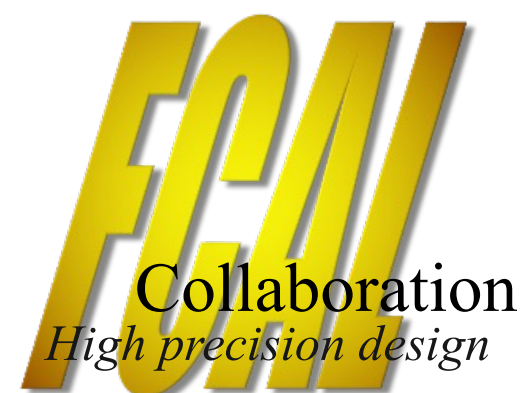




# ECAL Performance Towards the experiment

Shan Huang (Tel Aviv University)  
on behalf of the ECAL workforce

LUXE Workshop 2023

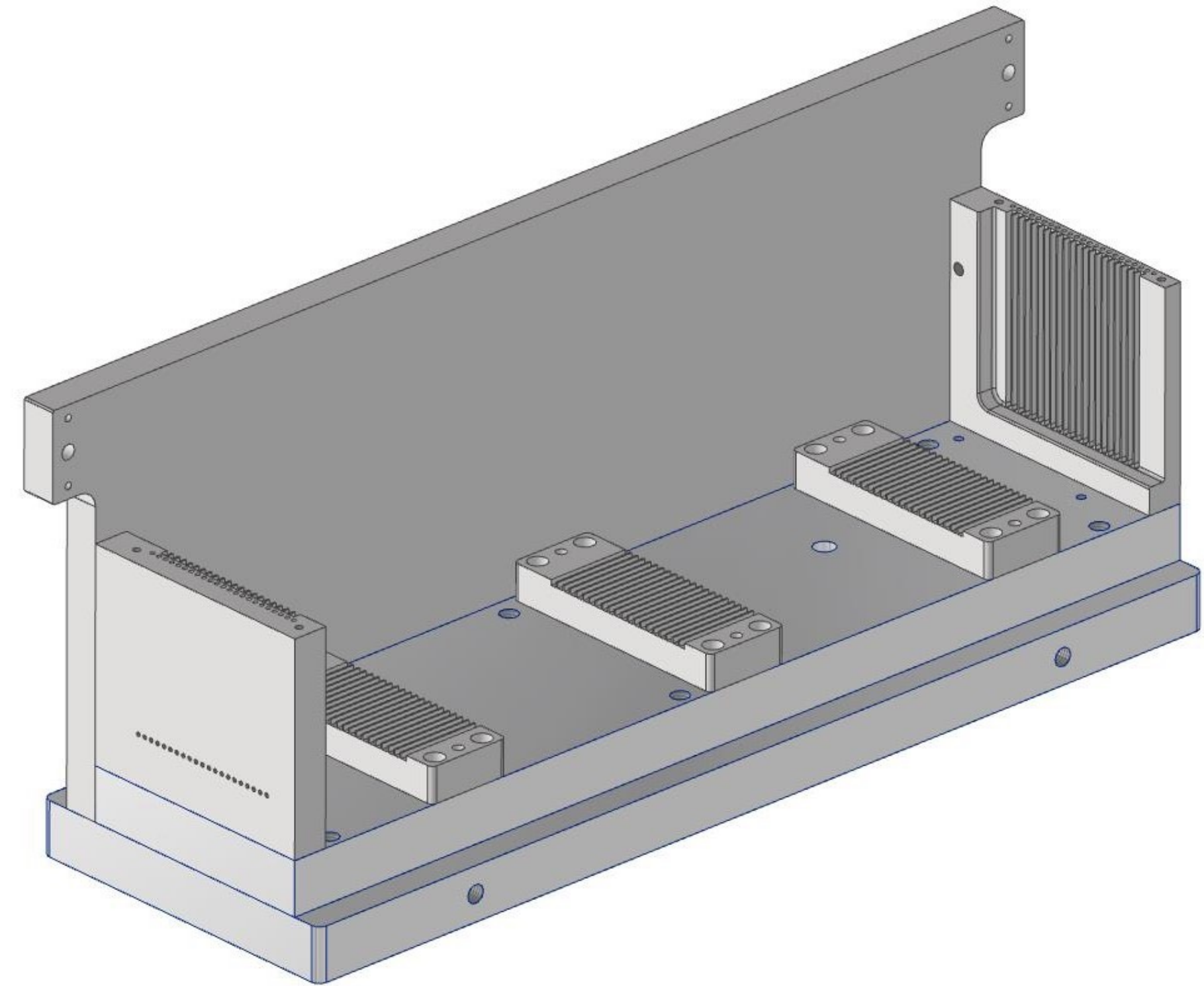


5 September 2023

# People on ECAL Performance

(Alphabetically)

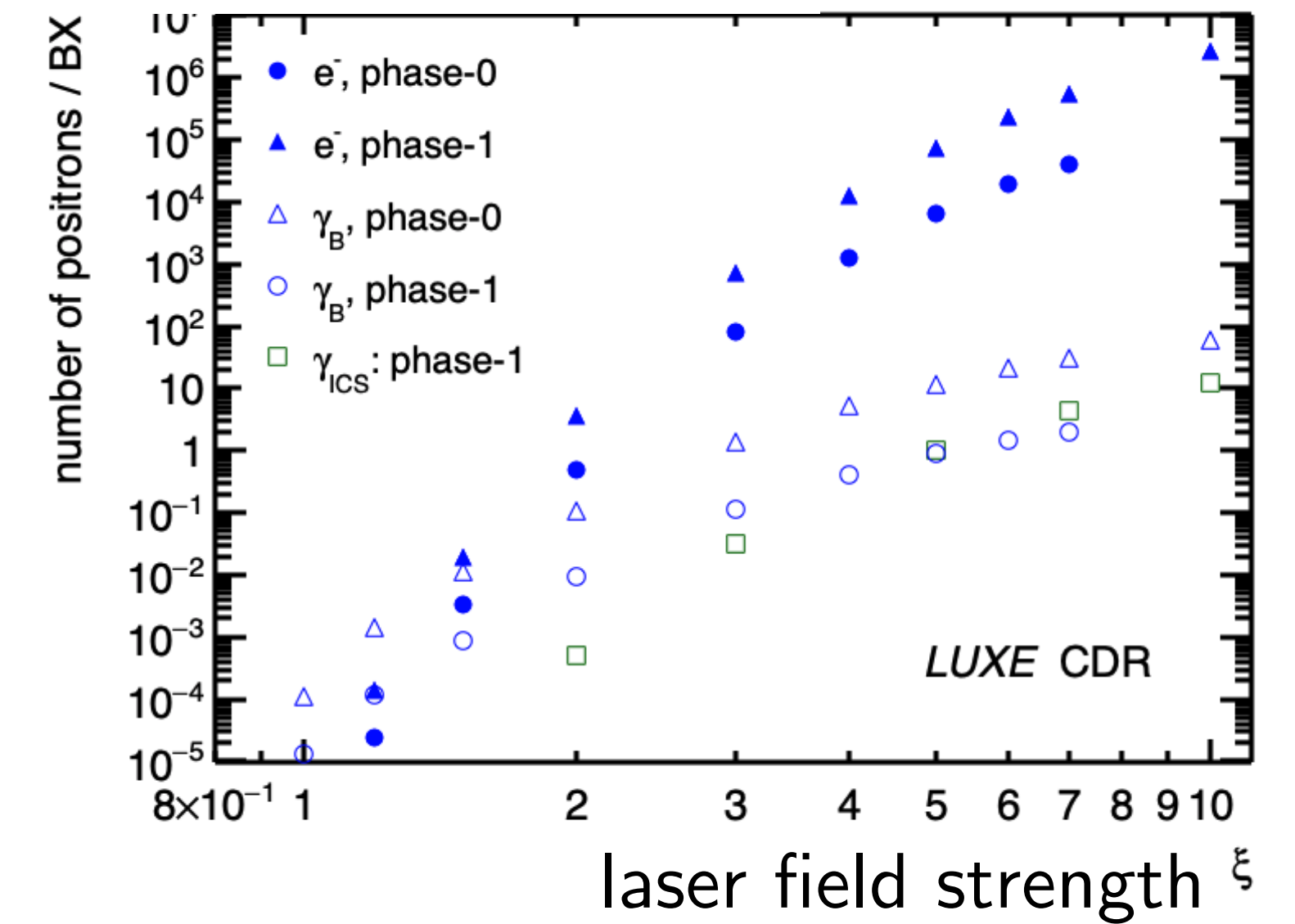
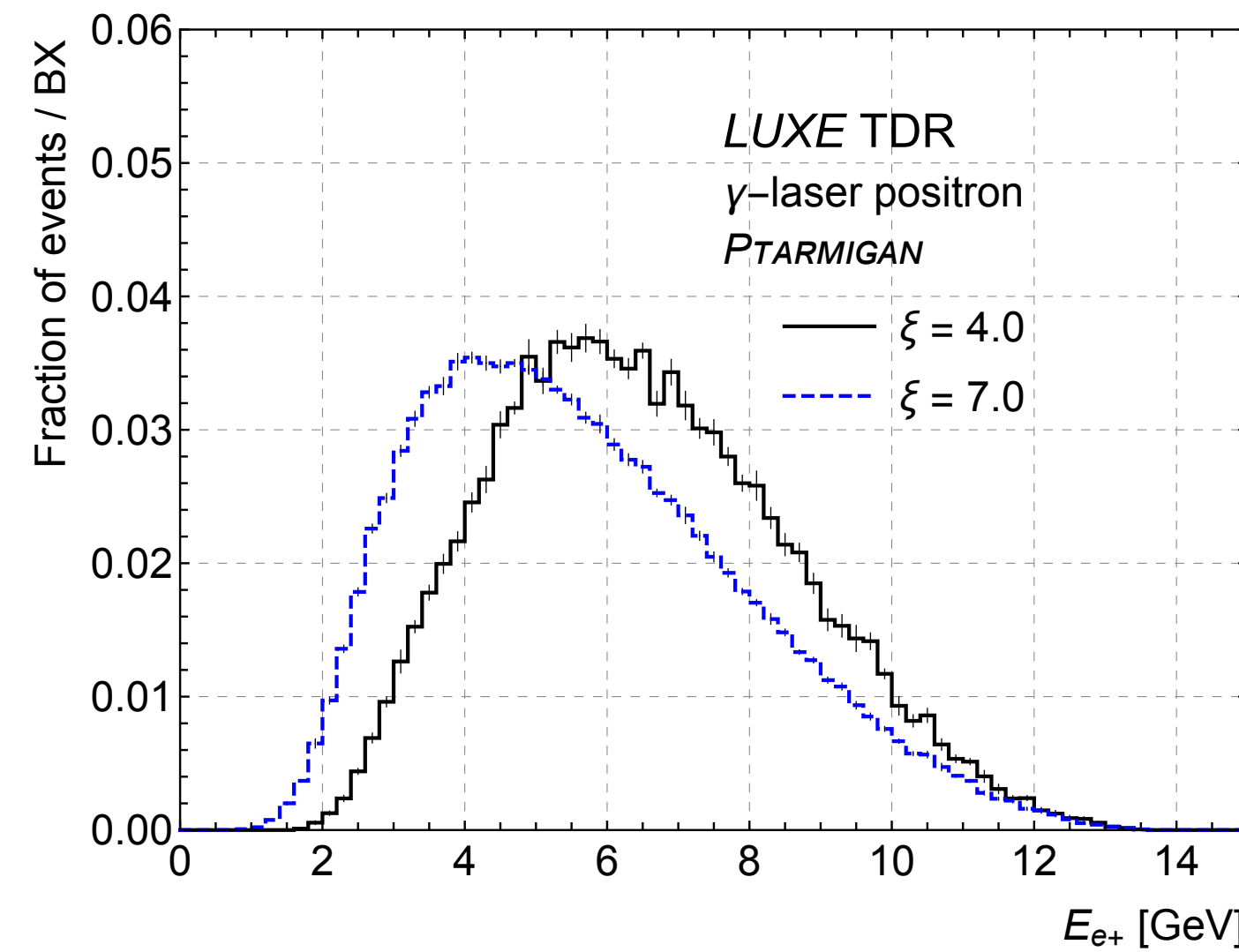
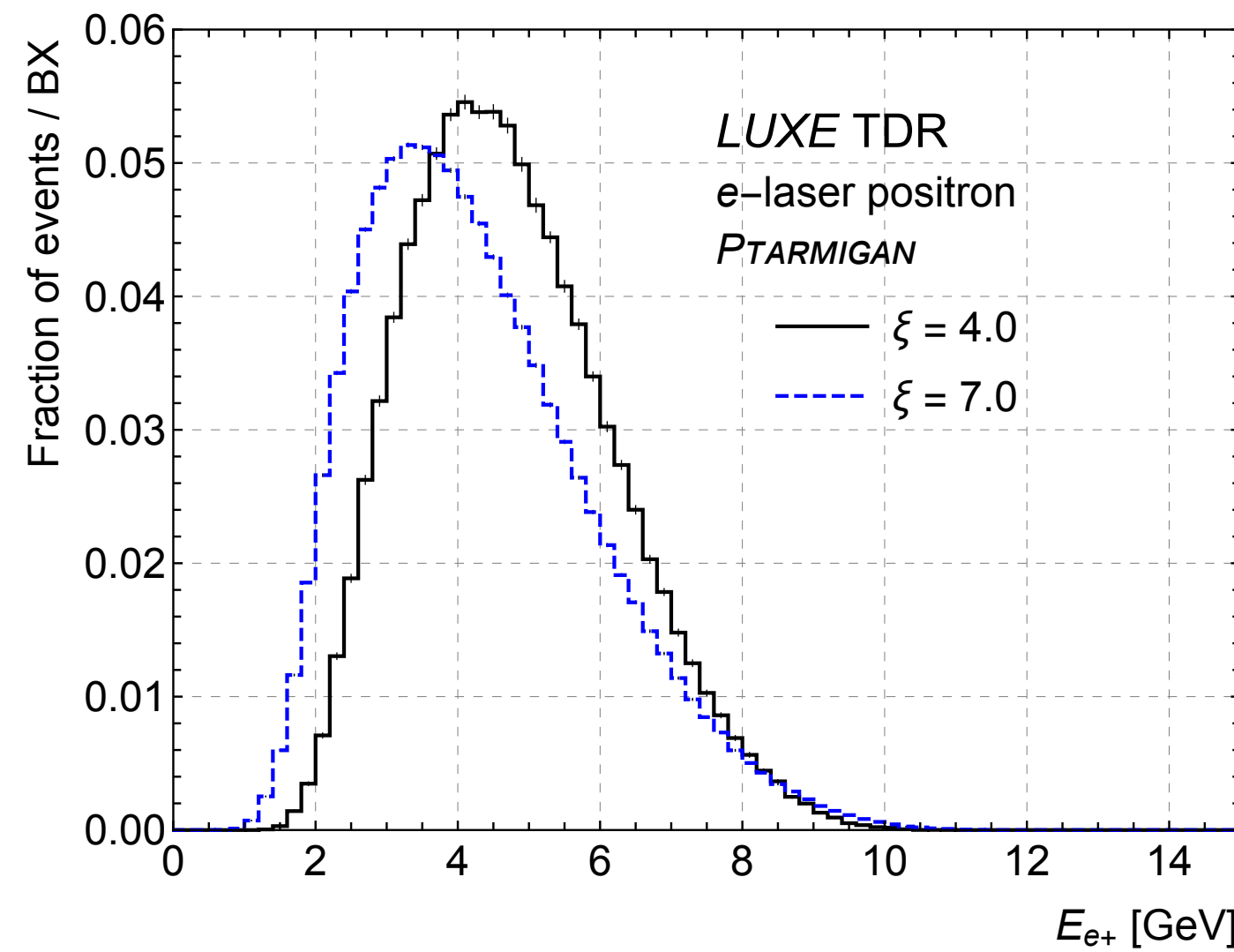
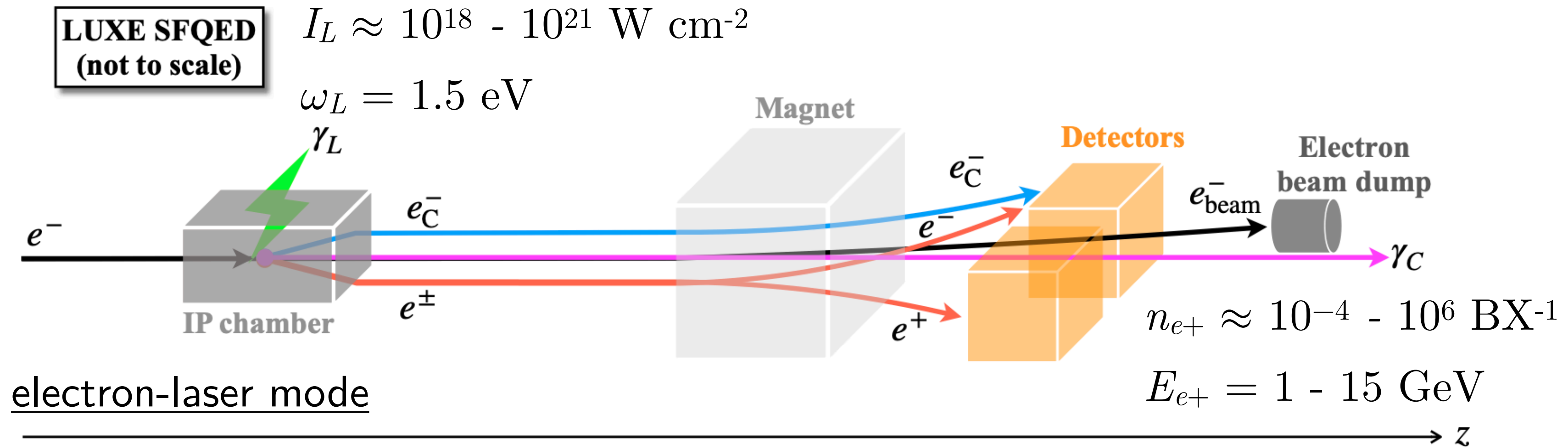
- AGH: Dawid, Jakub, Marek
- DESY: Wolfgang
- ISS: Alina, Mihai, Veta
- TAU: David, Dor, Halina, Michal, Nir, Shan
- UW: Filip, Grzegorz, Kamil, Piotr



# ECAL-P Overview

$$BX = 1.5 \times 10^9$$

$$E_e = 16.5 \text{ GeV}$$



# Outline

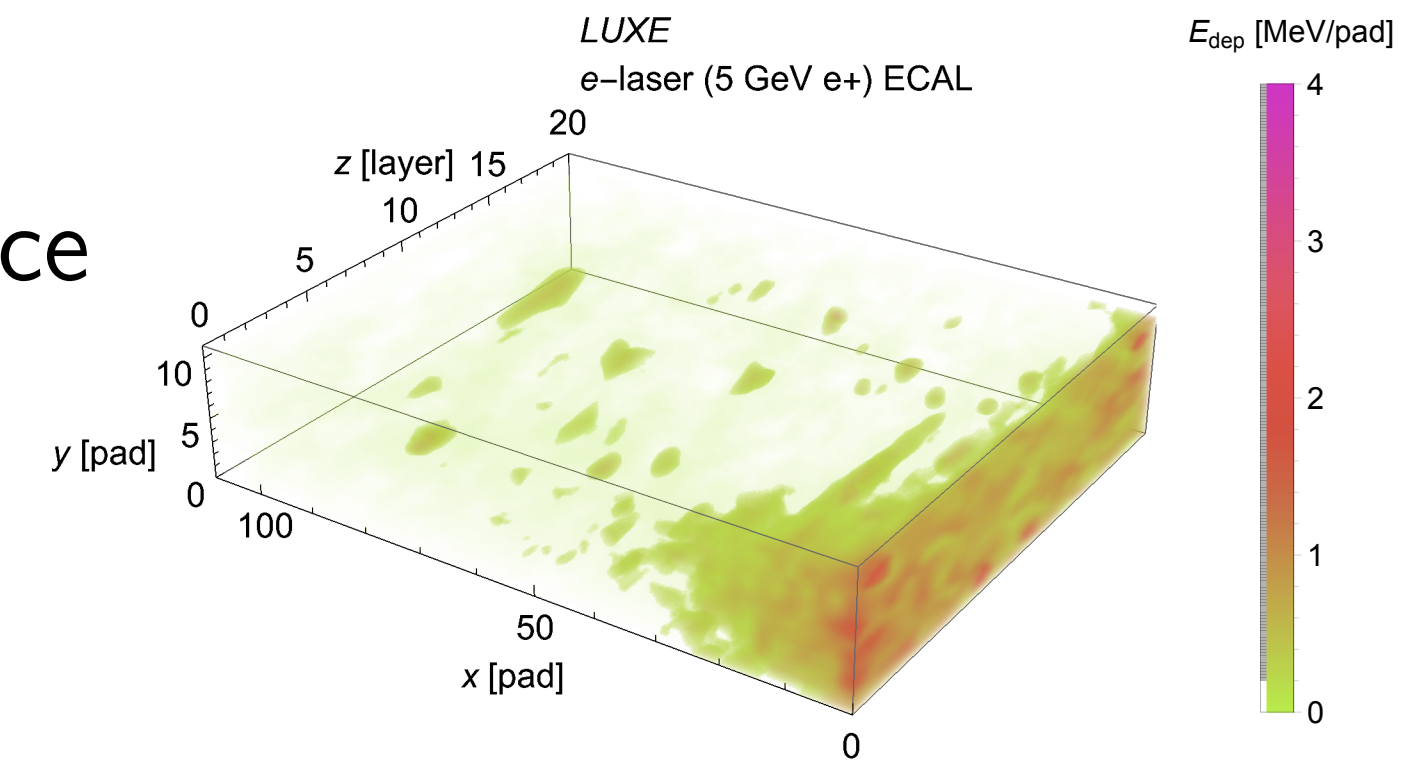
- Background
  - Calibration and Optimization
  - Position reconstruction
  - Spectrum reconstruction
- 
- One more thing ...



