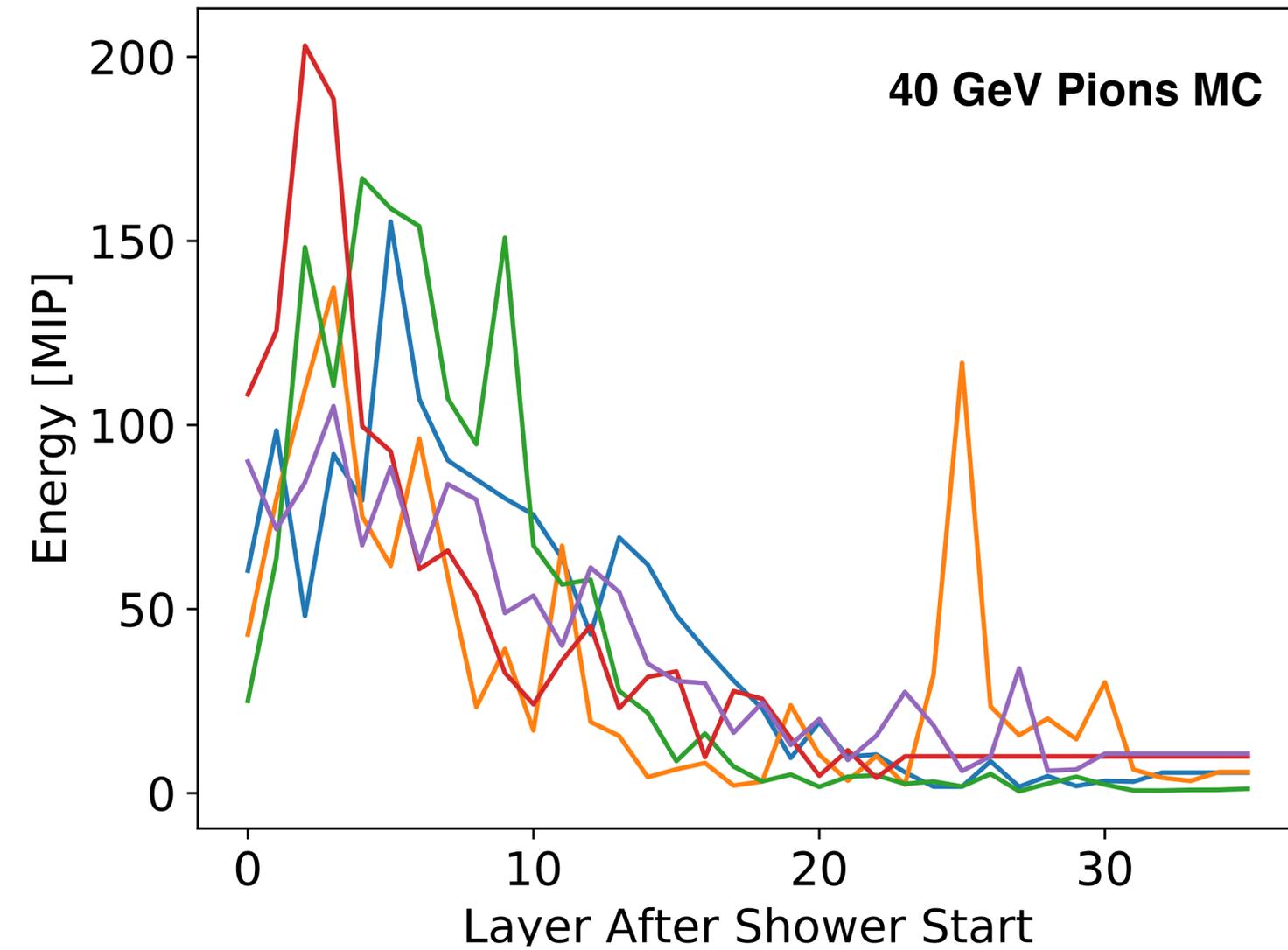
# NN Energy & Time - Linearity



- **Input:** energy (3 thresholds), time, #hits