# ECAL-P Background

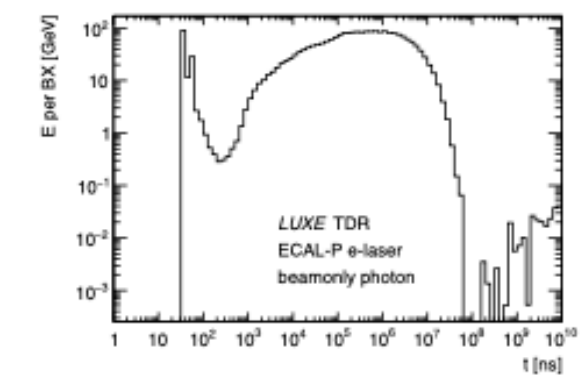
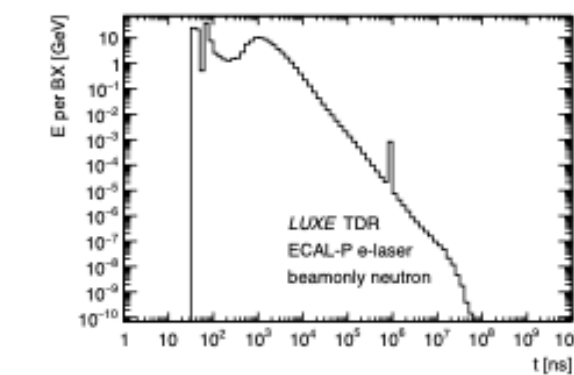
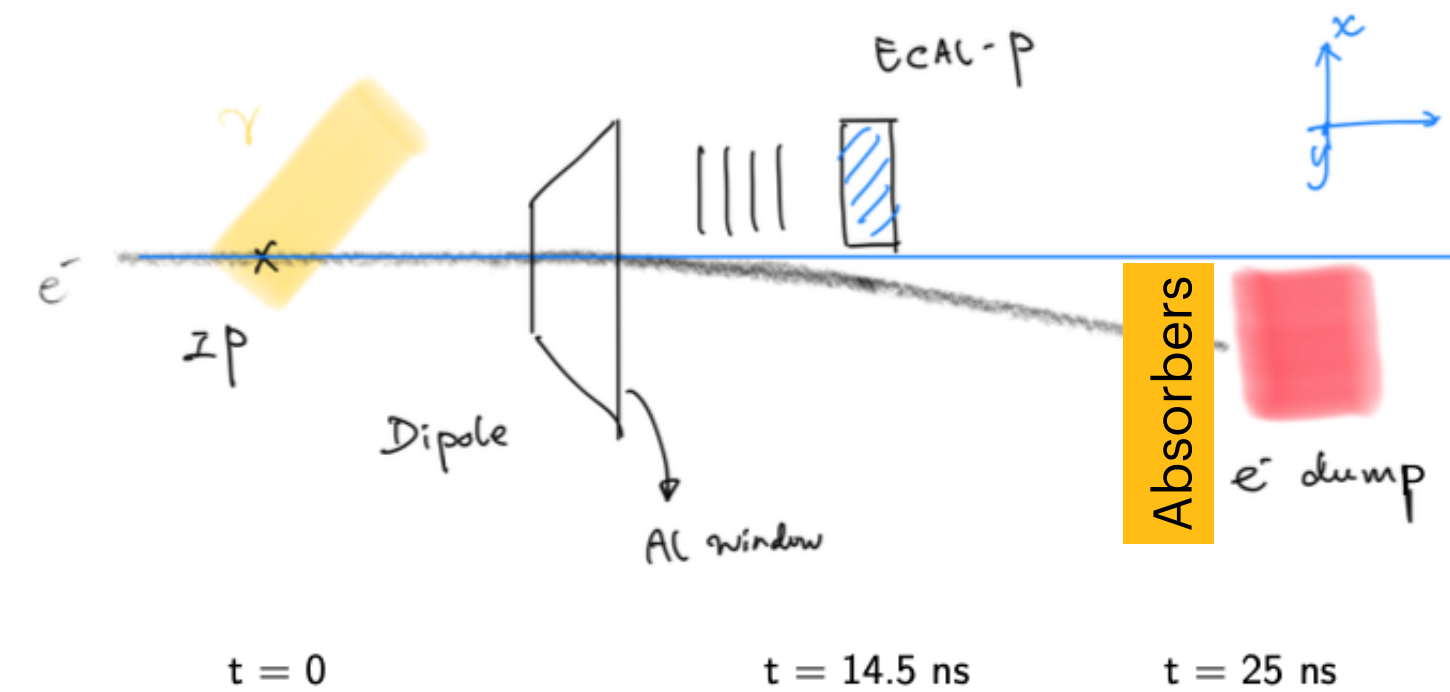
- Electromagnetic background of ECAL-P had been well discussed in the CDR

- electron-laser: beamline  $\Rightarrow$  extra shielding
  - gamma-laser: IP box  $\Rightarrow$  enlarged IP box entrance

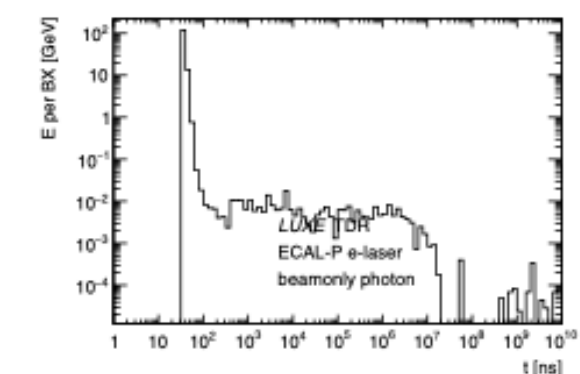
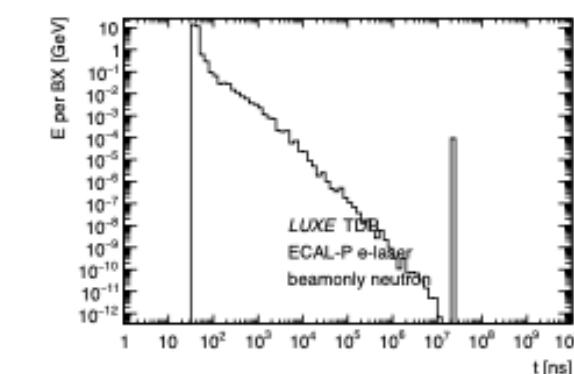


- Hadronic processes were off and particles were killed at dump in CDR simulations
- Severe hadronic background was found in the full simulation

- Electron dump
  - Al window
  - EM background

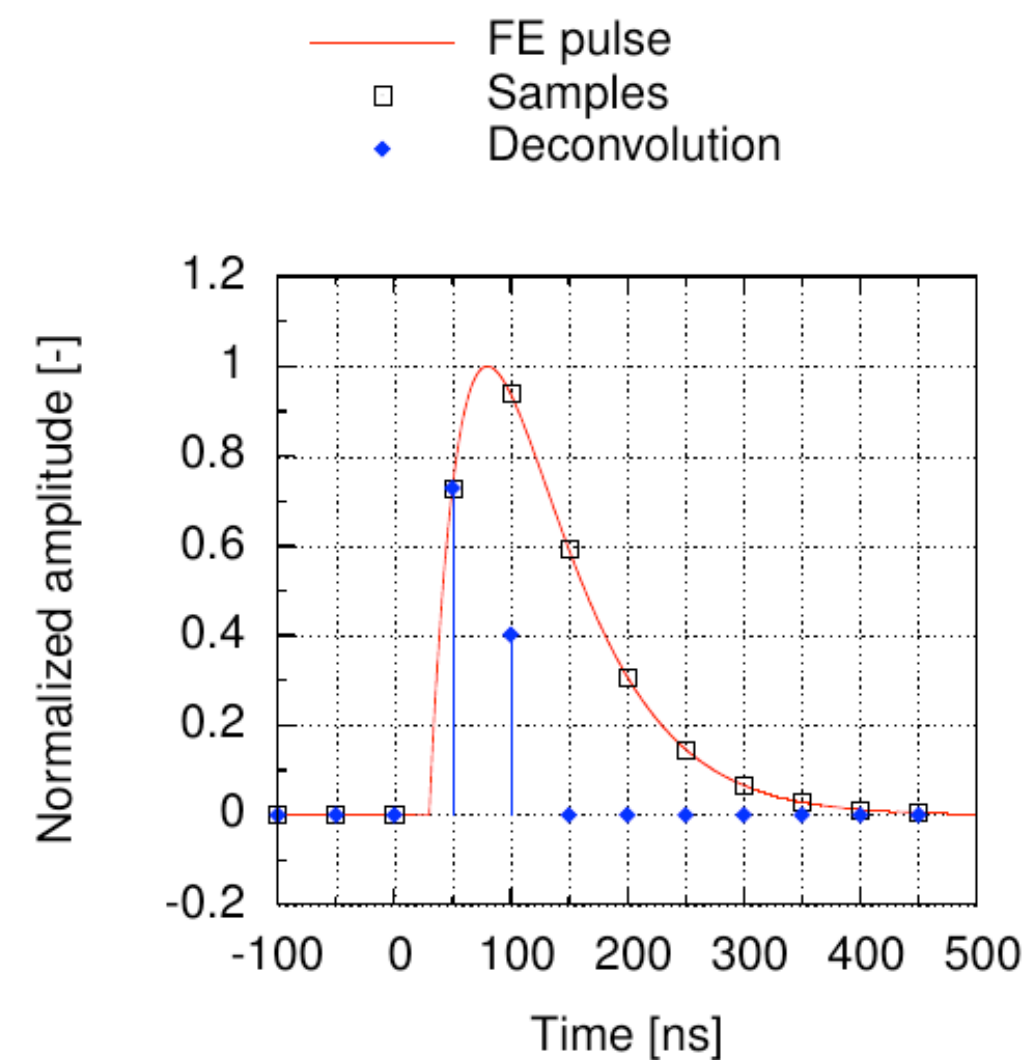


Hadronic BG

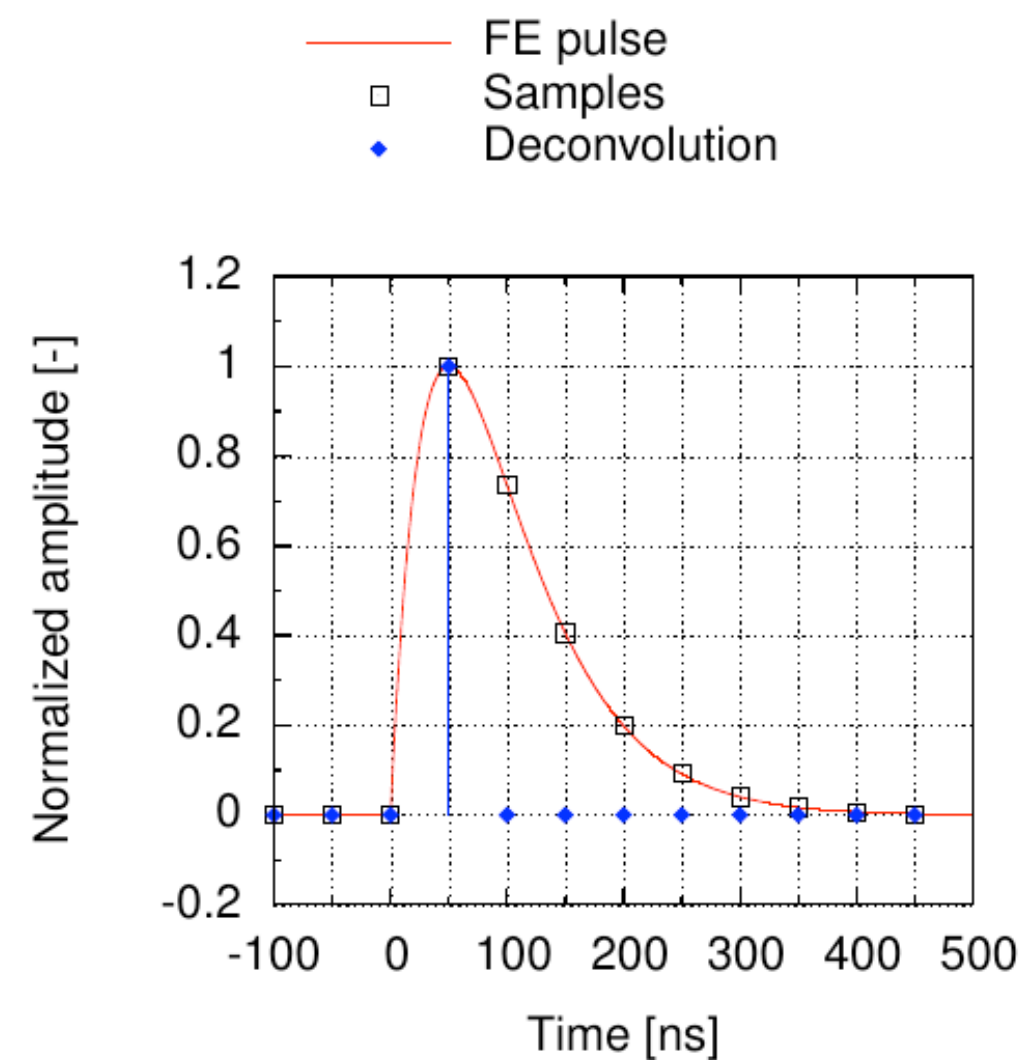


# ECAL-P Background

- The charge deposit is proportional to the peak amplitude of filter output
- The readout system needs at least two timeframes (100 ns) to obtain the peak value

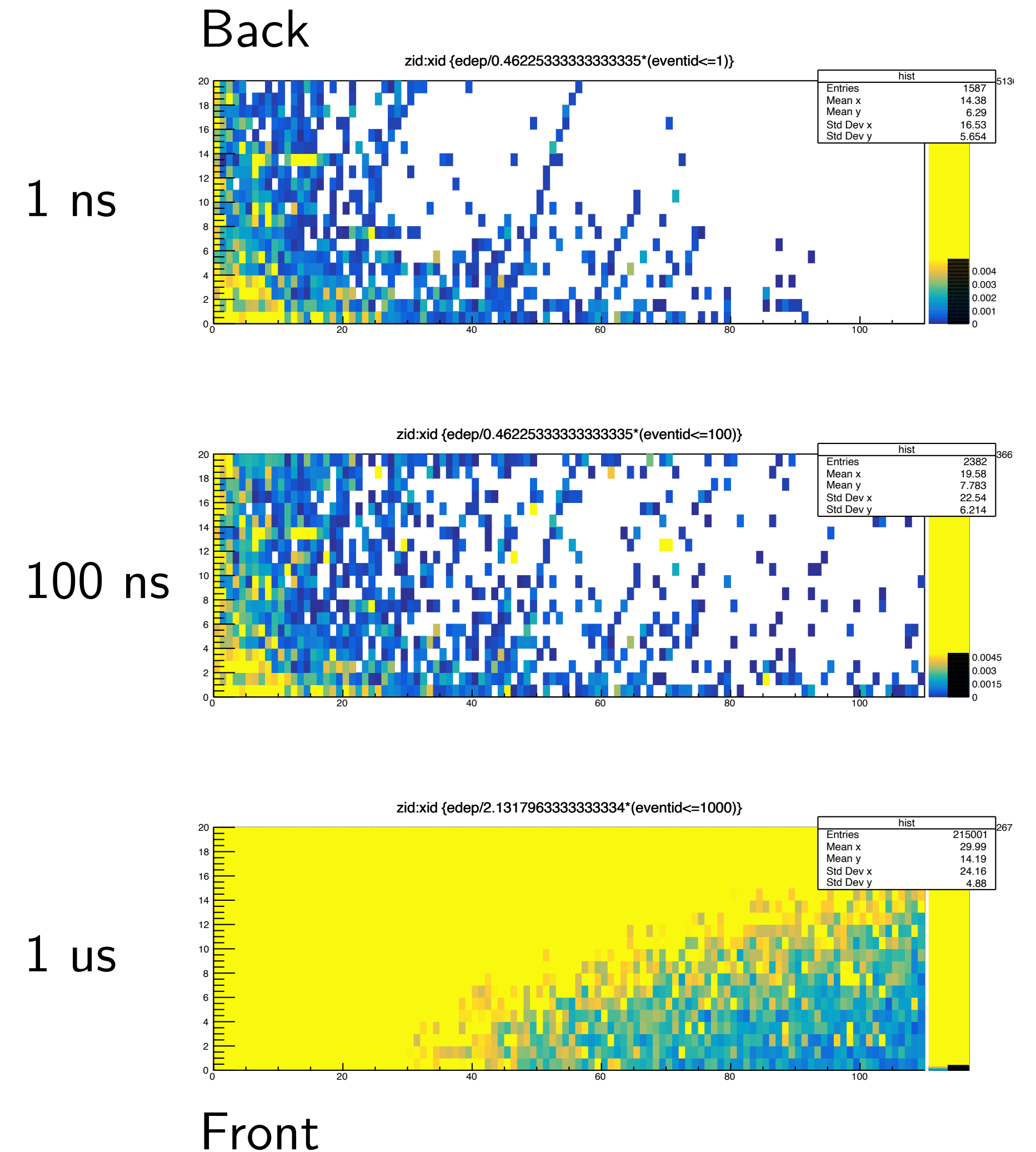


a) Example of asynchronous sampling with two non-zero filter output samples at  $t_0 = 30$  ns



b) Example of synchronous sampling with only one non-zero filter output sample at  $t_0 = 0$  ns