, radius, cogZ



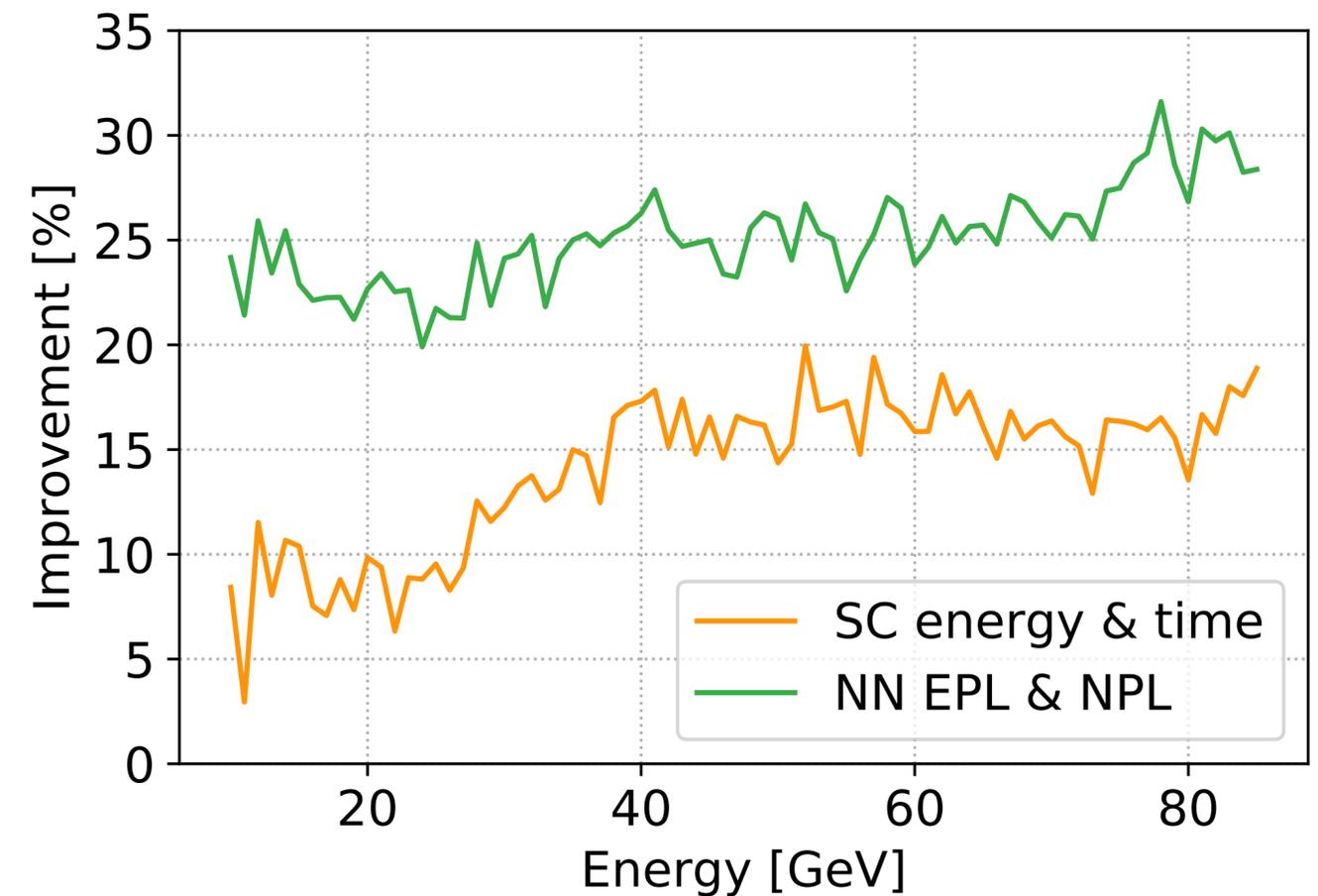
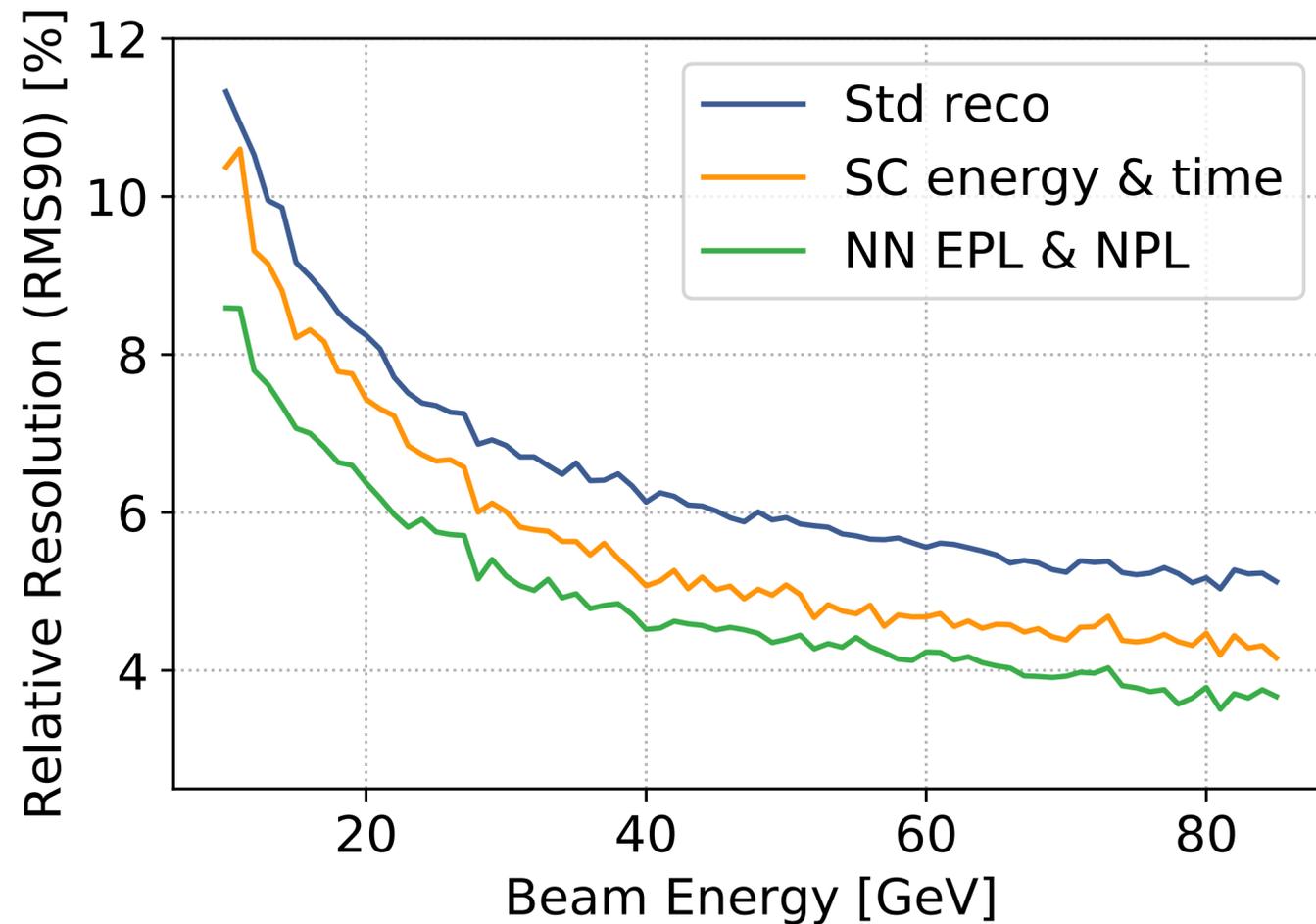
# Shower Profiles