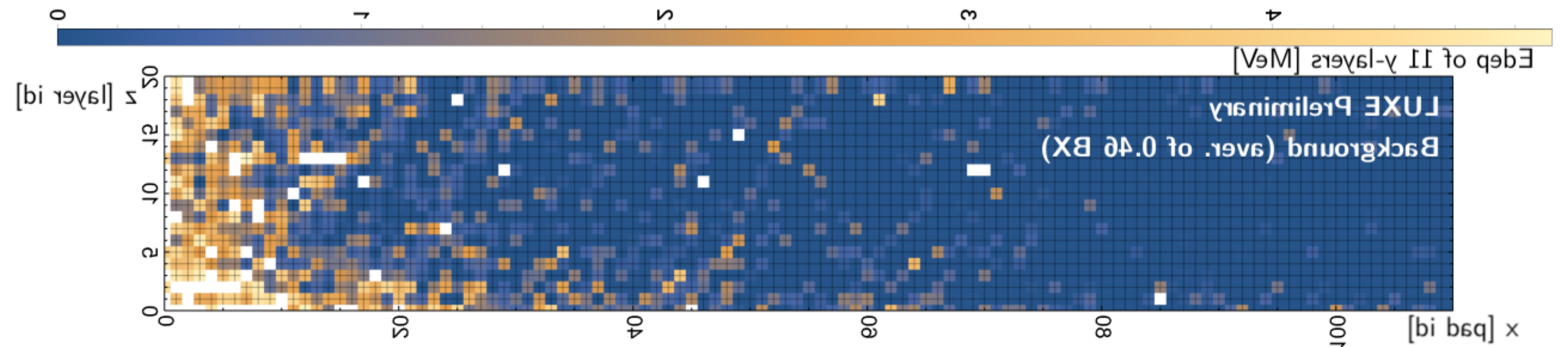
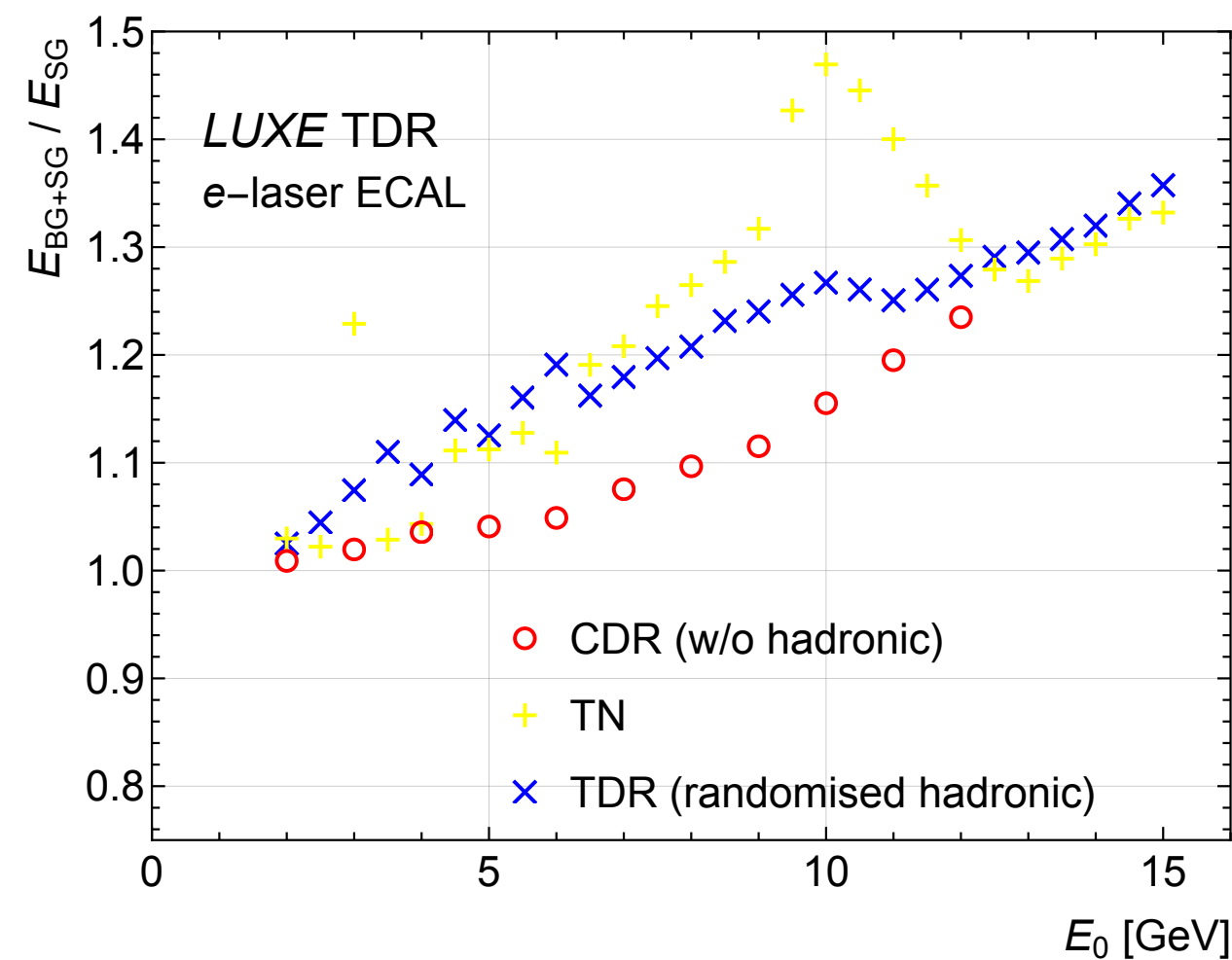
**Figure 2.29:** Examples of deconvolution filter output at  $T_{smp} = \tau_{sh} = 50$  ns.



# ECAL-P Background

- Irregularly high deposits ( $\sim 100$  MIPs) were found
- It was contributed by nuclear reaction, e.g.  $n + \text{Si} \rightarrow p + \text{D} + \text{Mg}$
- It took place tens of times per BX in the central area
- The high deposits were “smeared” in the whole area by randomized redistribution

Background sample = Constant background + redistributed high deposits

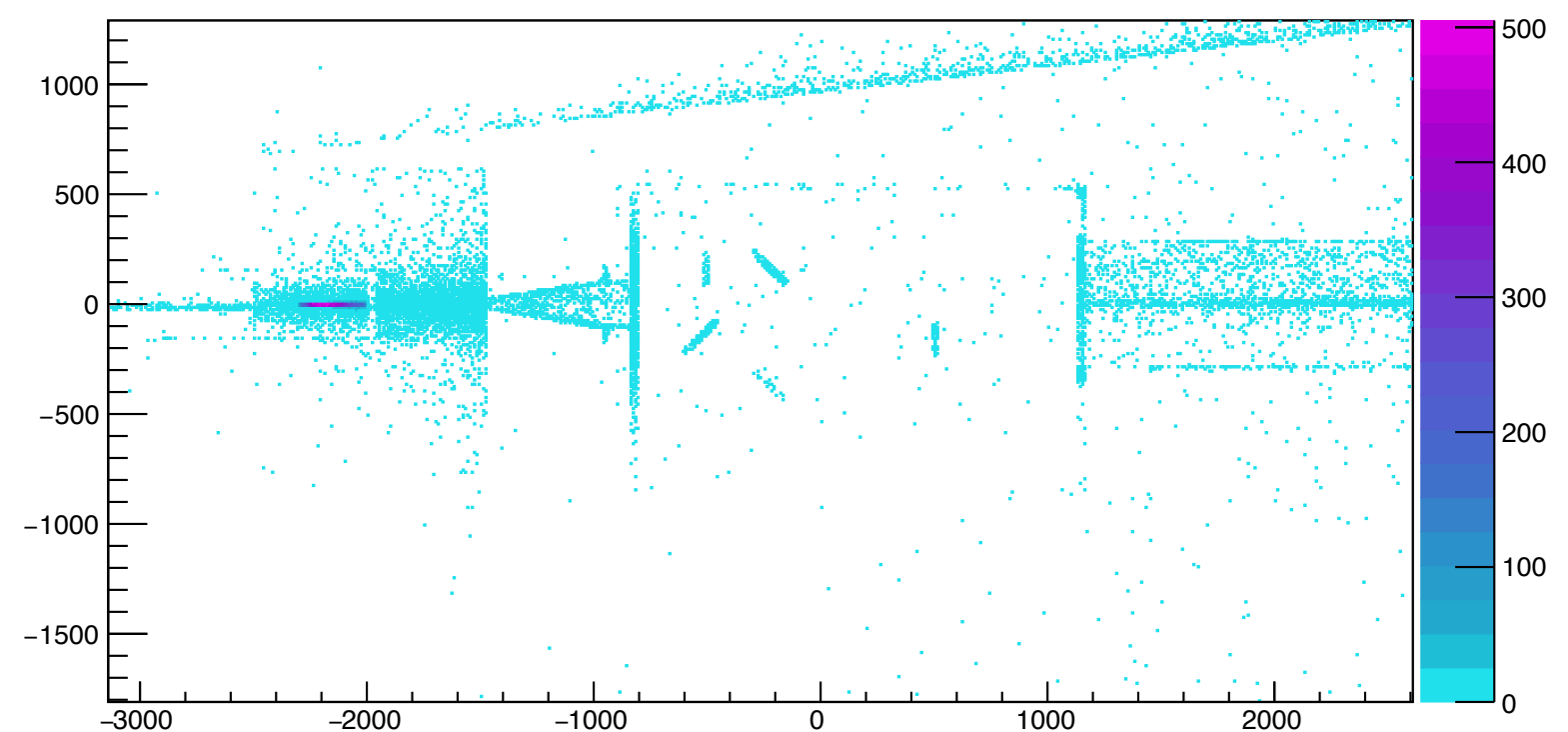
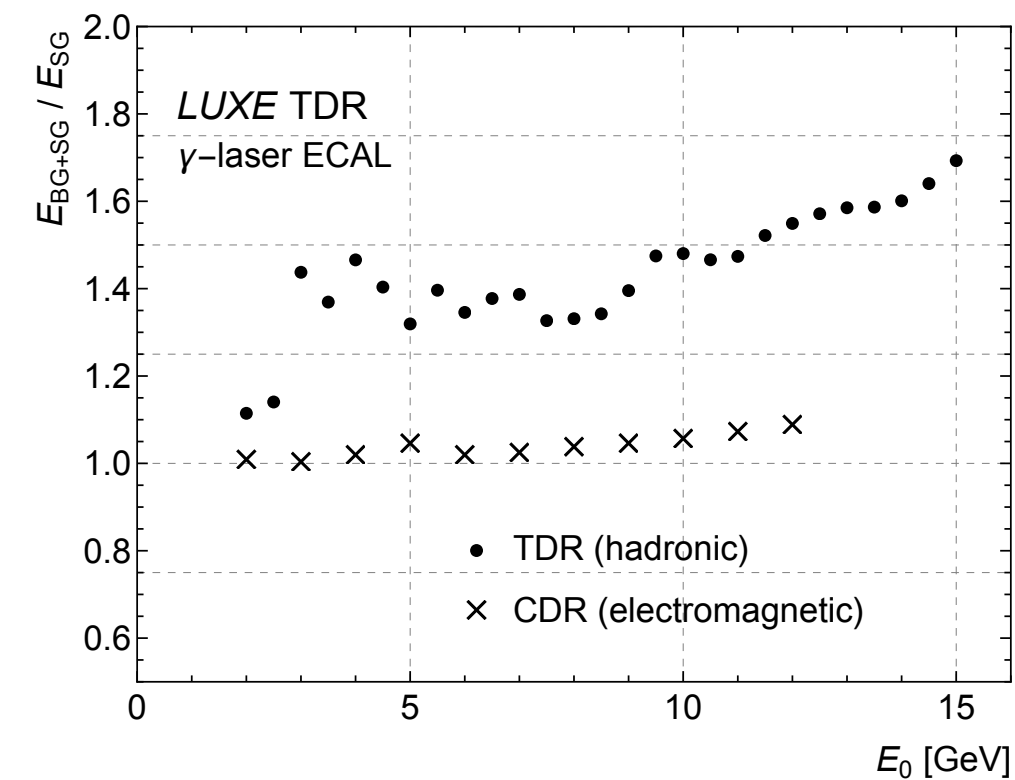
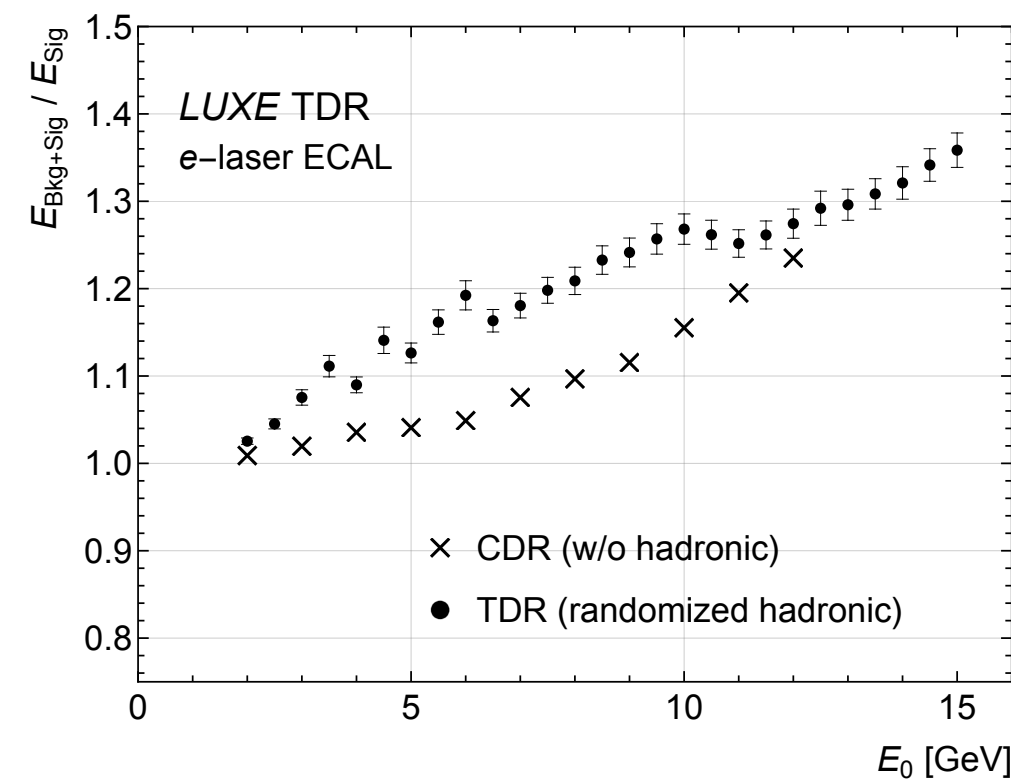
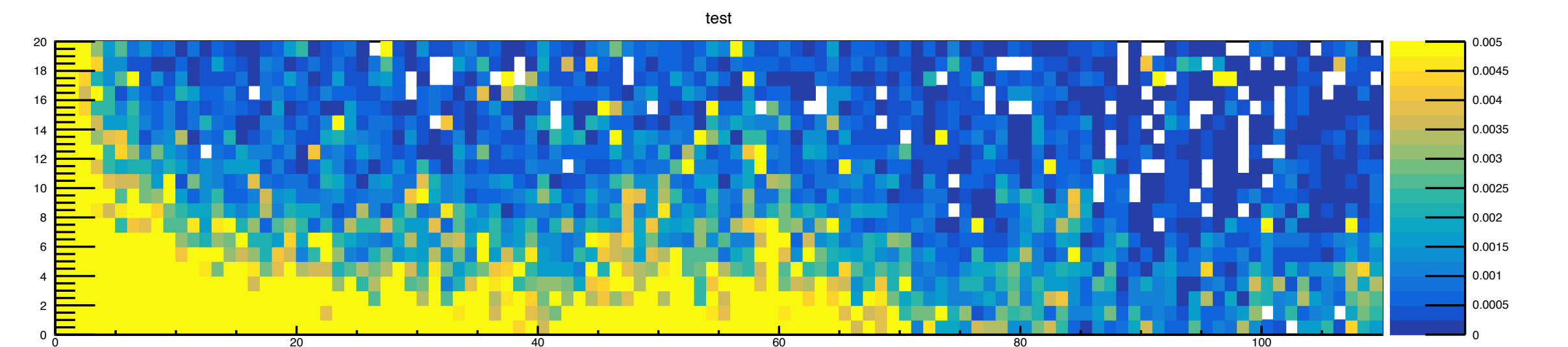
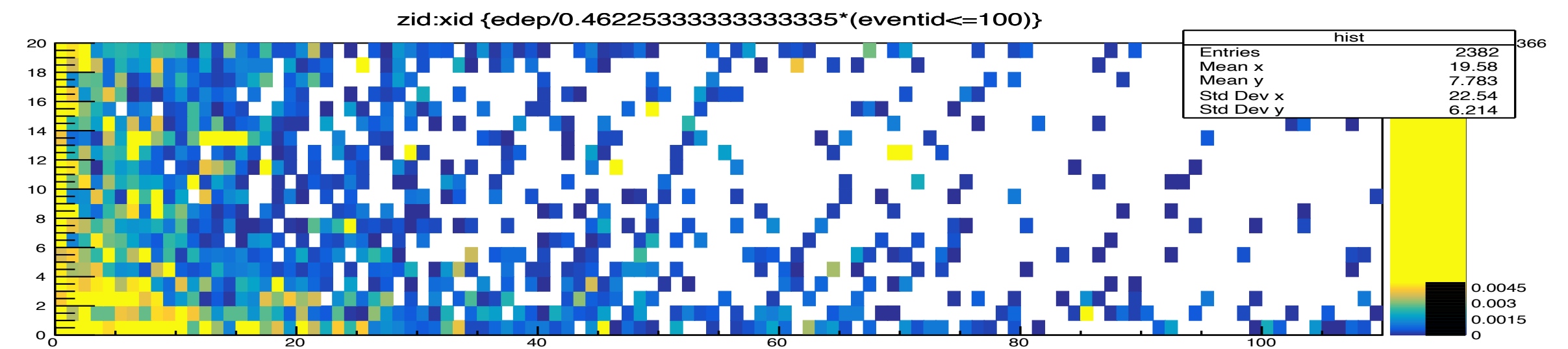




# ECAL-P Background

- The signal-to-noise ratio in gamma-laser setup is worse than in the electron-laser, but still acceptable (below 1.5 on the graph)
- More “consistent” in the central area
- Sourced back to the upstream dump

The accumulated background at 100 ns, topped at 5 MeV (50 MIPs) for e-laser and g-laser setups



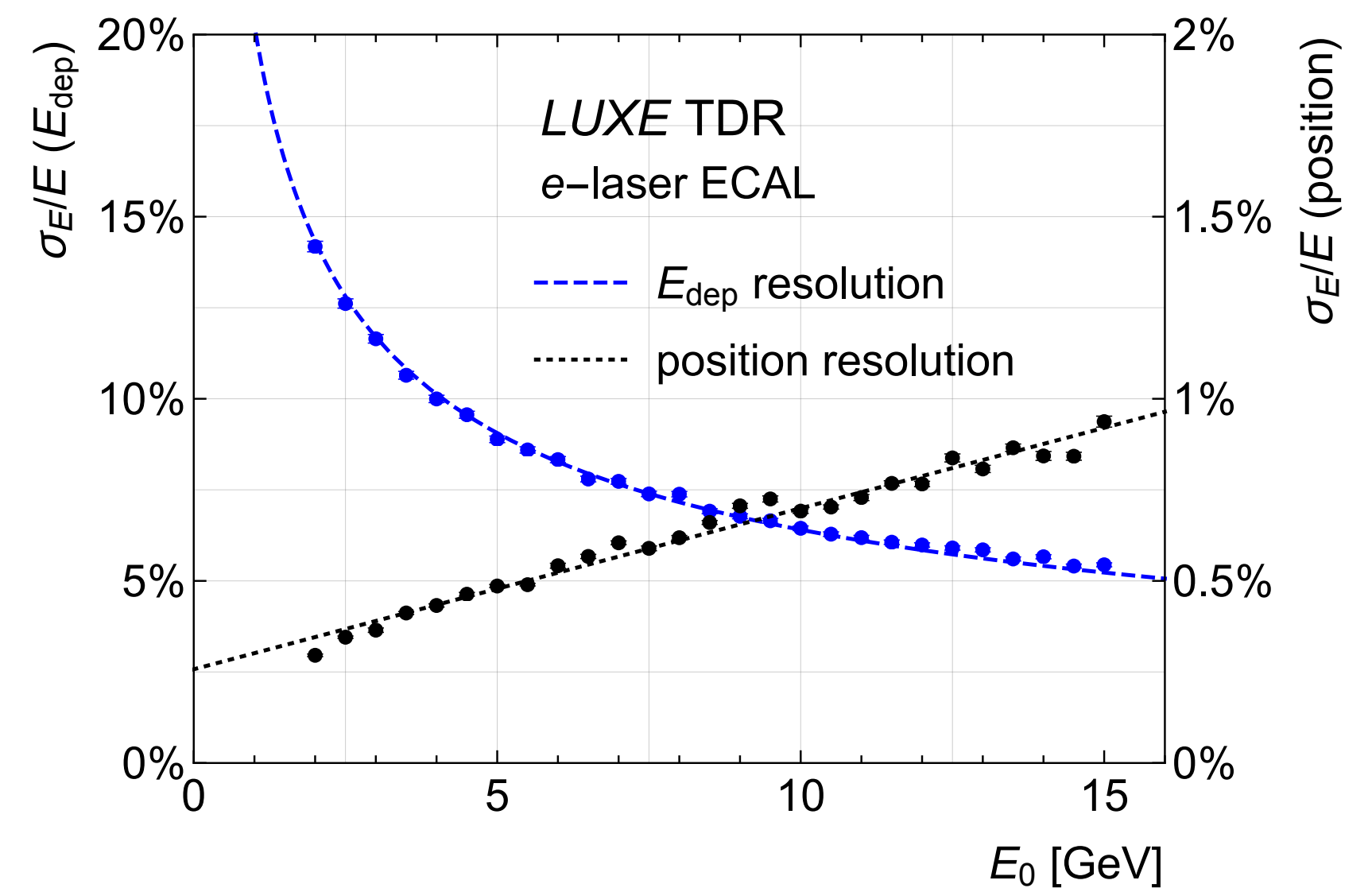
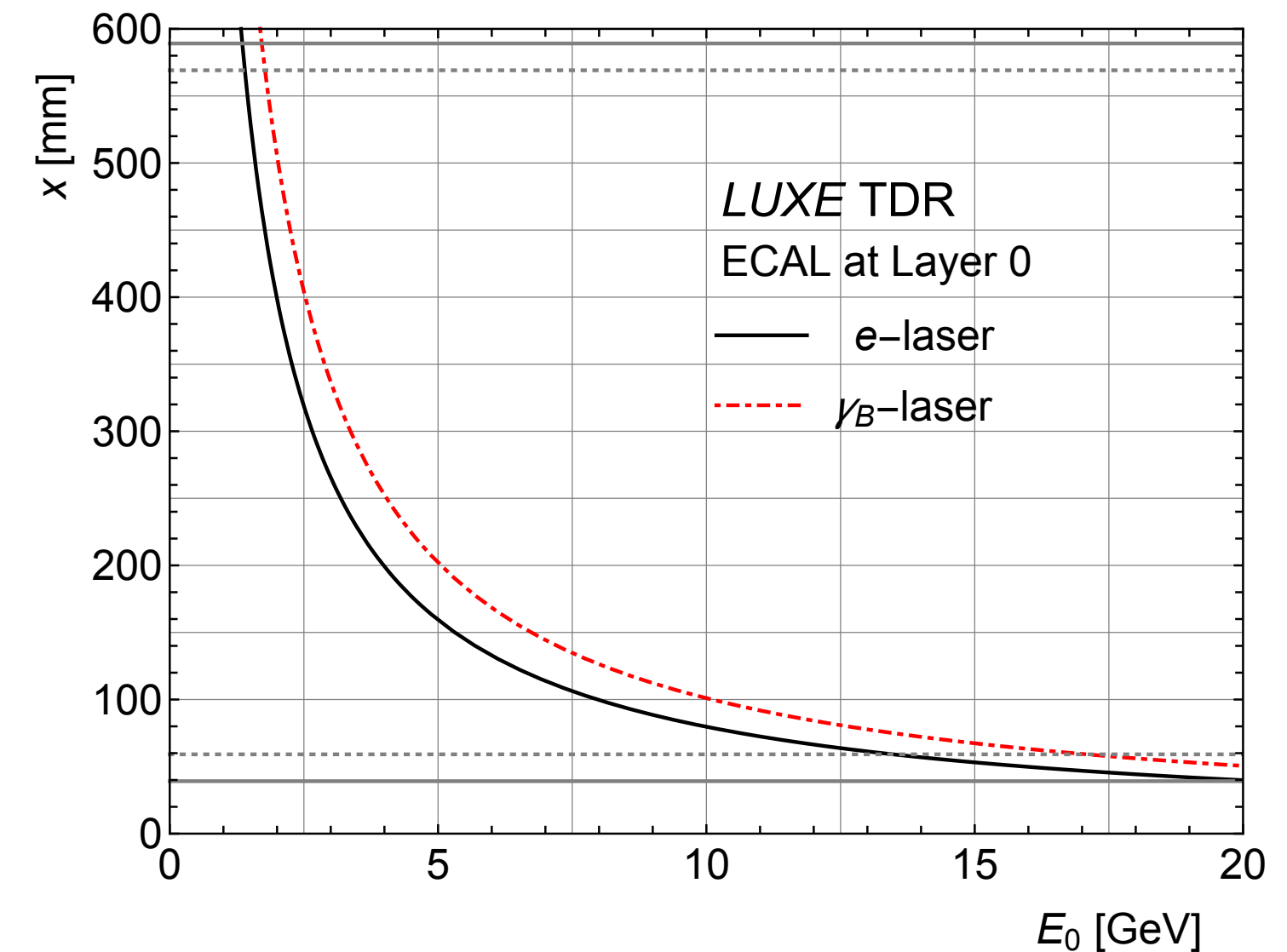


# Outline

- Background
- Calibration and Optimization
- Position reconstruction
- Spectrum reconstruction
- One more thing ...

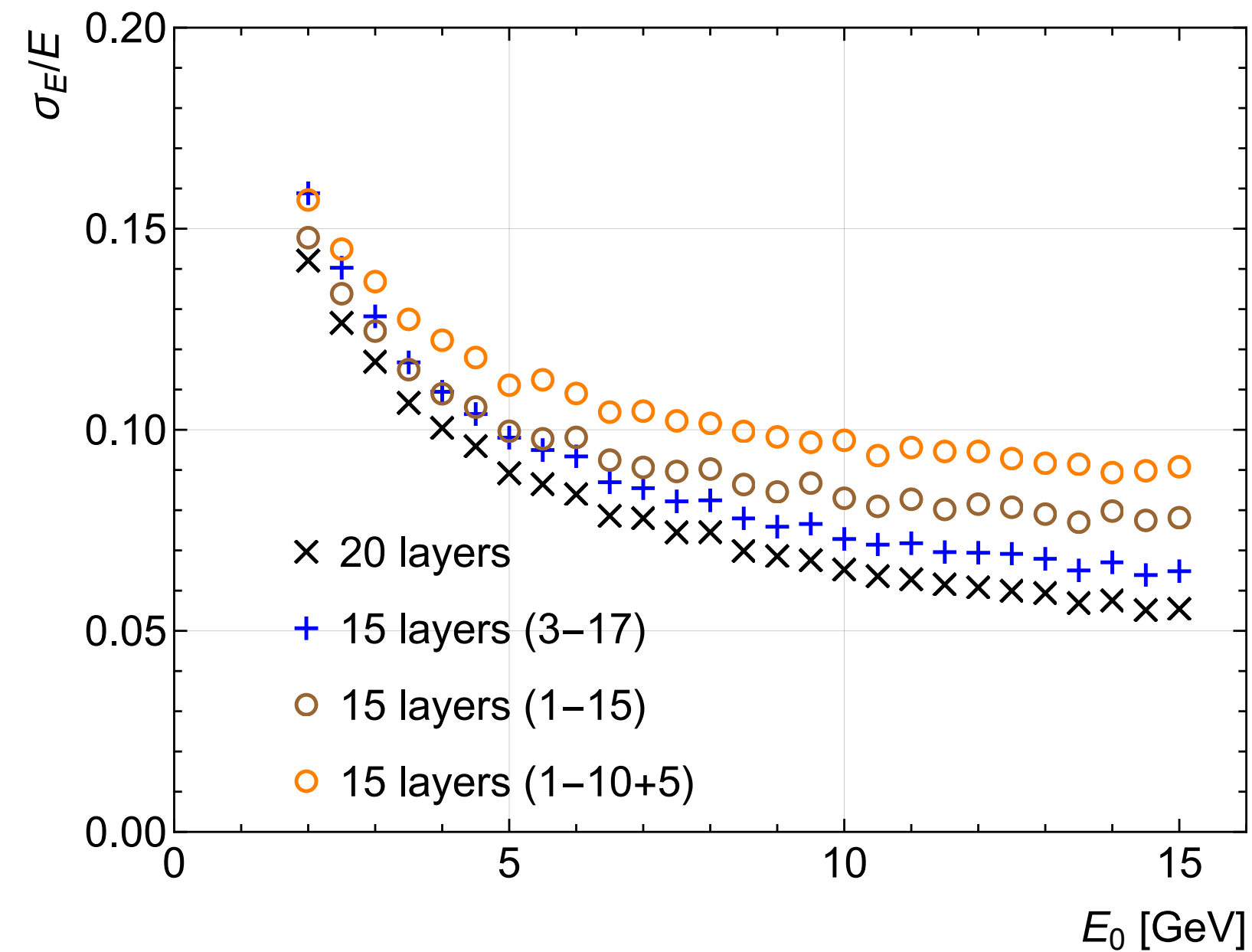
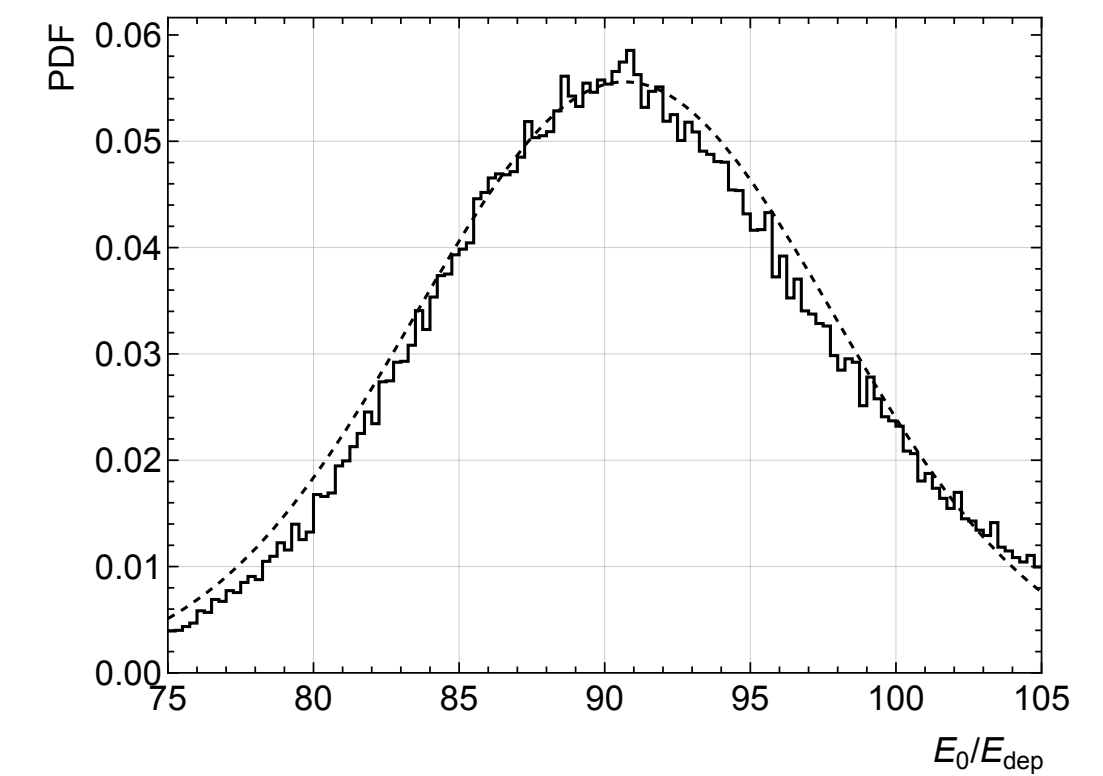
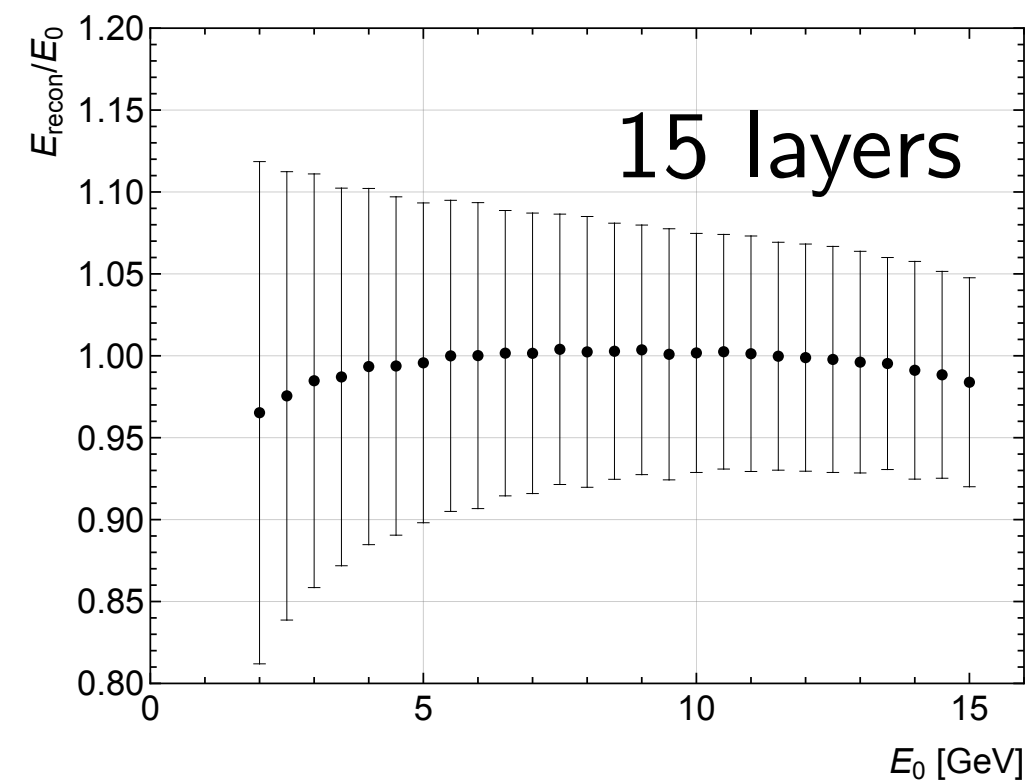
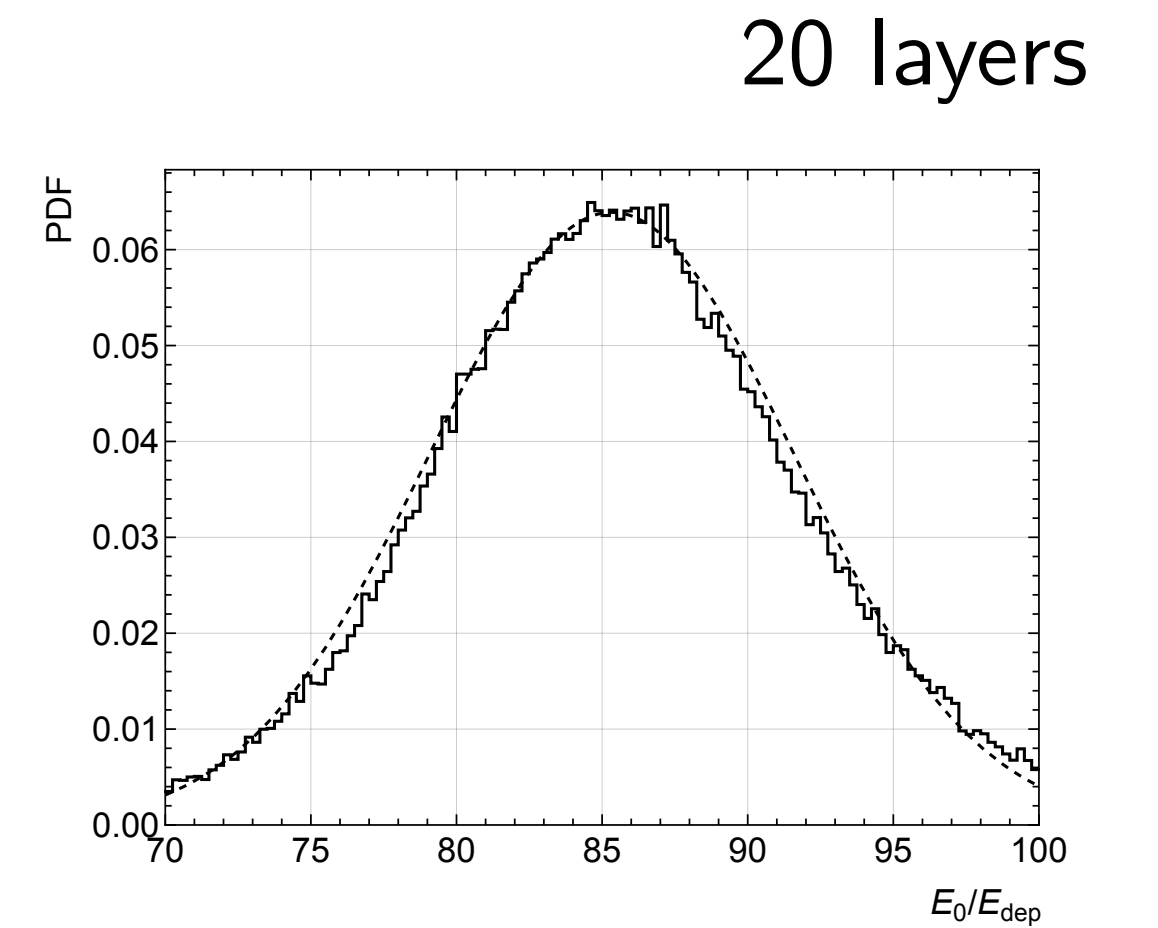
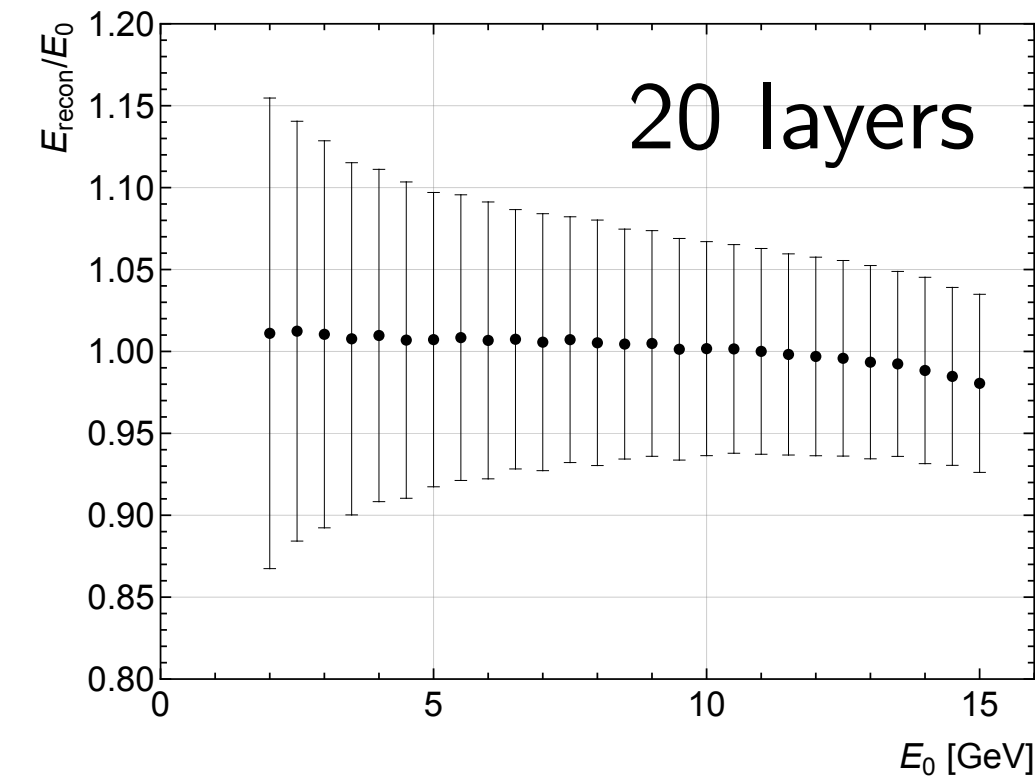
# ECAL-P Resolution

- The ECAL is going to calibrate the position-energy dependence created by the magnetic dipole
- A full 20-layer ECAL has the energy resolution of  $20\%/\sqrt{E/\text{GeV}}$  (in the CDR)
- Due to funding limit, it is likely that the ECAL-P will only have 15 active layers
- Several combinations of sampling schemes are under investigation

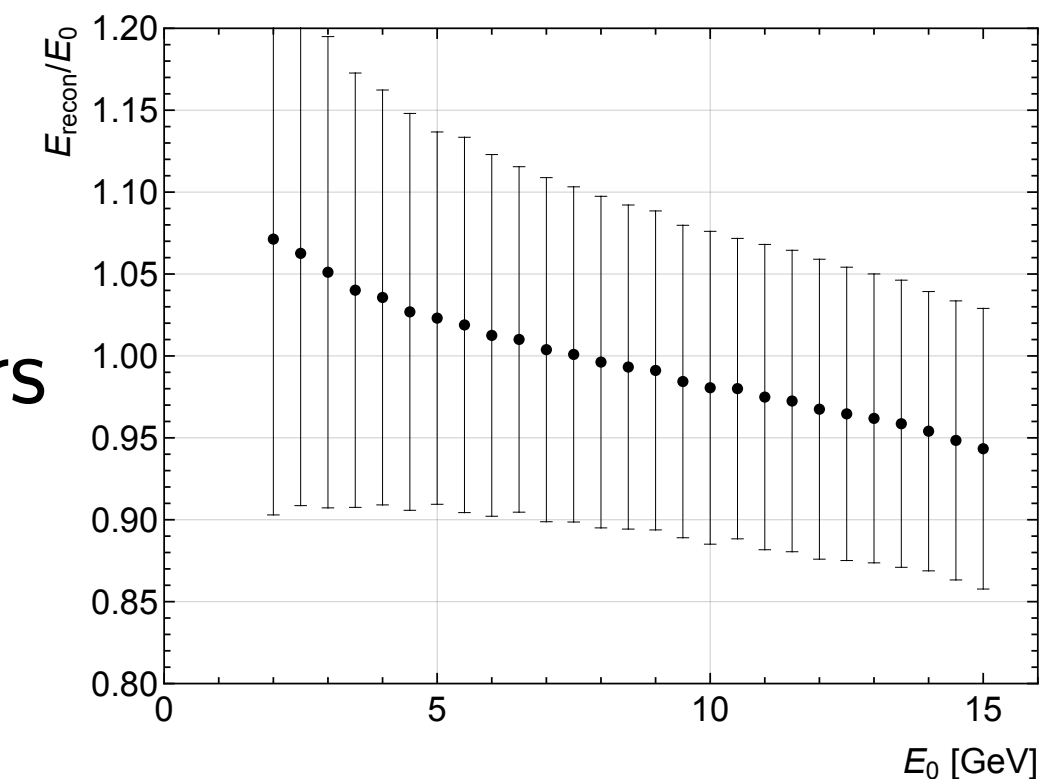


# Optimization

- An enumeration of all possible combinations were run to find the best 15 layers
- The simplest method, calibration by a constant ratio, was used