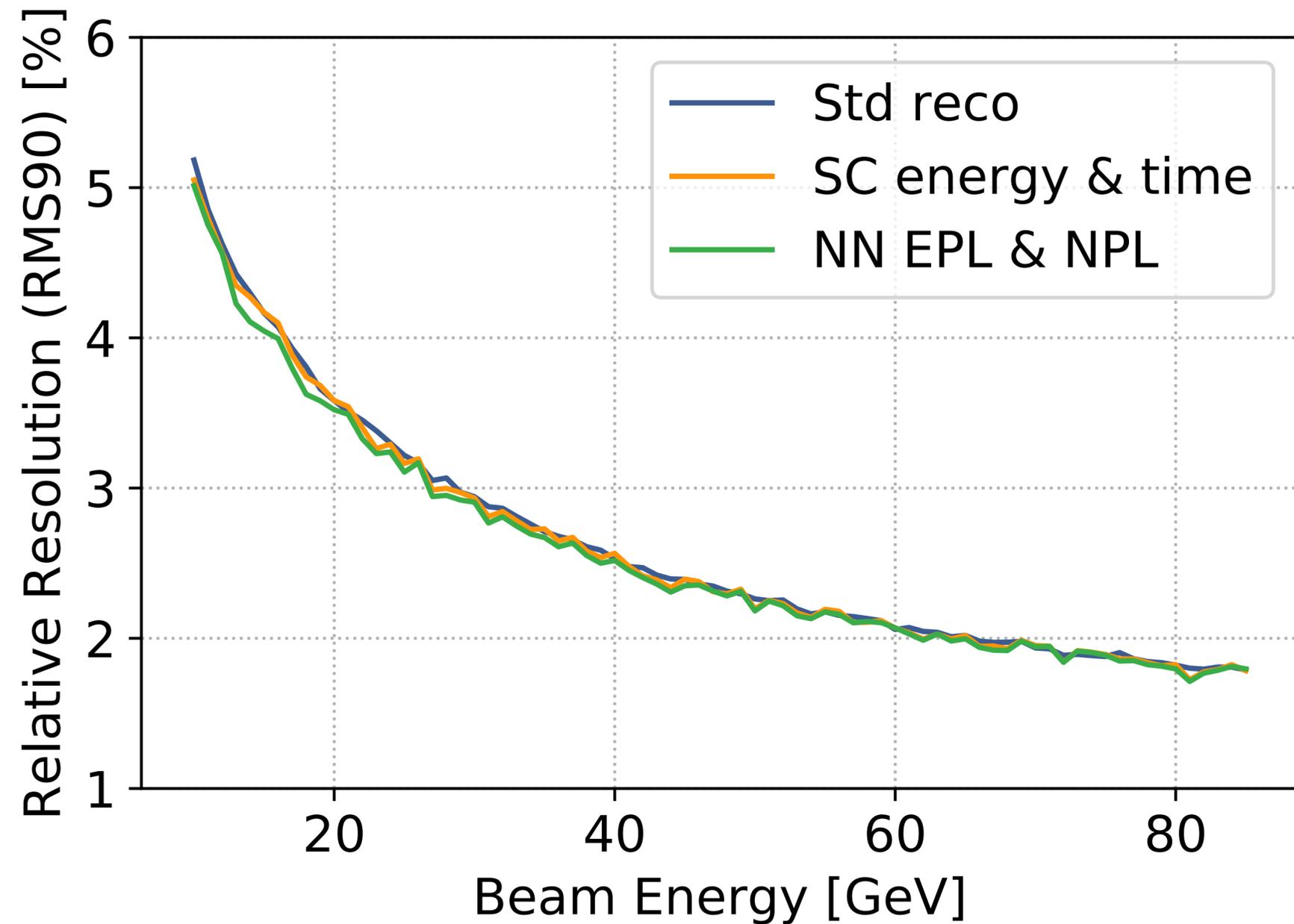


- **Input:** global, energy per layer, #hits per layer (total: 8+80 inputs)