10+5 layers  
(20 X0)

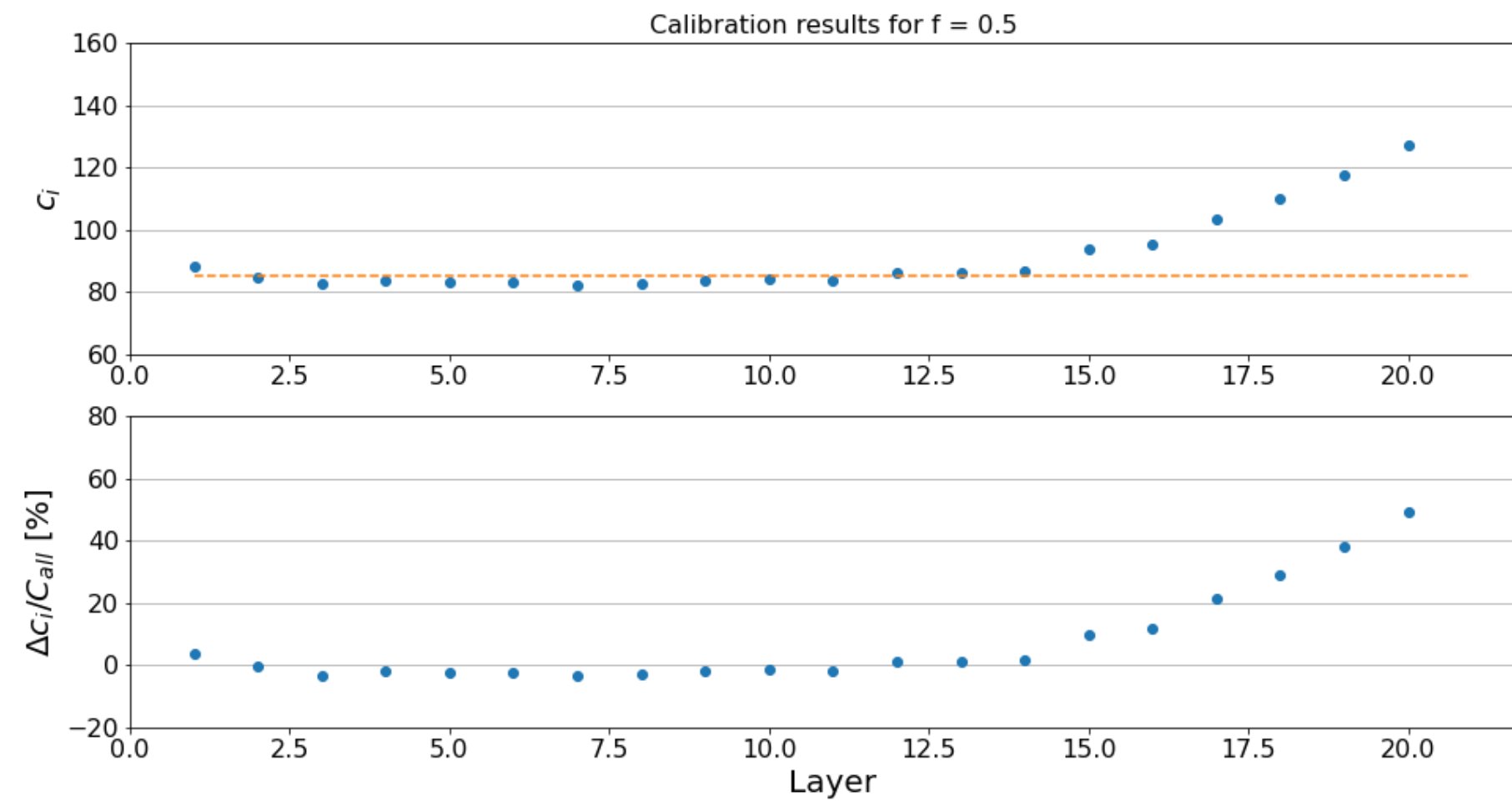
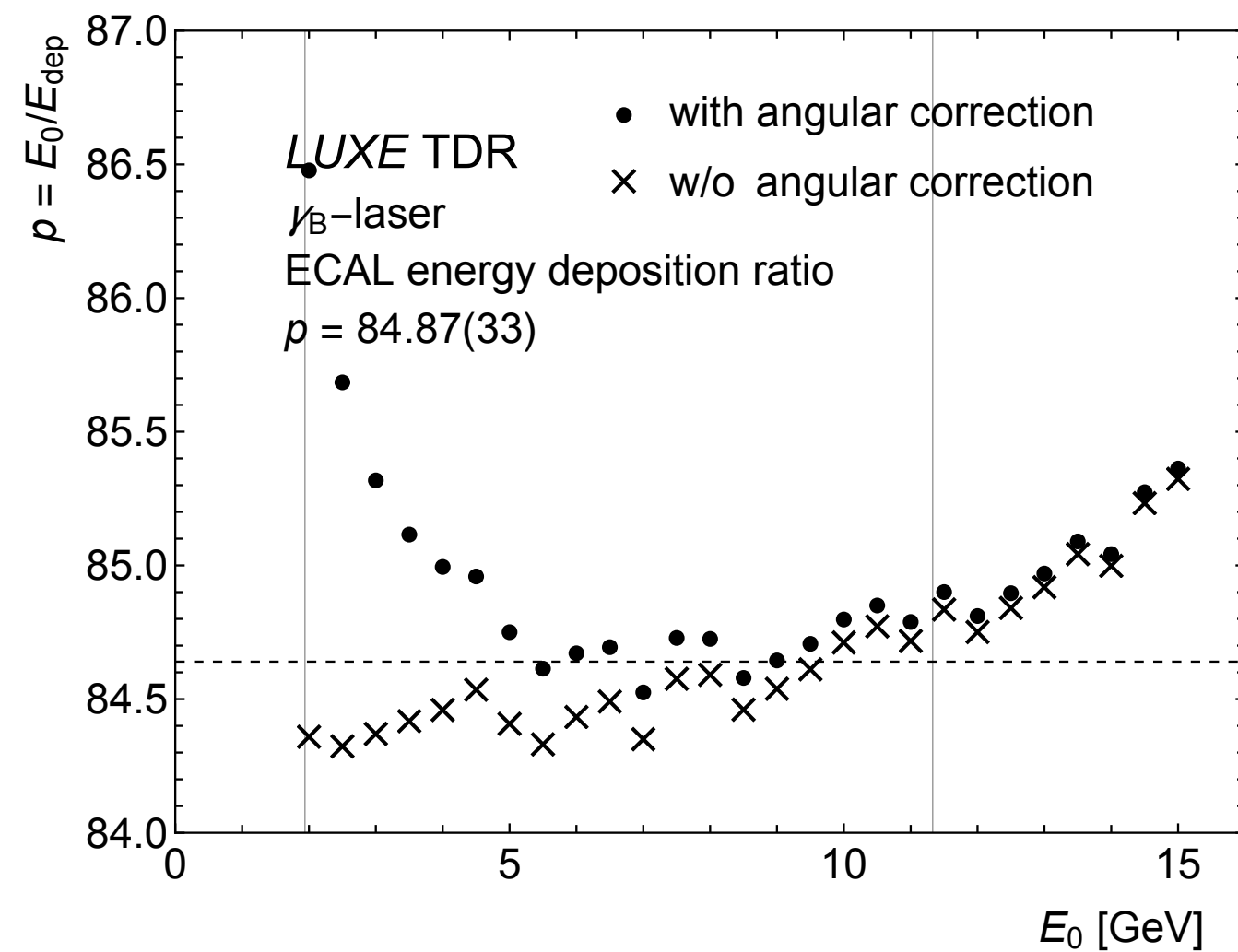
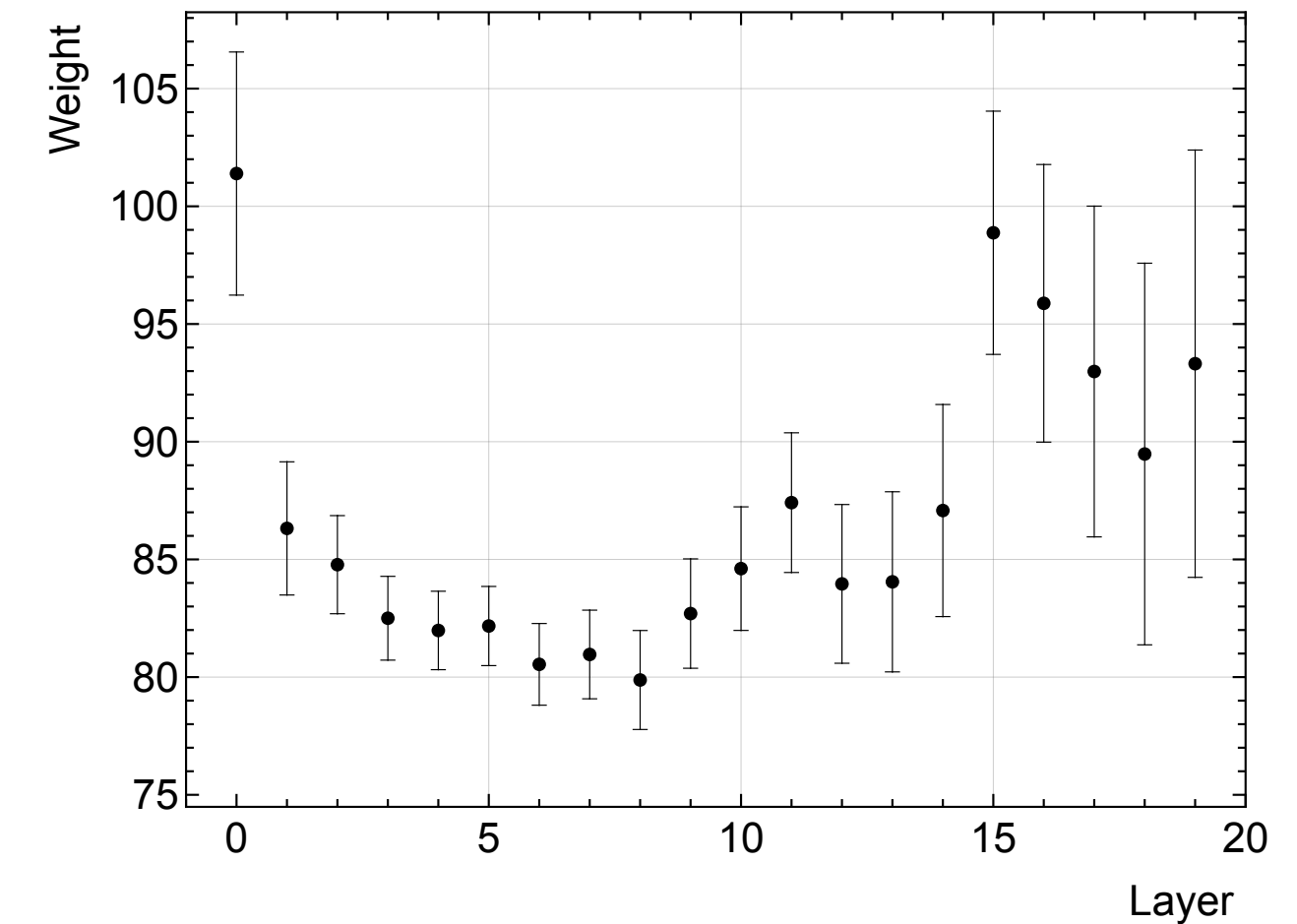


15 layers (15 X0)

Calibration & Optimization

# Calibration

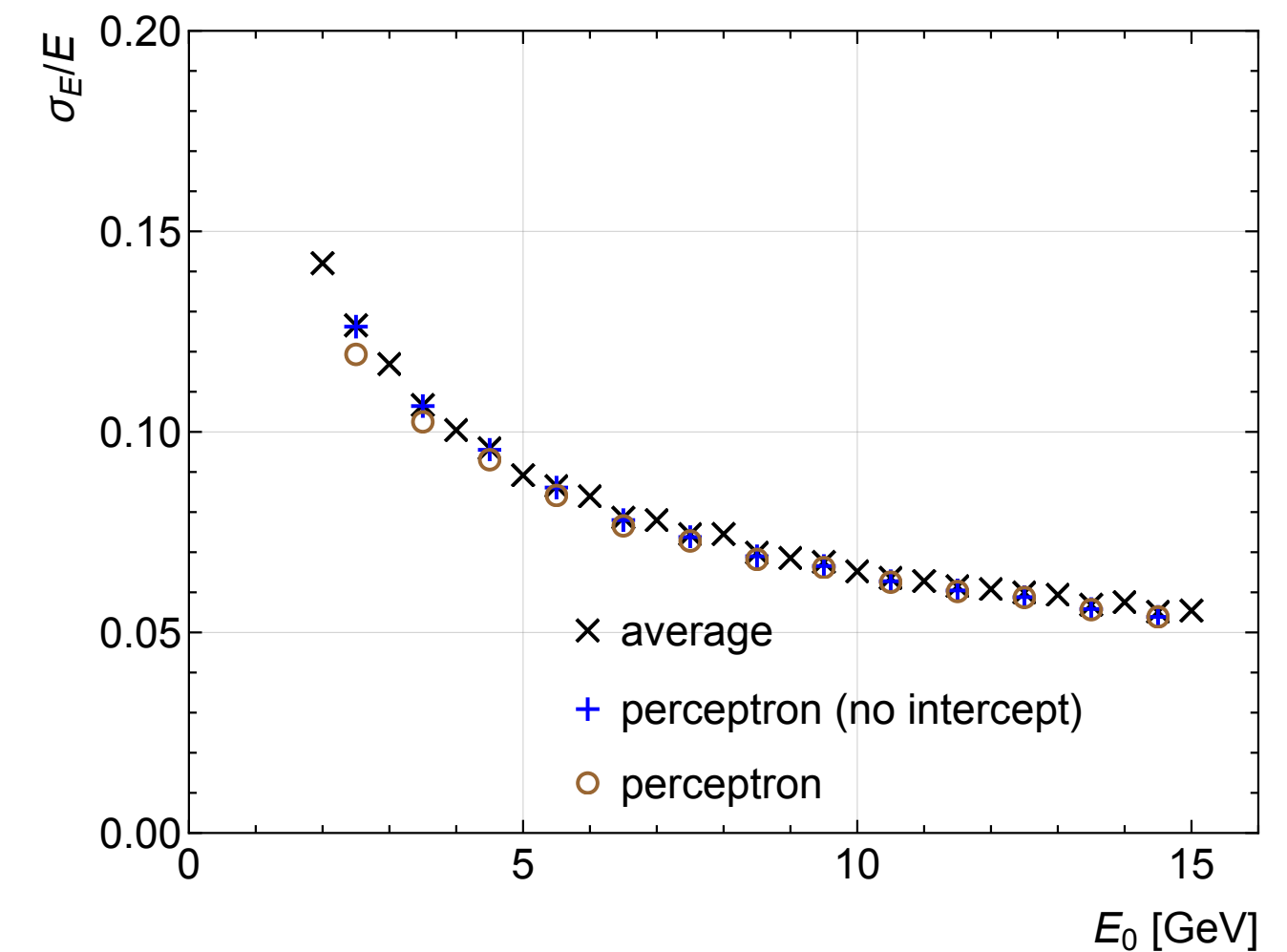
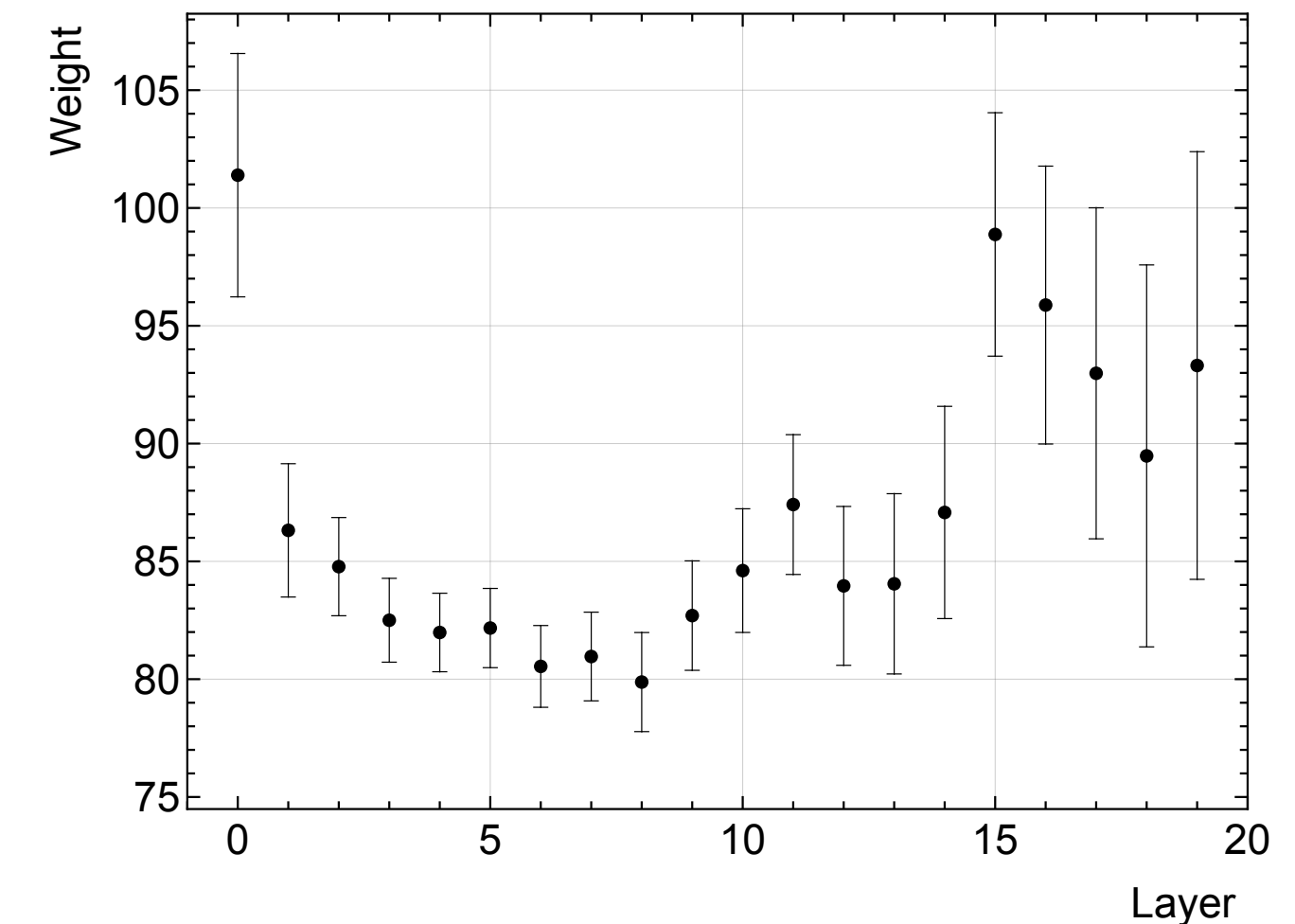
- In the CDR, a constant calibration factor was used for all layers
- It has been found out that different energy has different optimized calibration ratio
- An algorithm of compensation has been developed by giving different layer different calibration ratio