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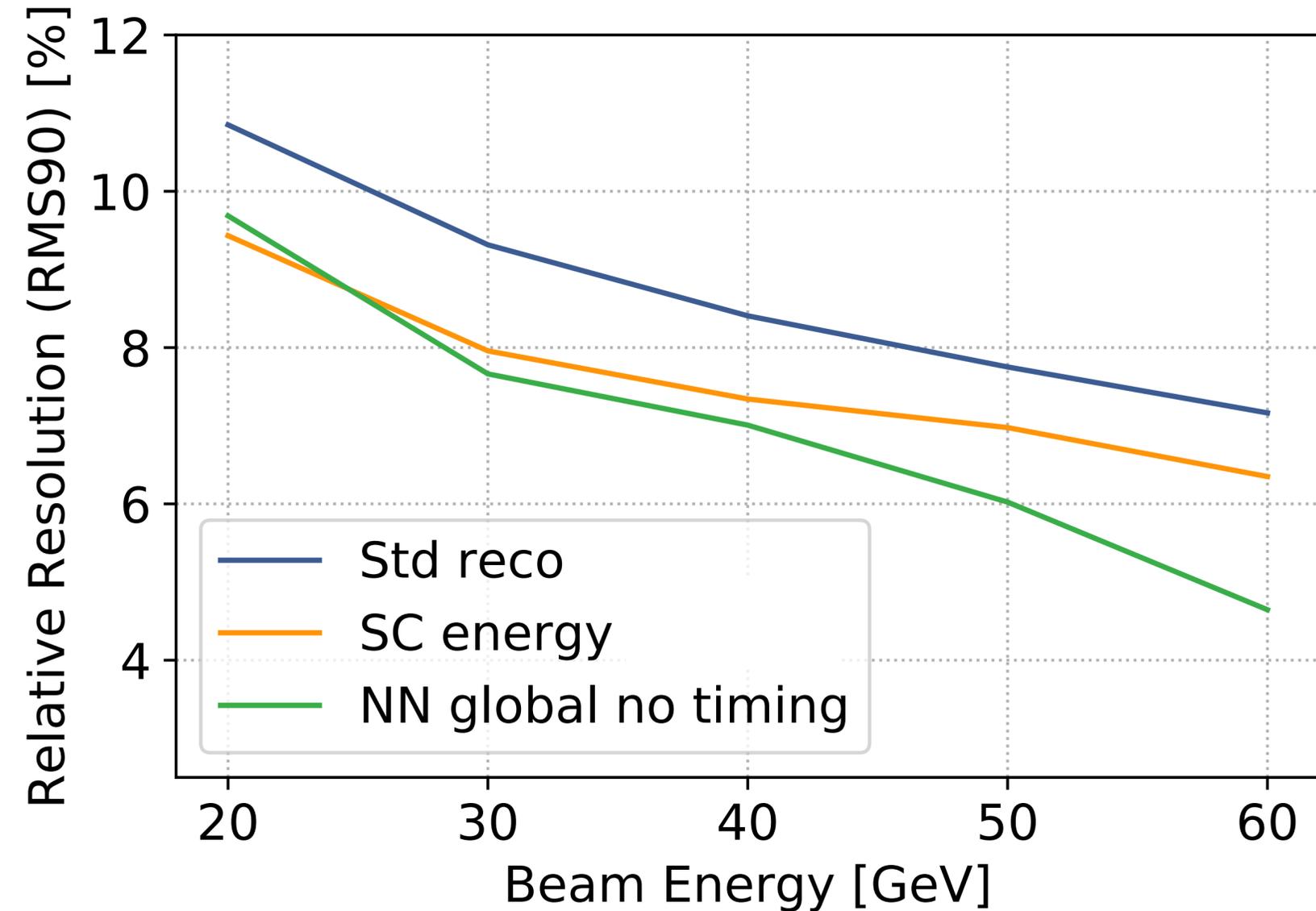
- Cross check with electrons
- No improvement visible



# Does it work on data?



- Training on 60 layers simulation, applying to data (20, 30, 40, 50, 60 GeV)
- Very preliminary: Insufficient event selection
- **Clear improvement observed**



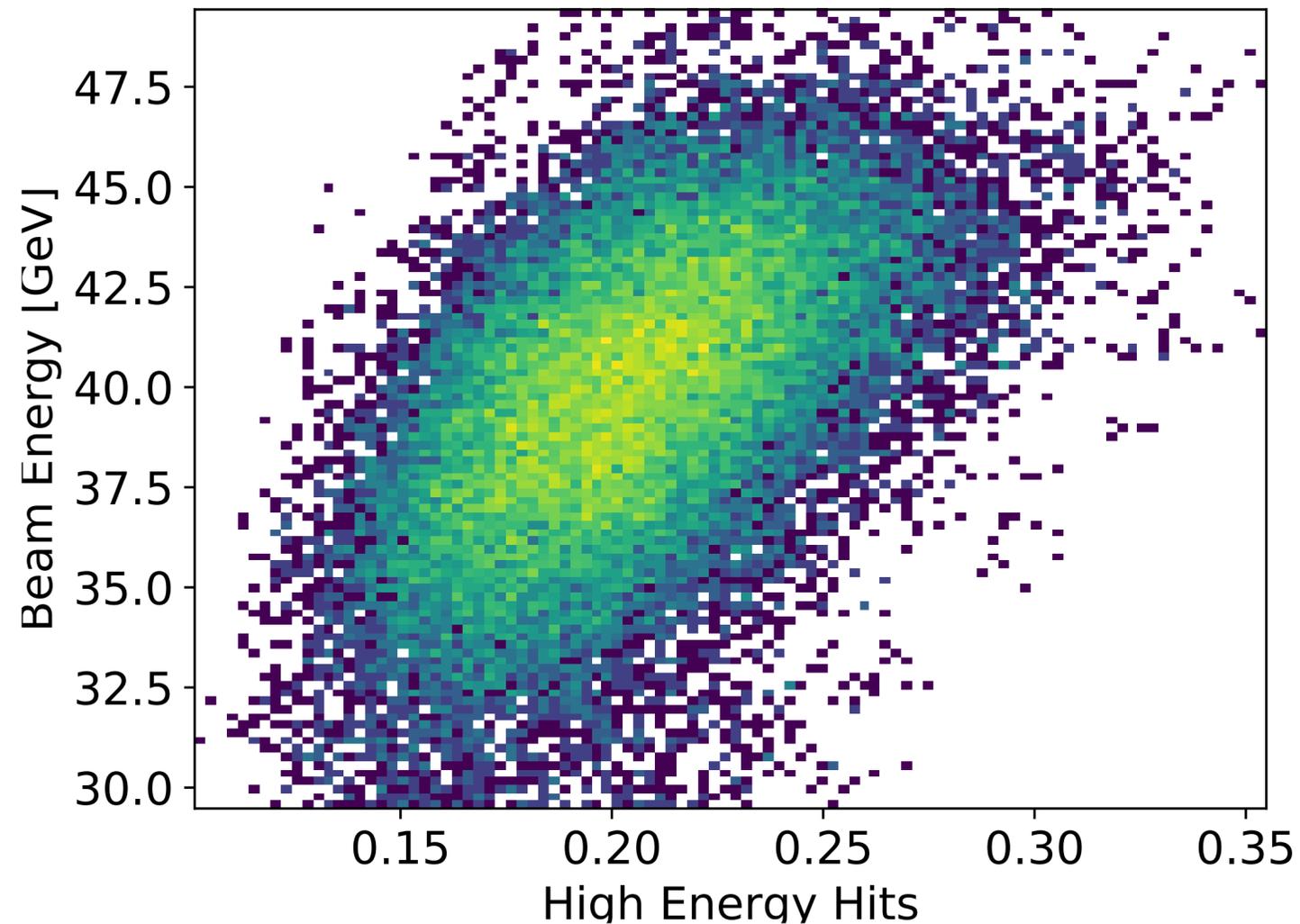
- Correlation of timing with energy visible in simulation
- Slight improvement in energy reconstruction in SC scheme
  - Good time resolution needed
  - Even harder in real-life: clustering
- Machine learning algorithms are capable of using much more information
  - Significant improvement for using energy profile of each event
  - Adding information improves energy reconstruction
  - First check on data looks promising
- **Outlook:**
  - Timing in tungsten?
  - Better architecture, better features, better event selection?
  - Deep learning (Erik)