# Calibration

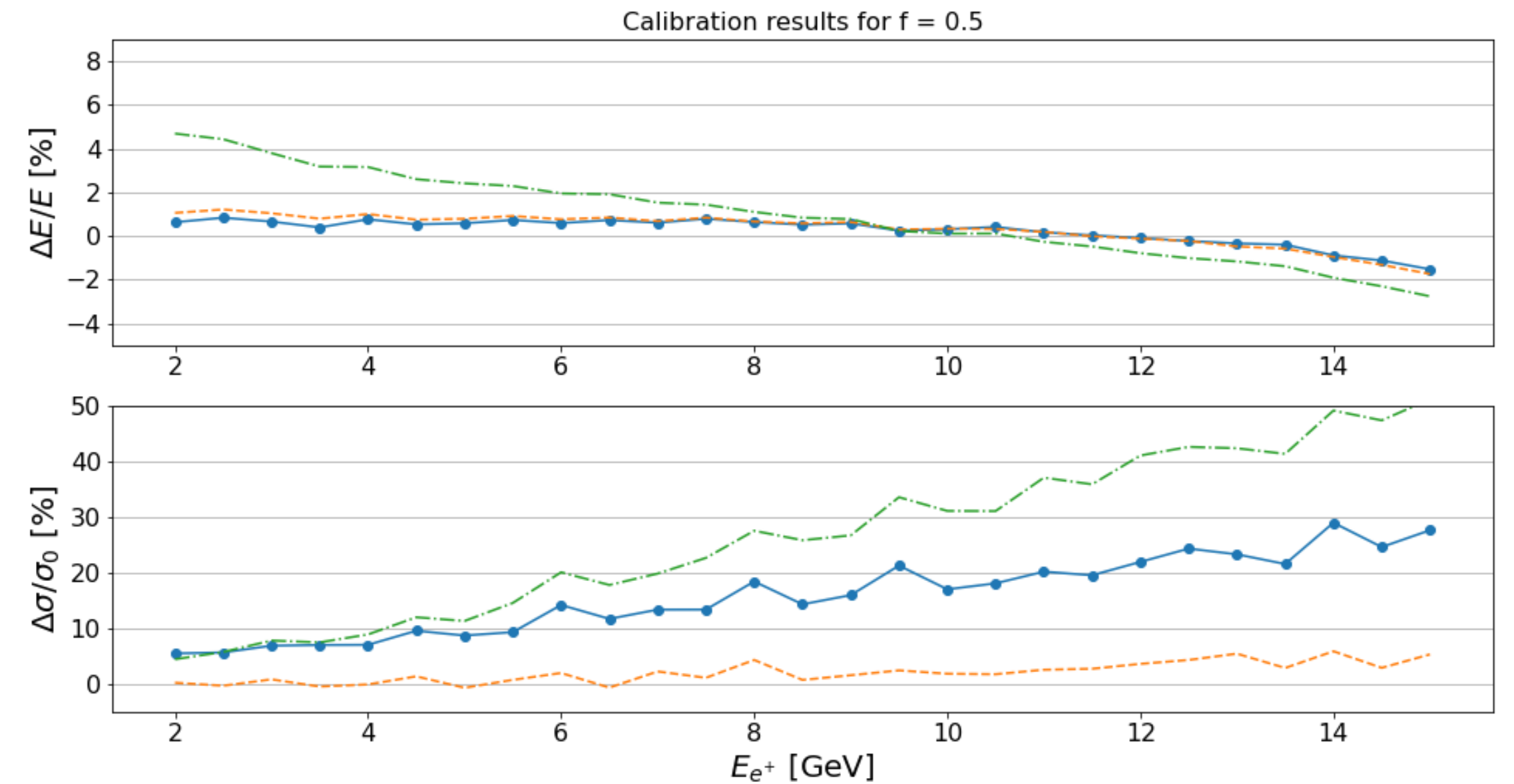
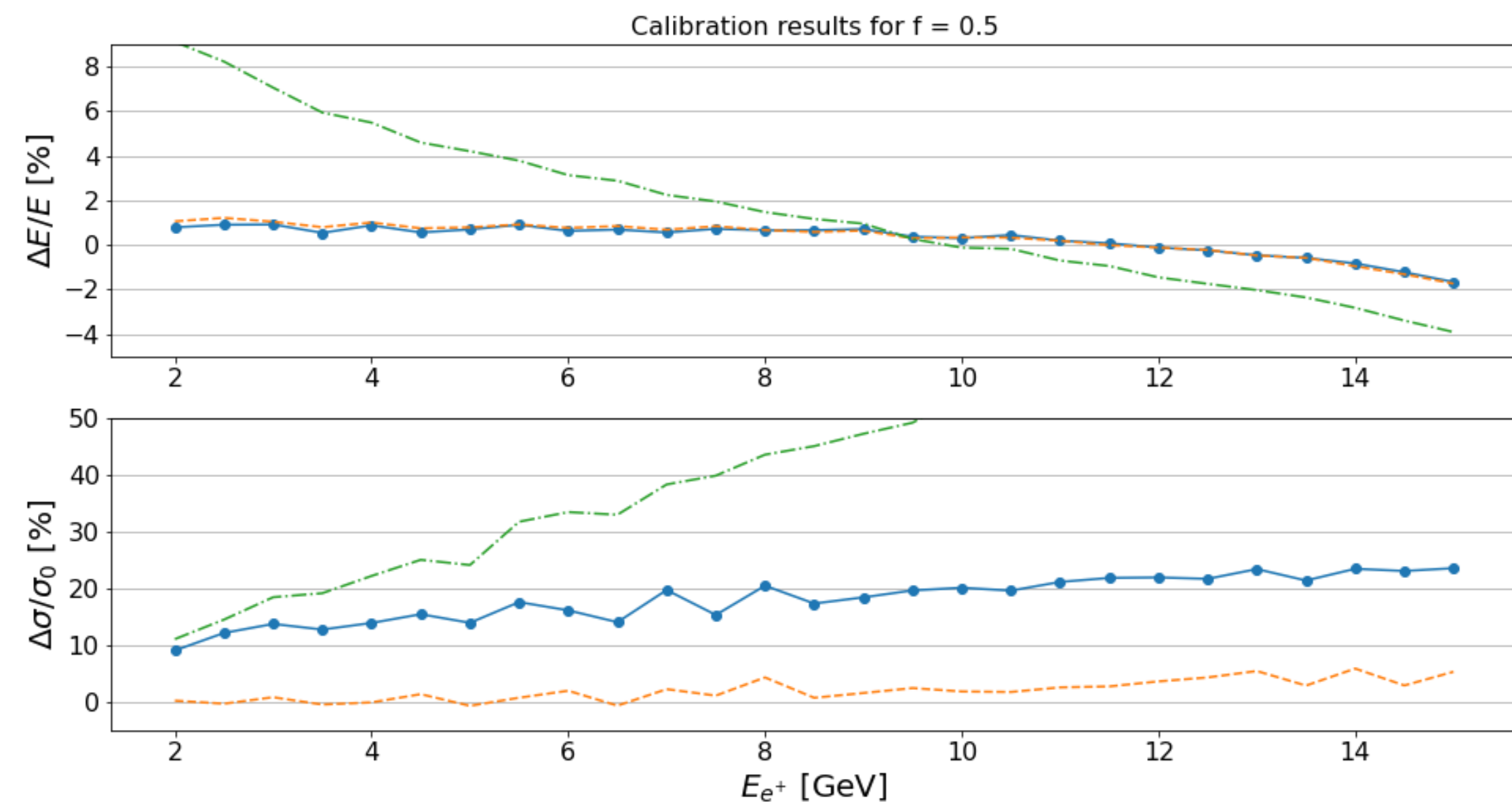
- In the CDR, a constant calibration factor was used for all layers
- It has been found out that different energy has different optimized calibration ratio
- An algorithm of compensation has been developed by giving different layer different calibration ratio
- A perceptron was also trained and confirmed the trending of weight distribution



# Calibration

- The framework developed by Filip gave a faster result with more flexibility

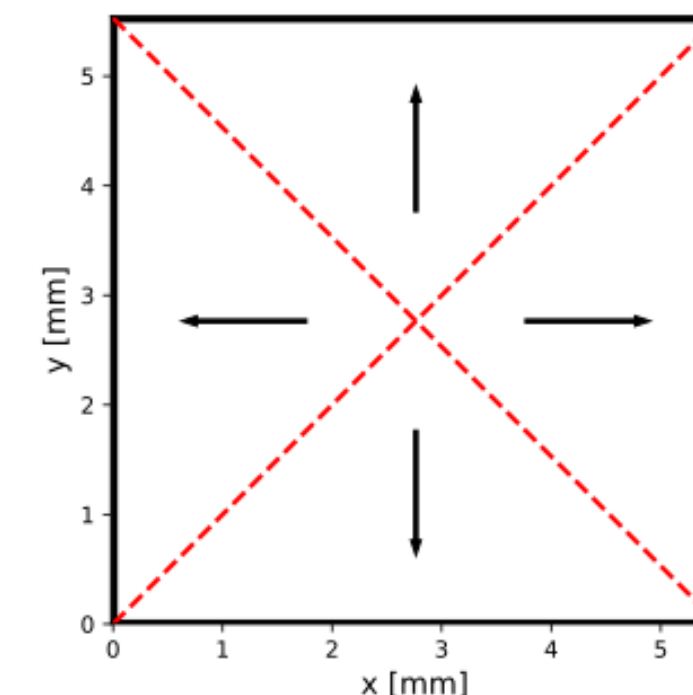
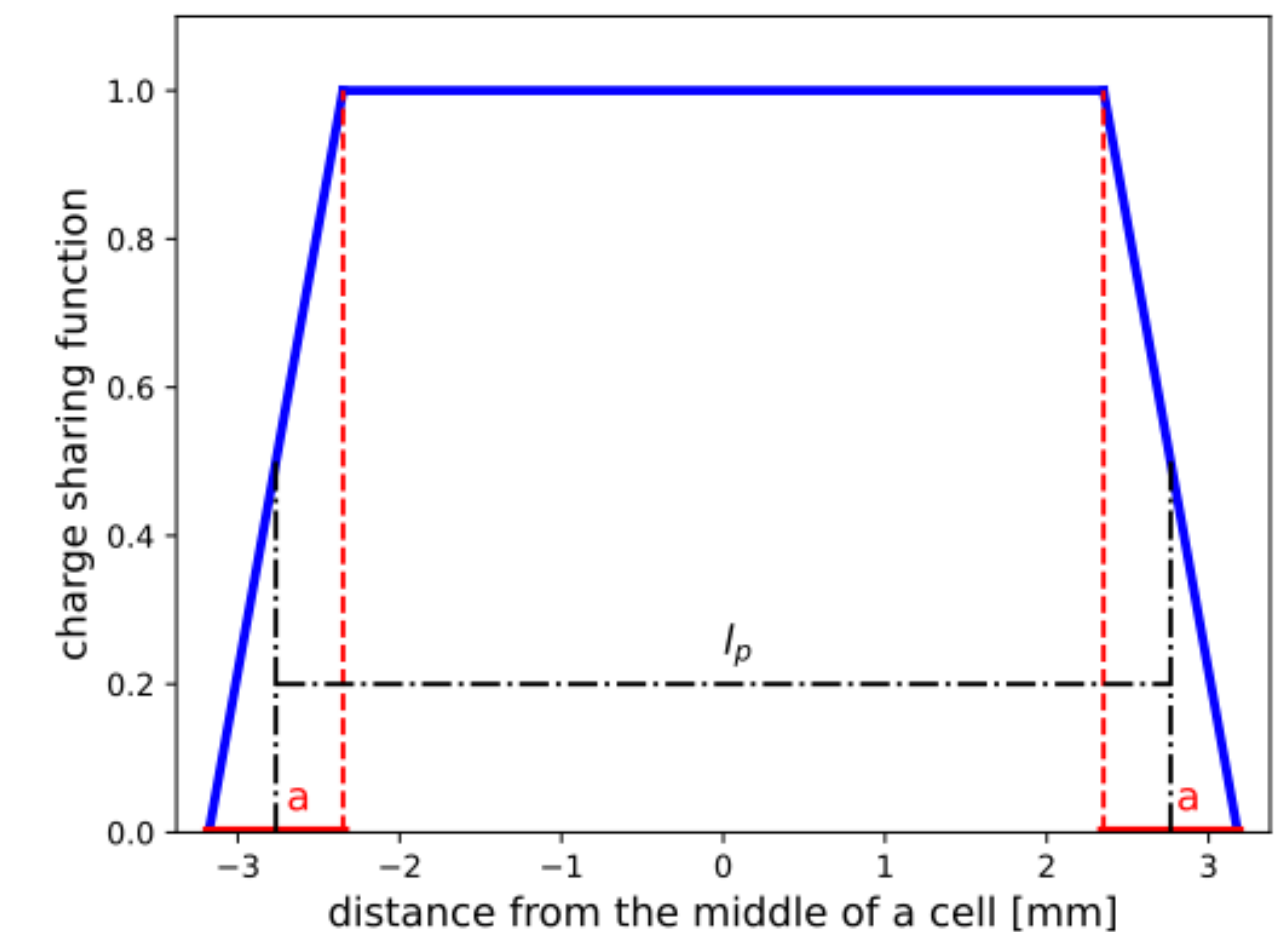
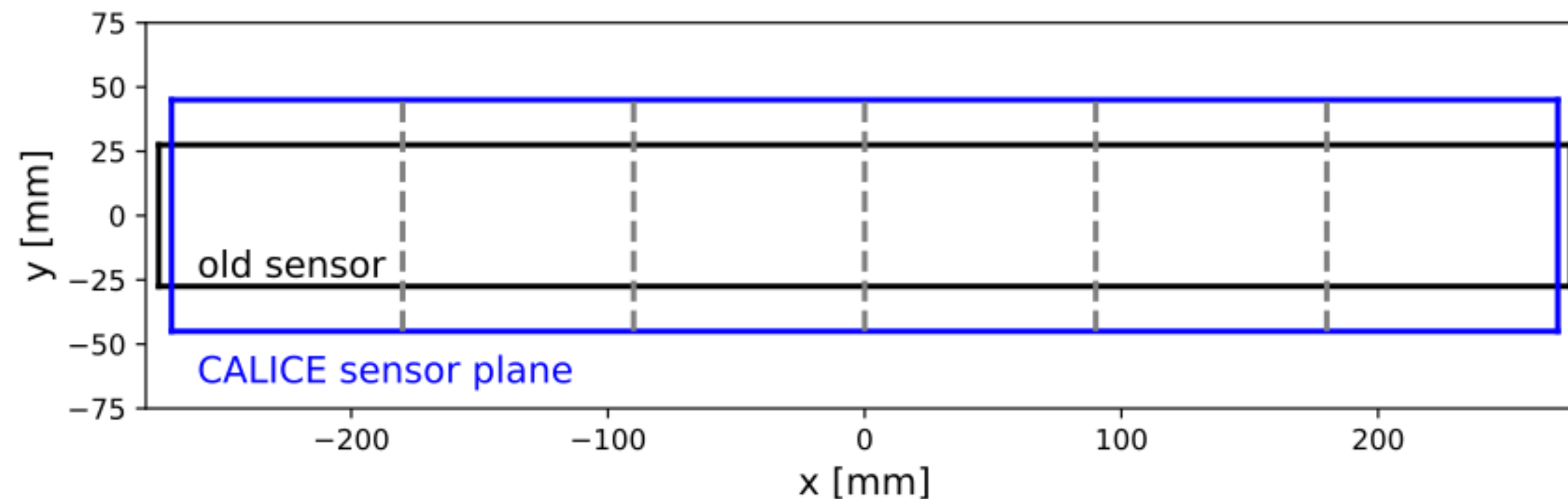
10+5 layers (20 X0)



15 layers (15 X0)

# Re-segmentation

- A new segmentation is going to be implemented in the analysis
  - Pad size: from  $5 \times 5 \text{ mm}^2$  to  $5.5 \times 5.5 \text{ mm}^2$ ; sensor size from  $55 \times 5.5 \text{ cm}^2$  to  $54 \times 9 \text{ cm}^2$
- The digitalization and charge sharing, and the effect of misalignment between layers will be considered



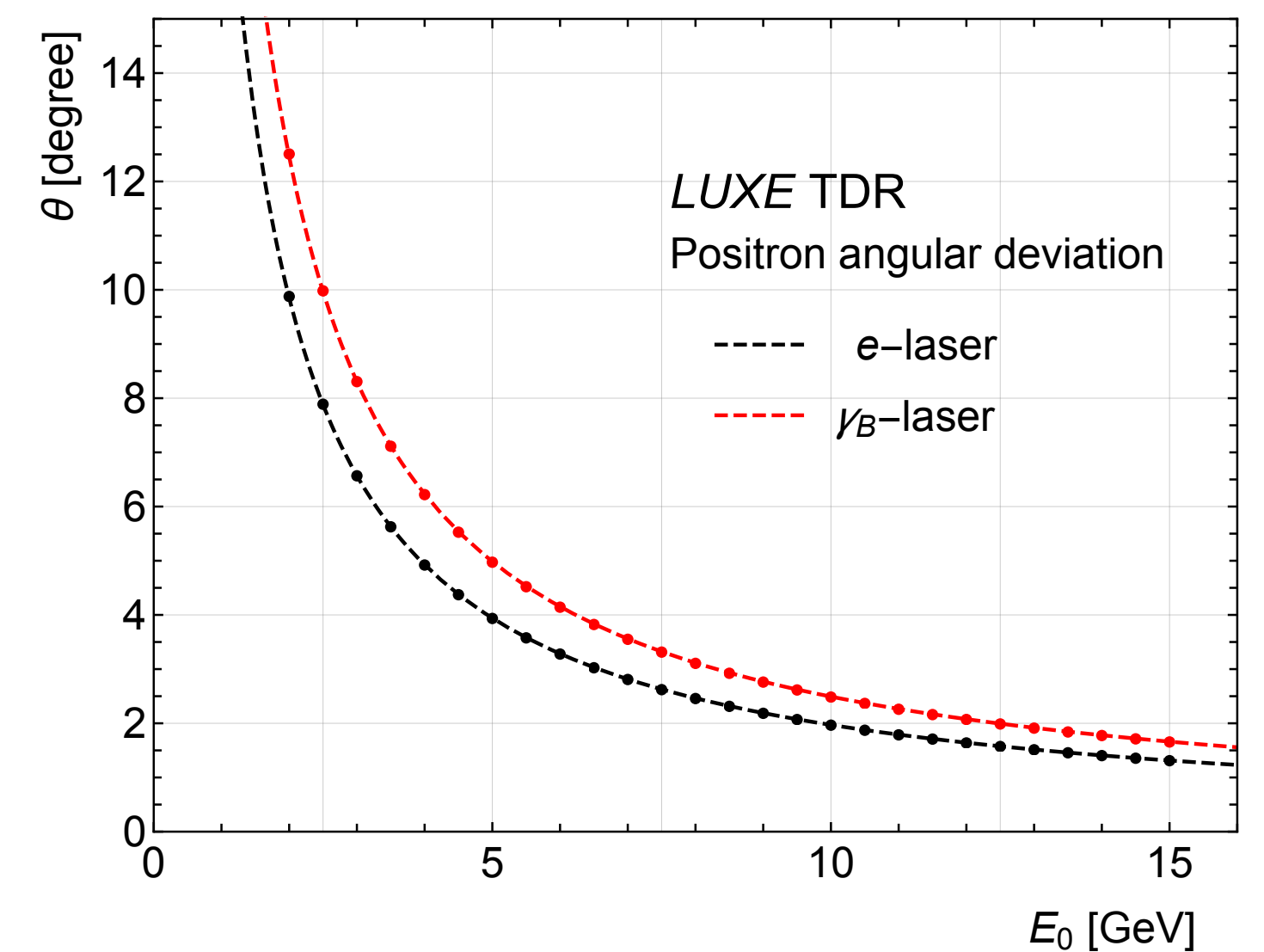
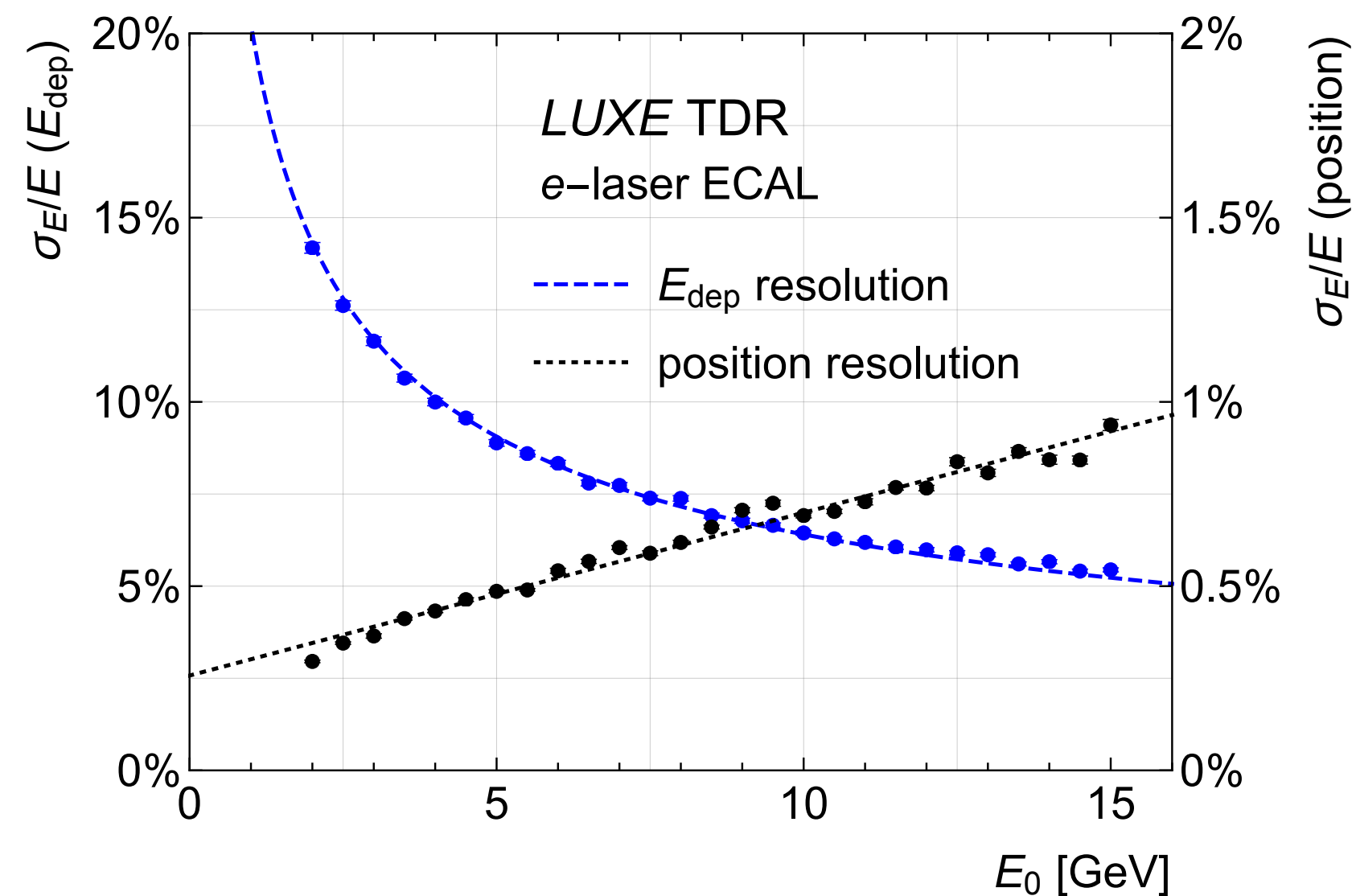
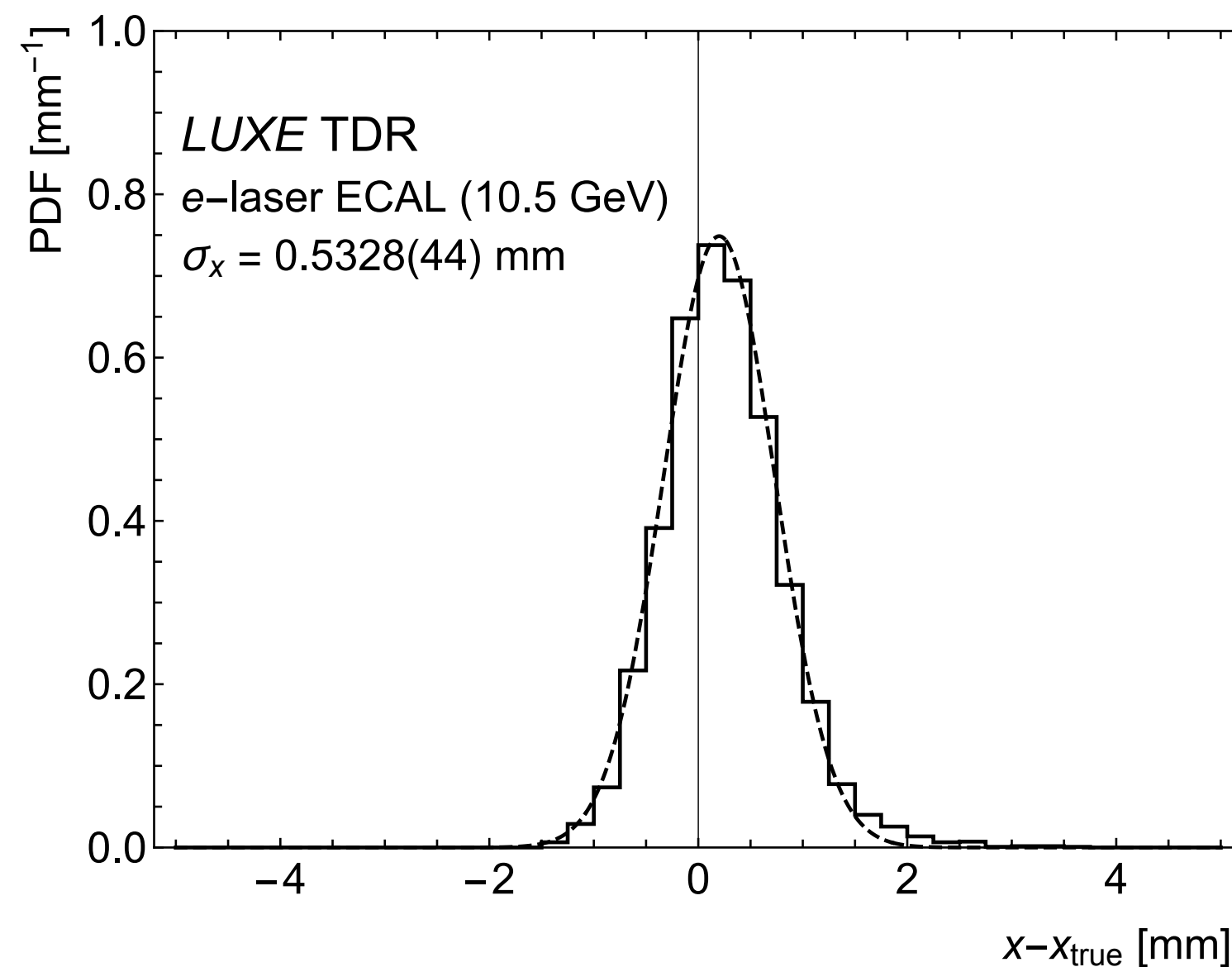
# Outline

- Background
- Calibration and Optimization
- Position reconstruction
- Spectrum reconstruction
- One more thing ...



# Logarithmic weighting

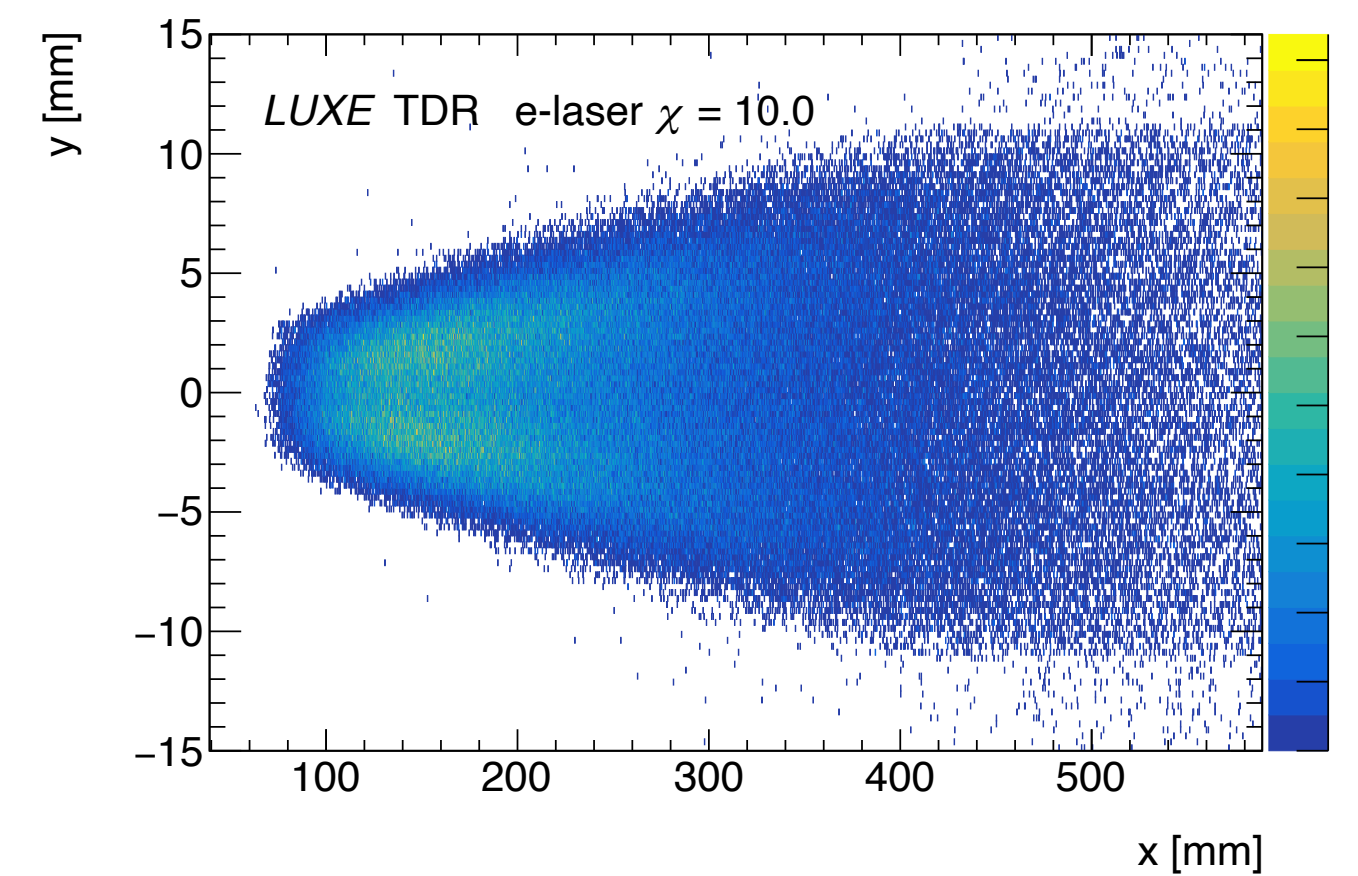
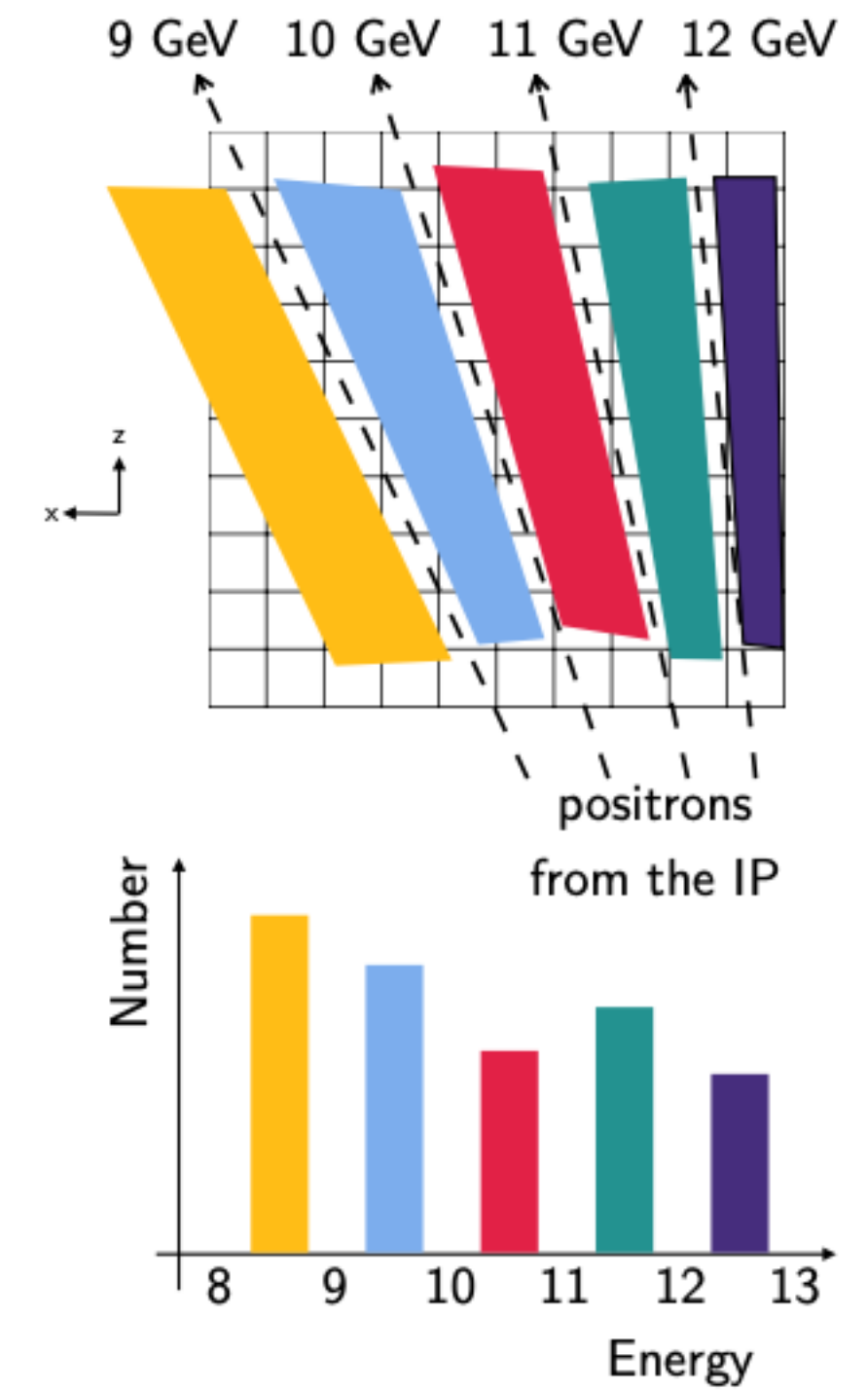
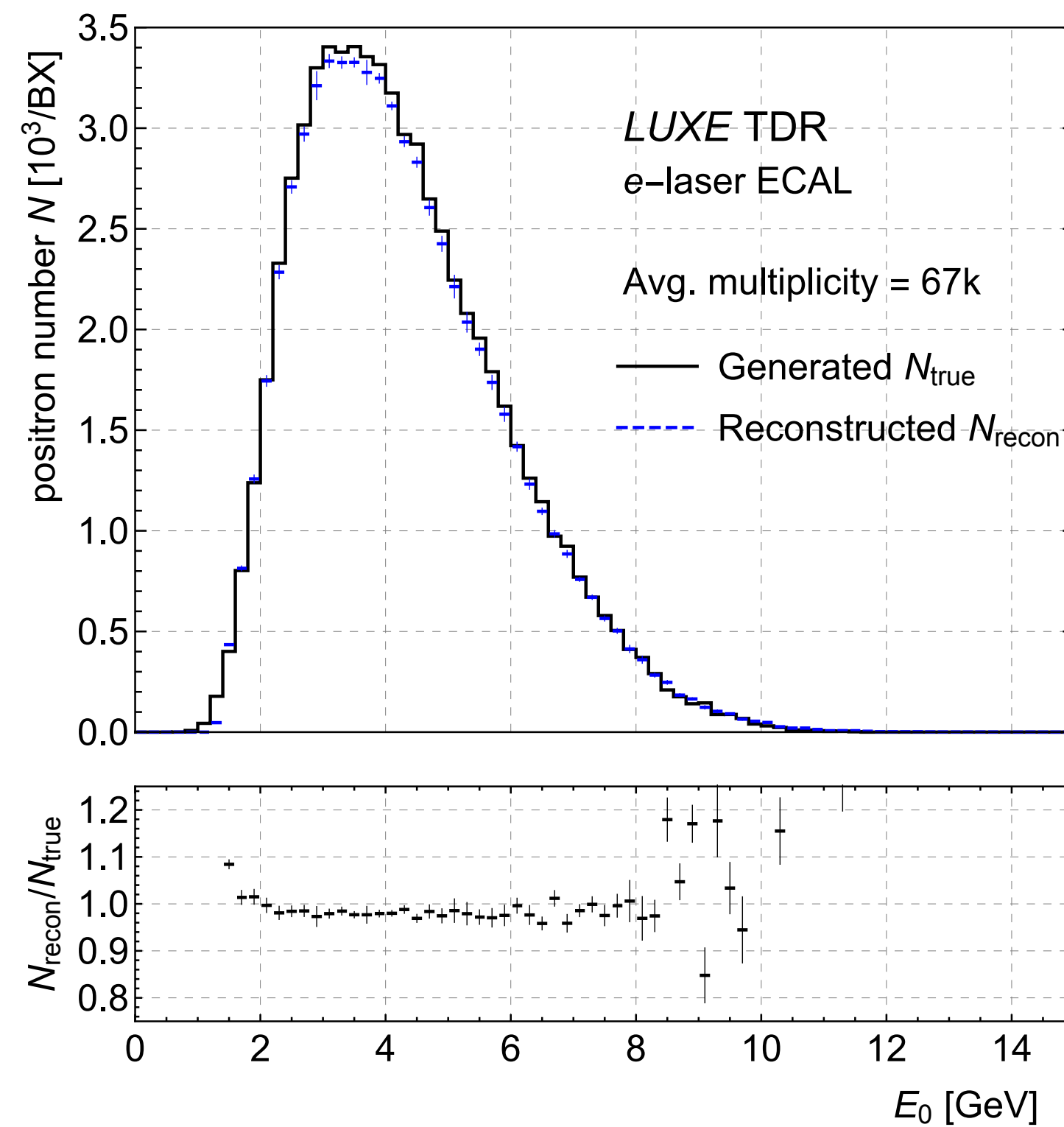
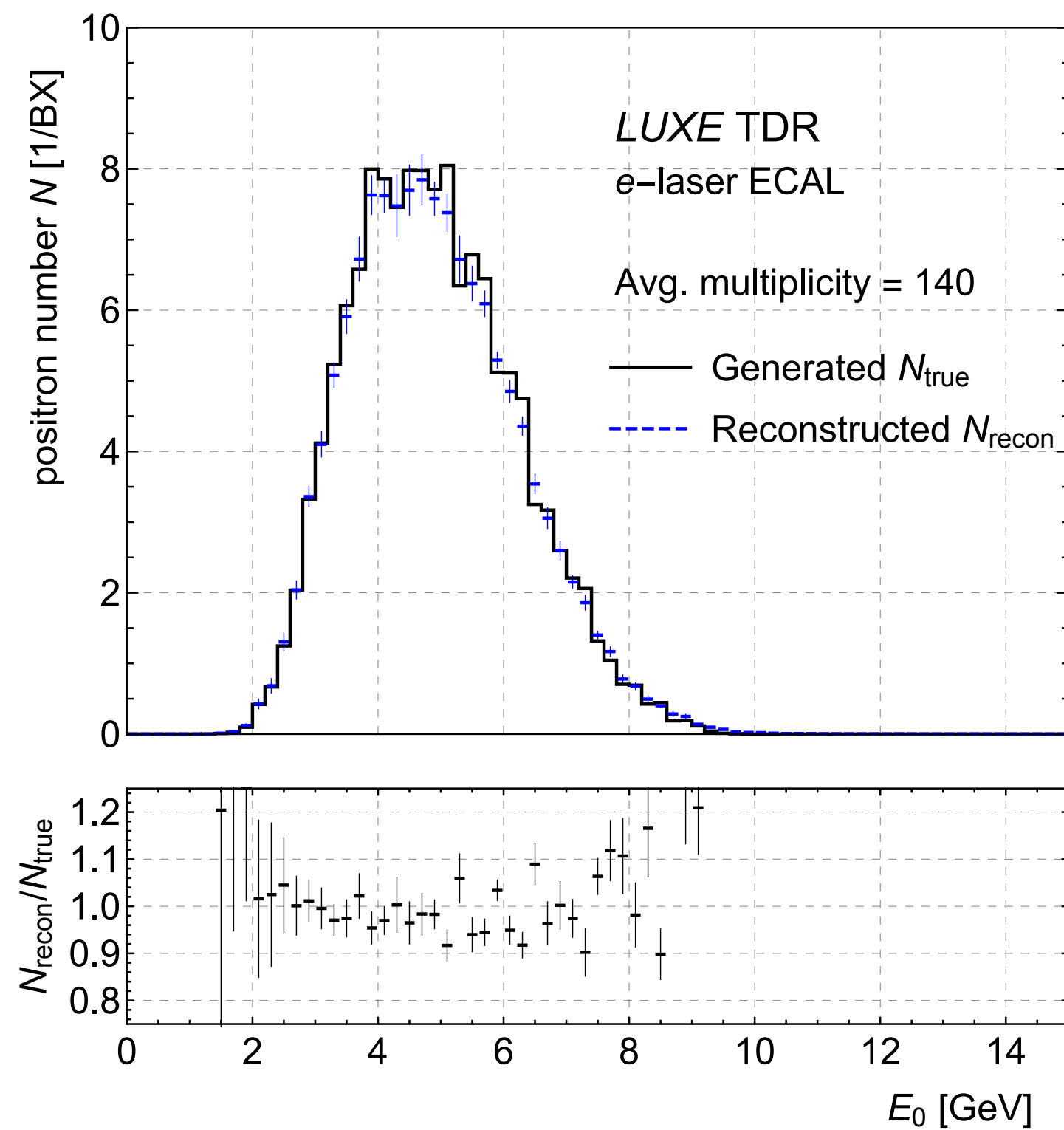
- The weighting method exploits the symmetry of cigar-shaped shower
- By introduced a logarithmic cut-off, the weighting method is more sensitive and more robust against random remote deposits (by photon)
- The magnetic dipole introduces an angle-energy distribution and creates bias in weighting method
- An algorithm based on machine learning is being developed



# Outline

- Background
- Calibration and Optimization
- Position reconstruction
- Spectrum reconstruction
- One more thing ...