# High Energy Hits Correlation

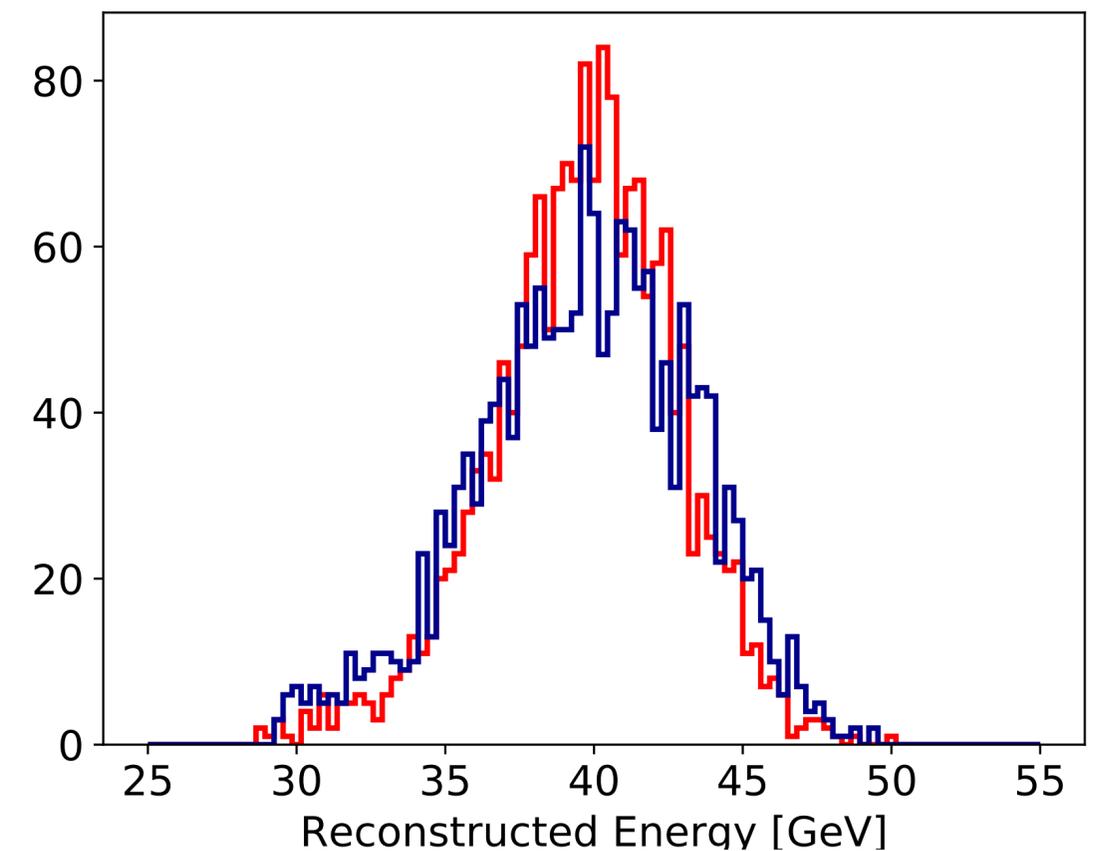


- Clear correlation of high energy hits > 5 MIPs and energy sum visible

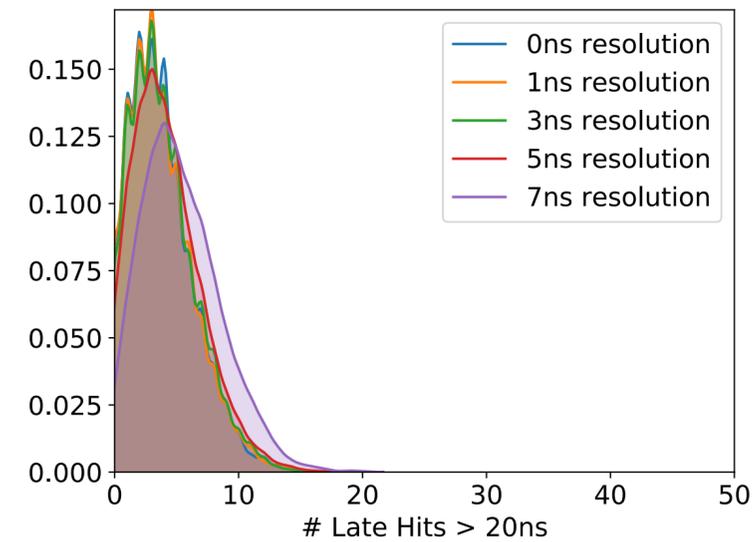
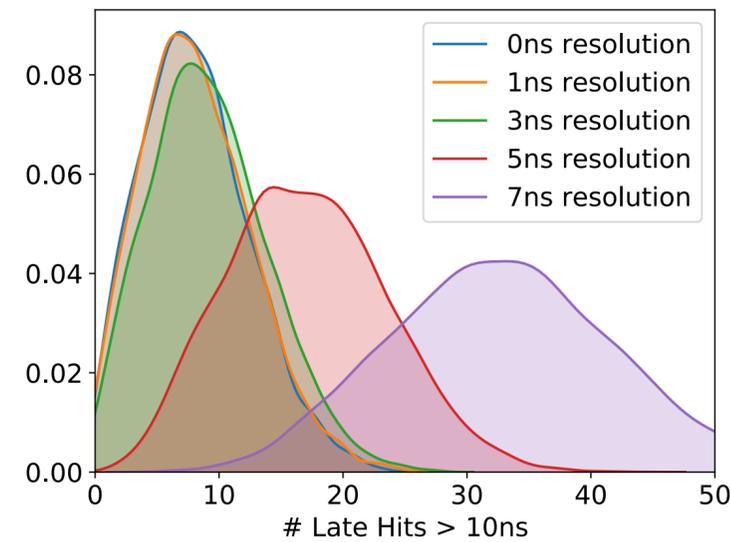
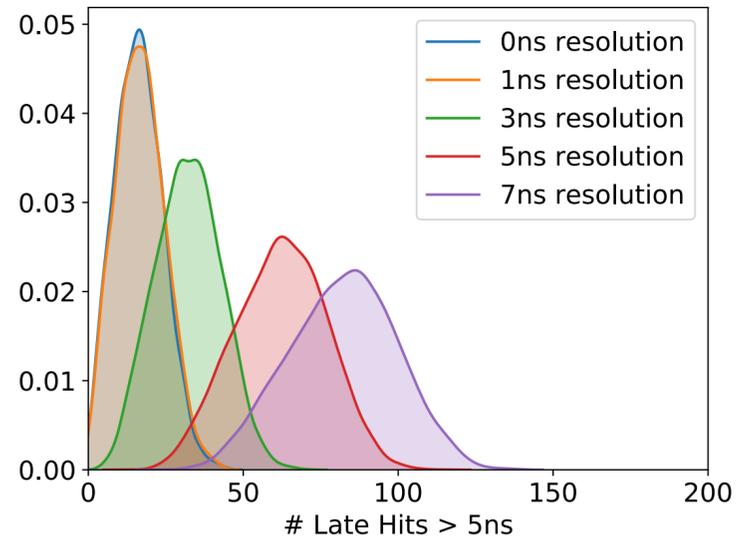
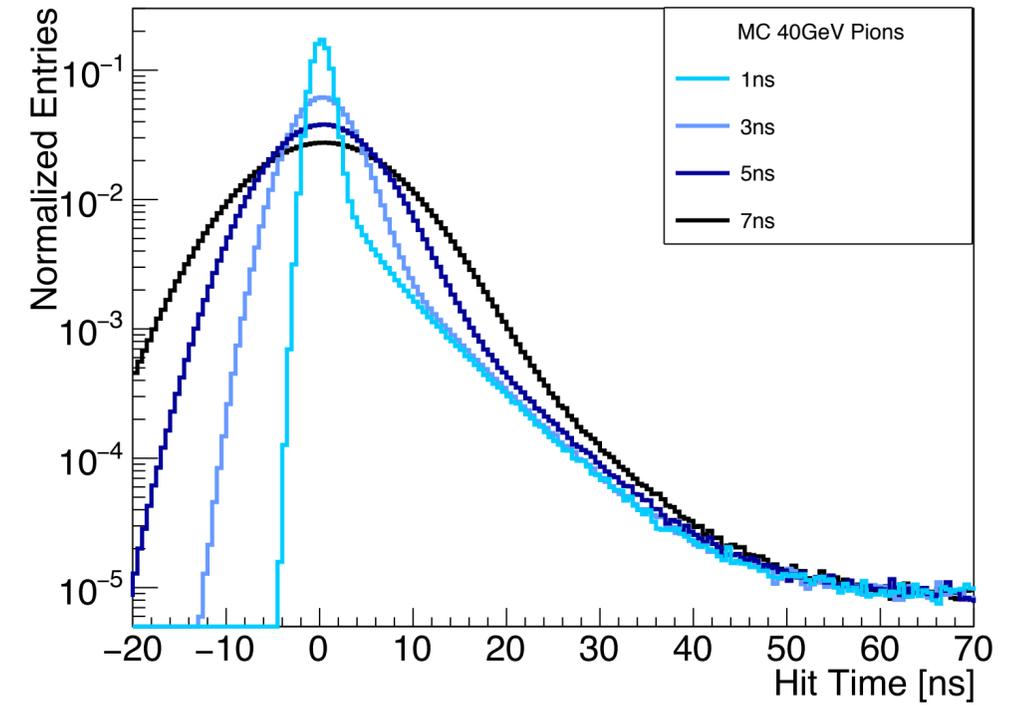
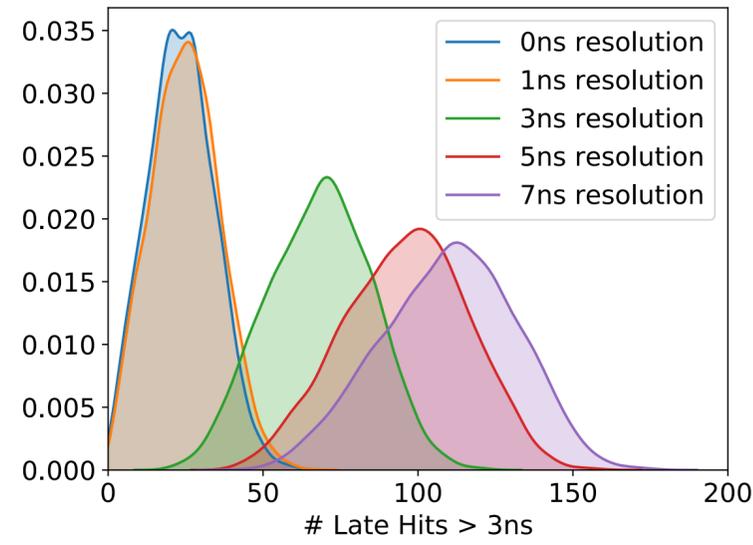
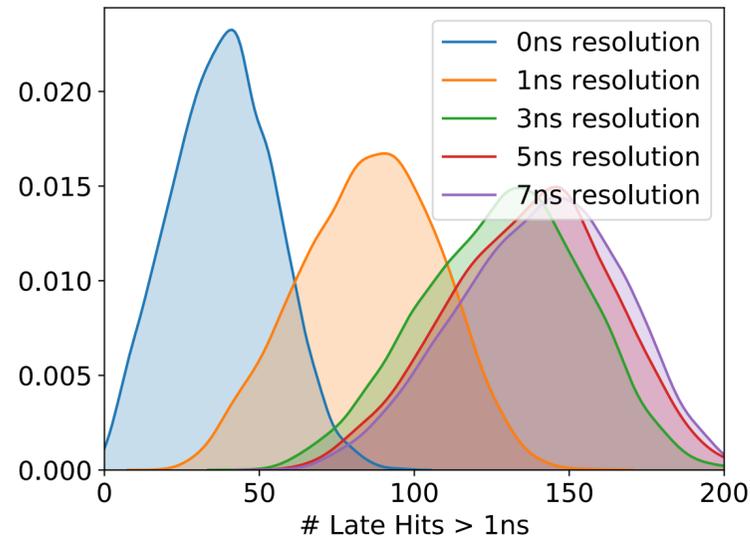