# Energy flow



# Reconstruction Network

- The energy flow method is essentially a hand-written “network” that associates the energy deposit of all pads to the bins of energy spectrum
- The idea of using machine learning is to find the weights in an efficient way
- A convolutional neural network is used to “recognize the image”
- The neural network is learning the dependency between position and energy that benefits the resolution

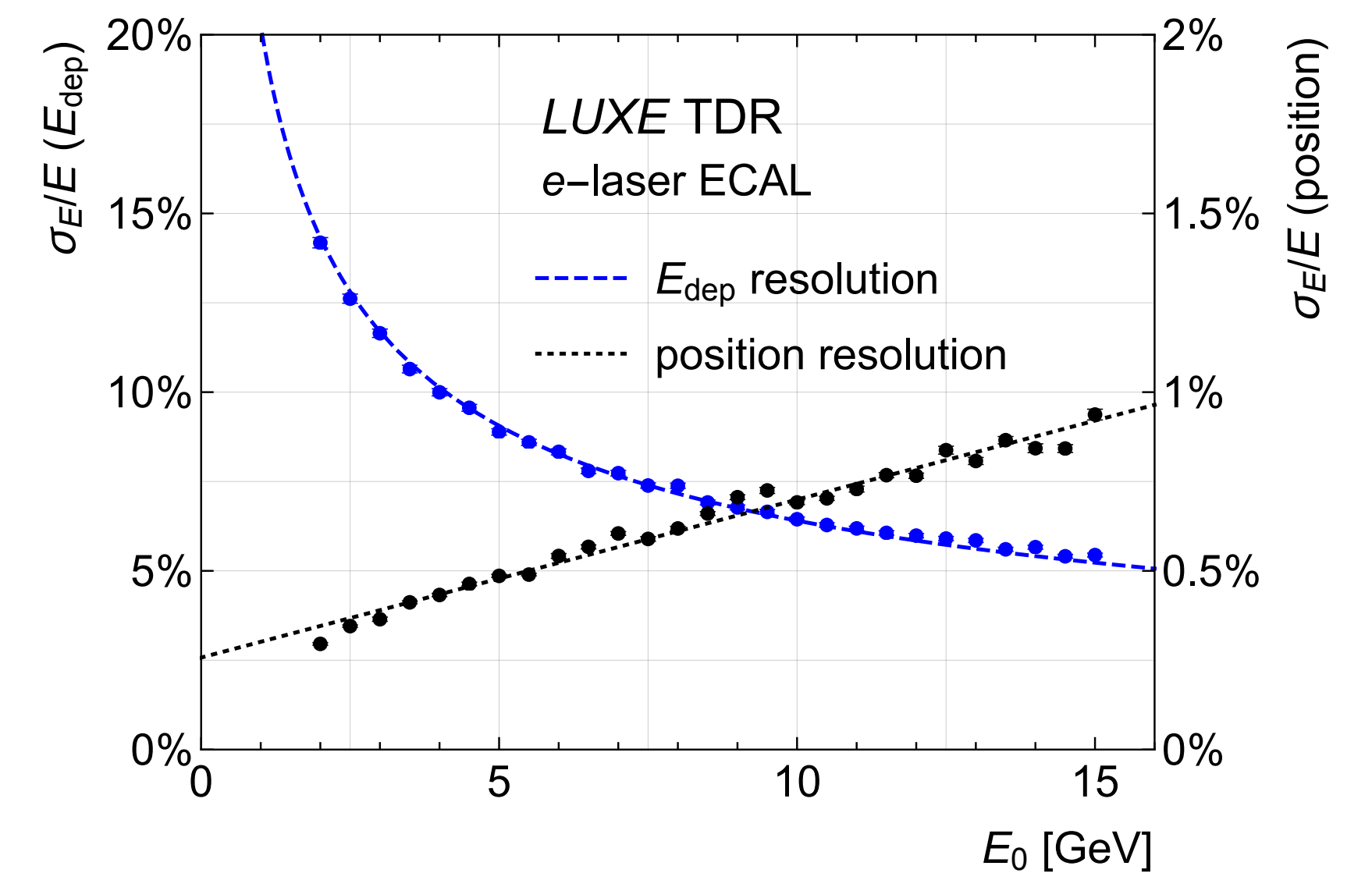
Model architecture

```
=====
```

Layer (type:depth-idx)	Input Shape	Output Shape
Model	[128, 1, 110, 21]	[128, 60]
└─Conv2d: 1-1	[128, 1, 110, 21]	[128, 8, 56, 11]
└─Conv2d: 1-2	[128, 8, 56, 11]	[128, 16, 28, 6]
└─Linear: 1-3	[128, 2688]	[128, 256]
└─Linear: 1-4	[128, 256]	[128, 60]

```
=====
```

Total params: 705,012  
Trainable params: 705,012  
Non-trainable params: 0  
Total mult-adds (M): 118.36

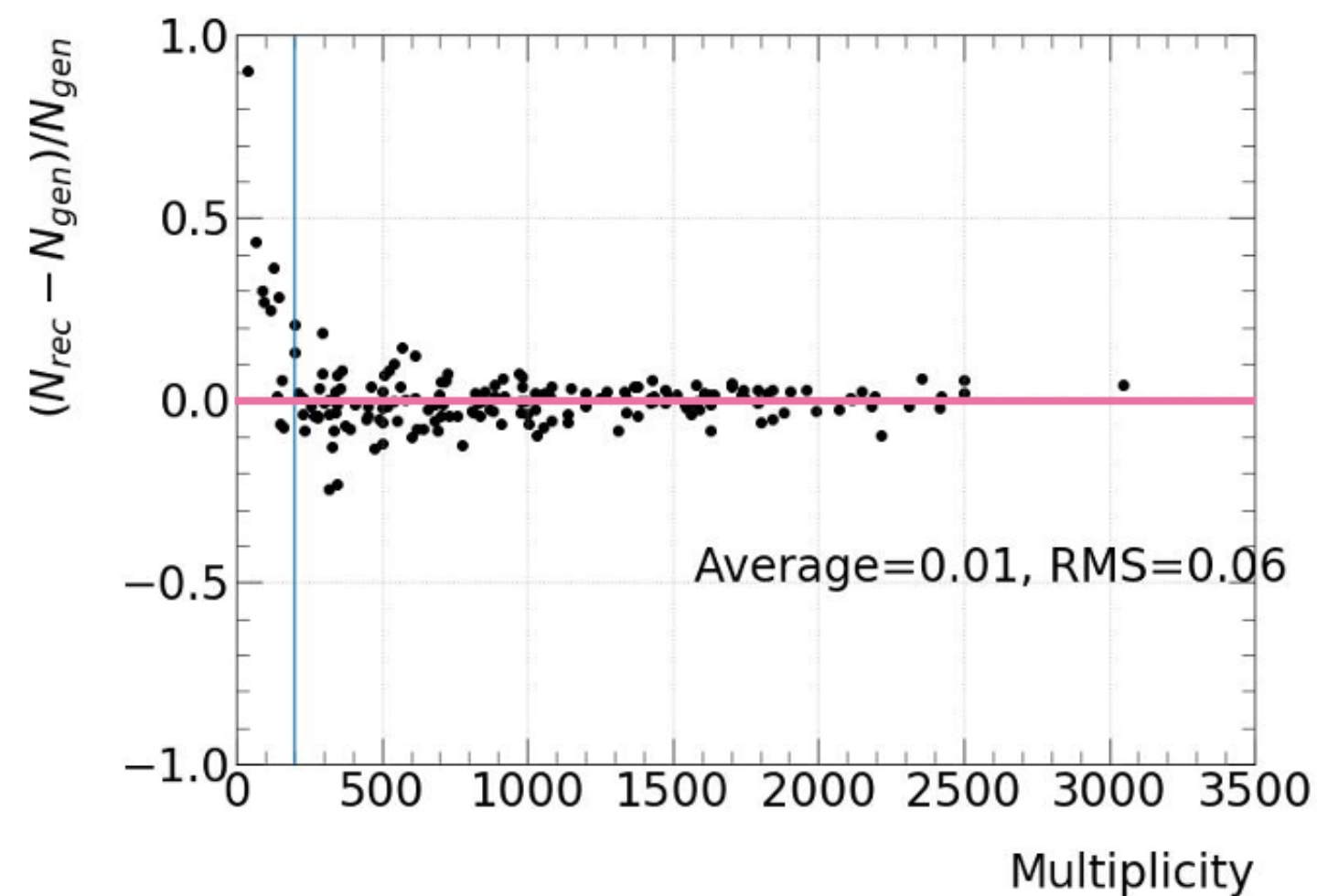




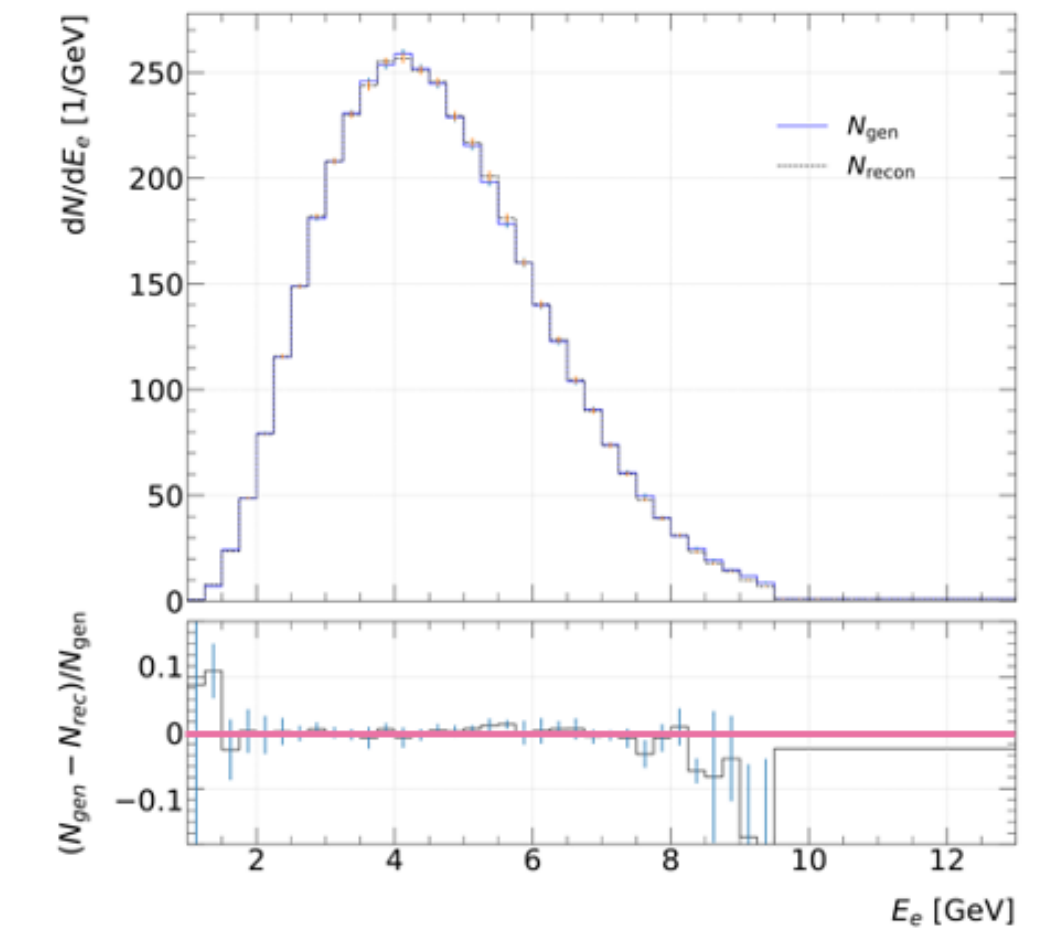
# CNN based algorithm

- The CNN successfully reconstruct the energy spectrum not only with 20 or 15 layers, but also even with only 5 layers above a multiplicity threshold
- The method has a difficulty with generalization
- We are working on stripping out the information of multiplicity

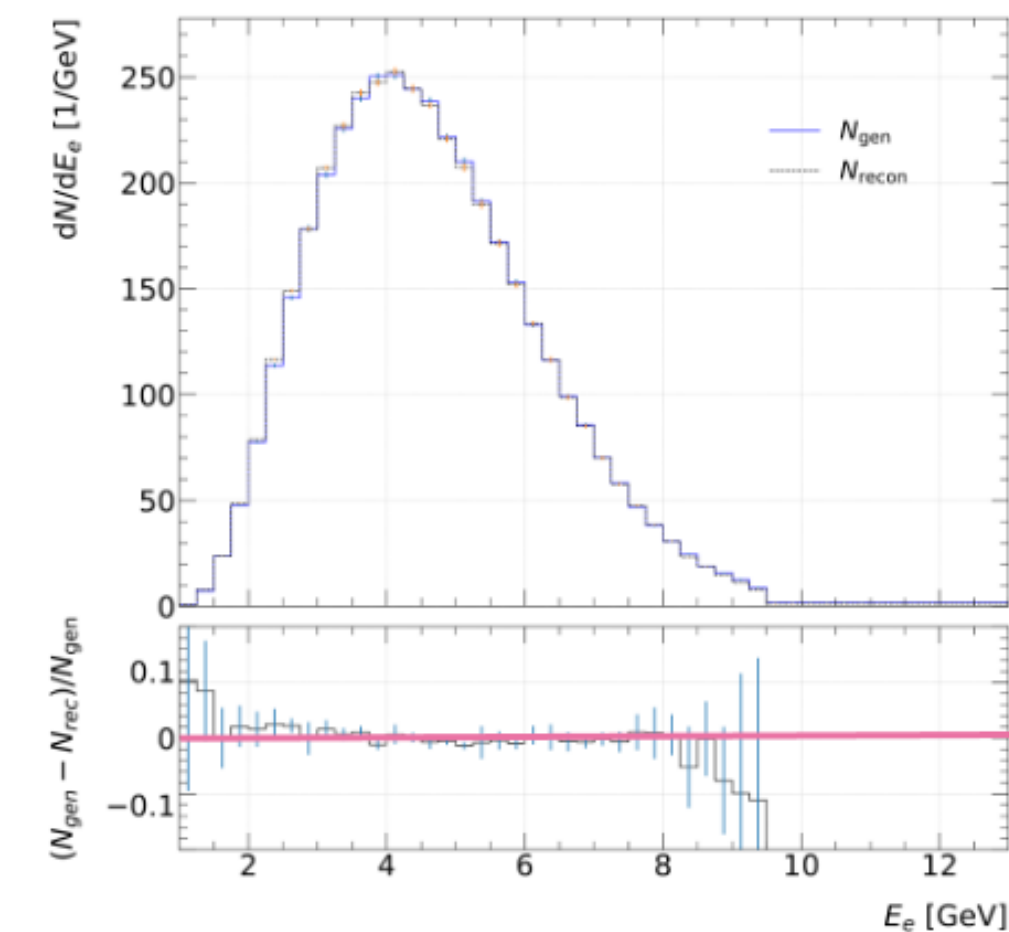
5-LAYER ECAL  
WITH THE BACKGROUND



20-LAYER ECAL  
WITH THE BACKGROUND



5-LAYER ECAL  
WITH THE BACKGROUND

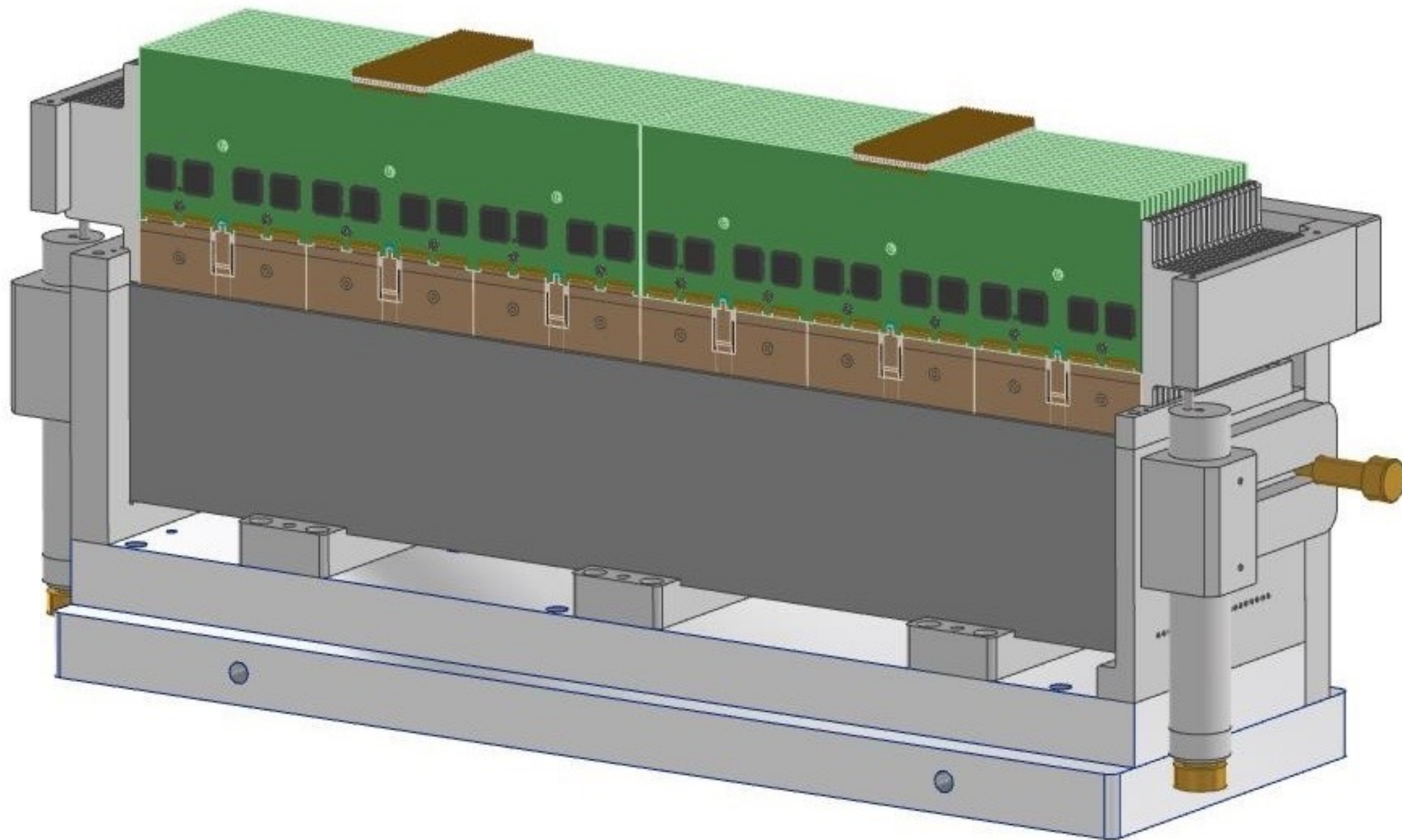


# Outline

- Background
- Calibration and Optimization
- Position reconstruction
- Spectrum reconstruction
  
- One more thing ...

# Summary

- The LUXE ECAL group is working towards the final design of the ECAL-P and a ready-use reconstruction toolkit
- The background study so far confirms that the signal-to-noise ratios are generally acceptable
- A framework has been developed to find out the optimized design of ECAL-P
- Logarithmic weighting is so far one of our best methods for position reconstruction
- A machine learning method based on CNN is successfully reconstruct the positron beam's spectrum with very few  $X_0$ , comparing to the benchmark of energy flow
- Things are still on our list:
  - Uncertainty of background subtraction (fast simulation)
  - Digitalization and other effects
  - Clustering (for the cases of multiplicity around 10)
  - Thorough studies on ECAL-E



**Thank you for your attention!**



Backup of TB