$$E_{\text{reco}} = a + b * f_{\text{HighEnergyHits}} + c * f_{\text{HighEnergyHits}}^2$$

- Global software compensation approach yields **15.2%** improvement over std reco



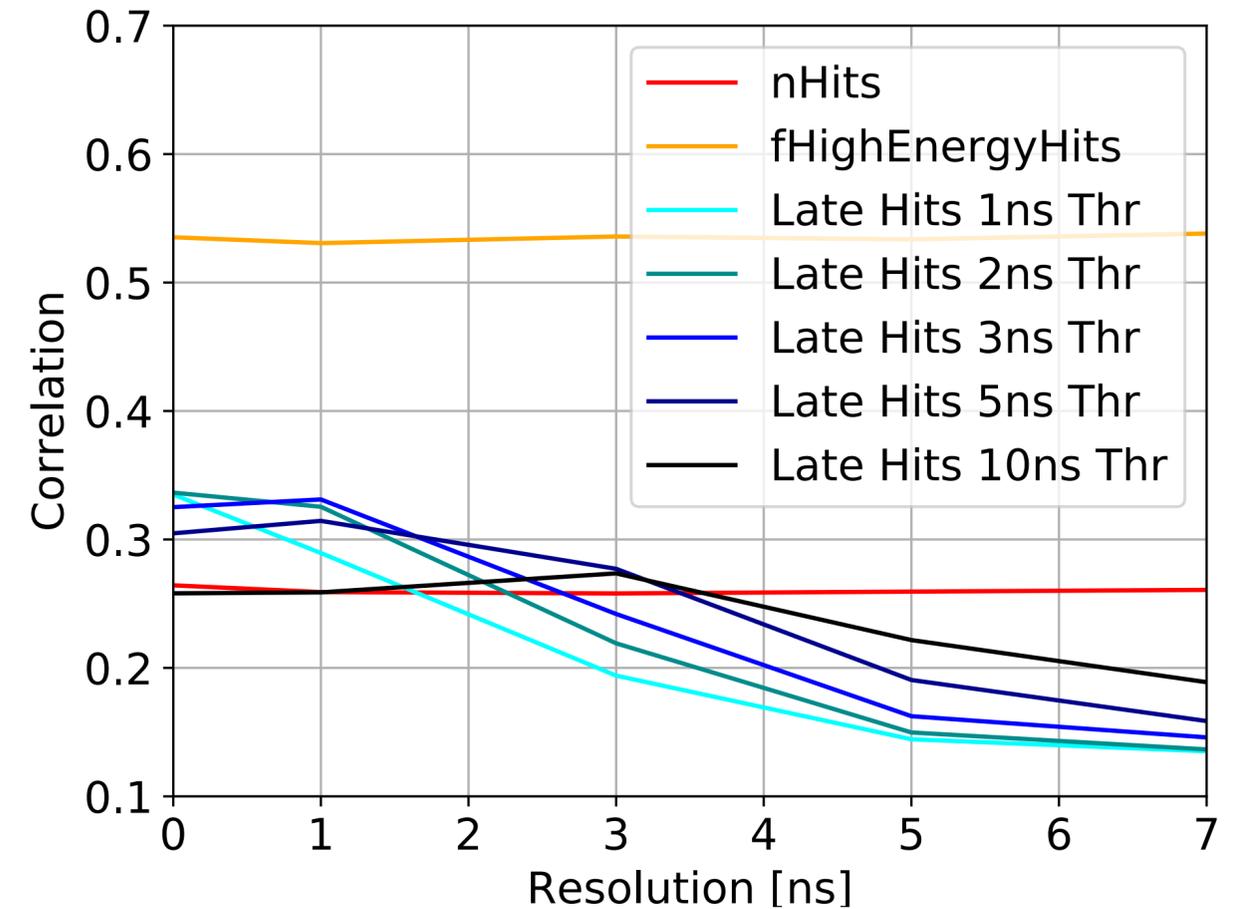
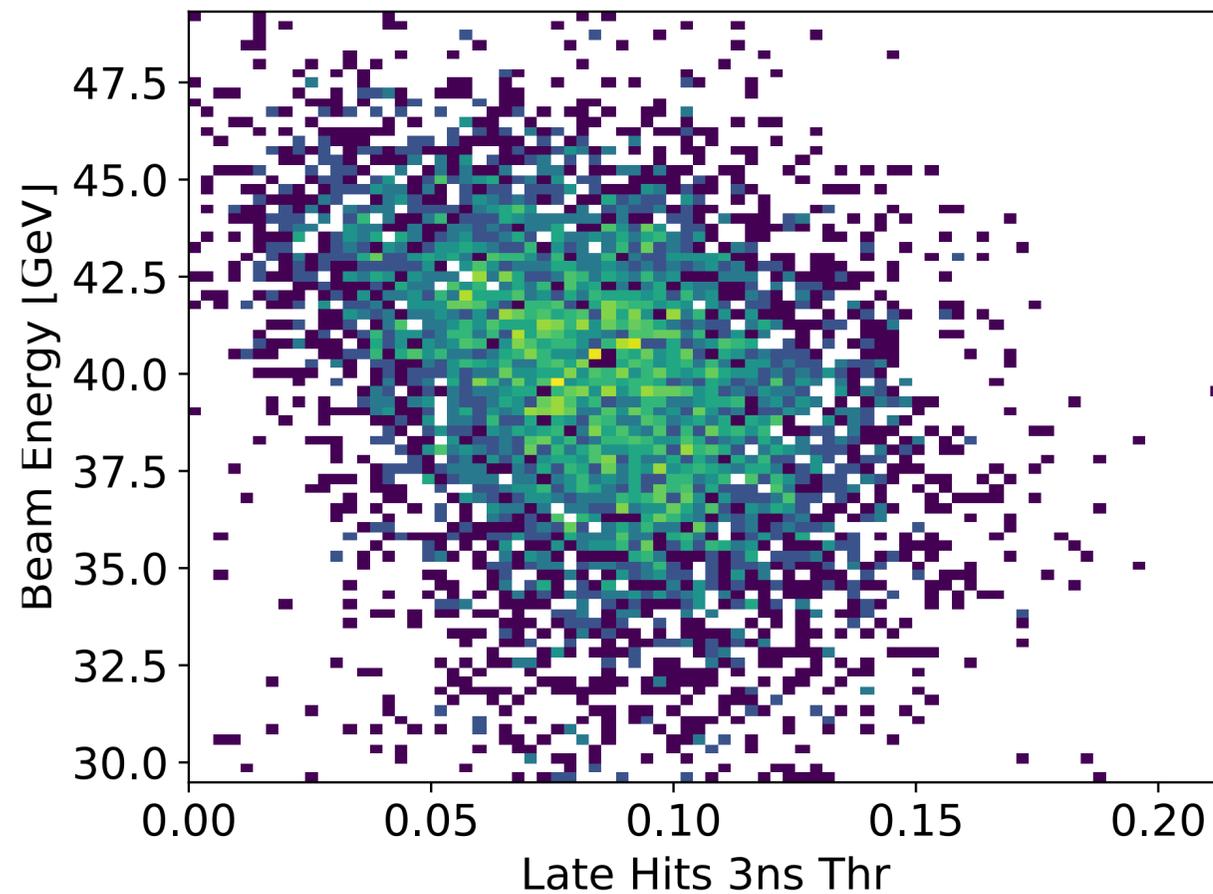
# Late Hits



# Fraction of Late Hits



- Correlation between fraction of late hits and energy sum visible



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