

Exploring position reconstruction of HPGe detector events in LEGEND with a deep neural network



Christoph Seibt¹ (christoph.seibt@tu-dresden.de), Aobo Li², Kai Zuber¹, Frank Siegert¹ and Mariia Redchuk³ on behalf of the **LEGEND Collaboration**

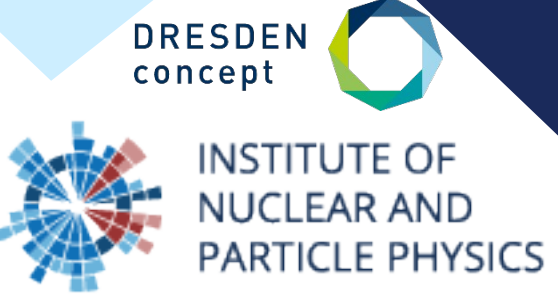
¹ Technical University Dresden

² University of California San Diego

³ University of Padova and INFN



Large Enriched Germanium Experiment for Neutrinoless $\beta\beta$ Decay



Introduction

The LEGEND experiment:

- Goal: Measure $0\nu\beta\beta$ decay in ^{76}Ge
- LEGEND-1000: Sensitivity on half-lives of up to 10^{28}y
- To reach this sensitivity, background reduction is very important

Motivation for position reconstruction:

- The position of an event inside a Ge-detector can indicate signal or background origin:
 - Surface events and events near the contact
 - Spot local detector impurities
- Position reconstruction has purpose beyond background reduction:
 - Exploring possibilities of pulse shape simulation and machine learning

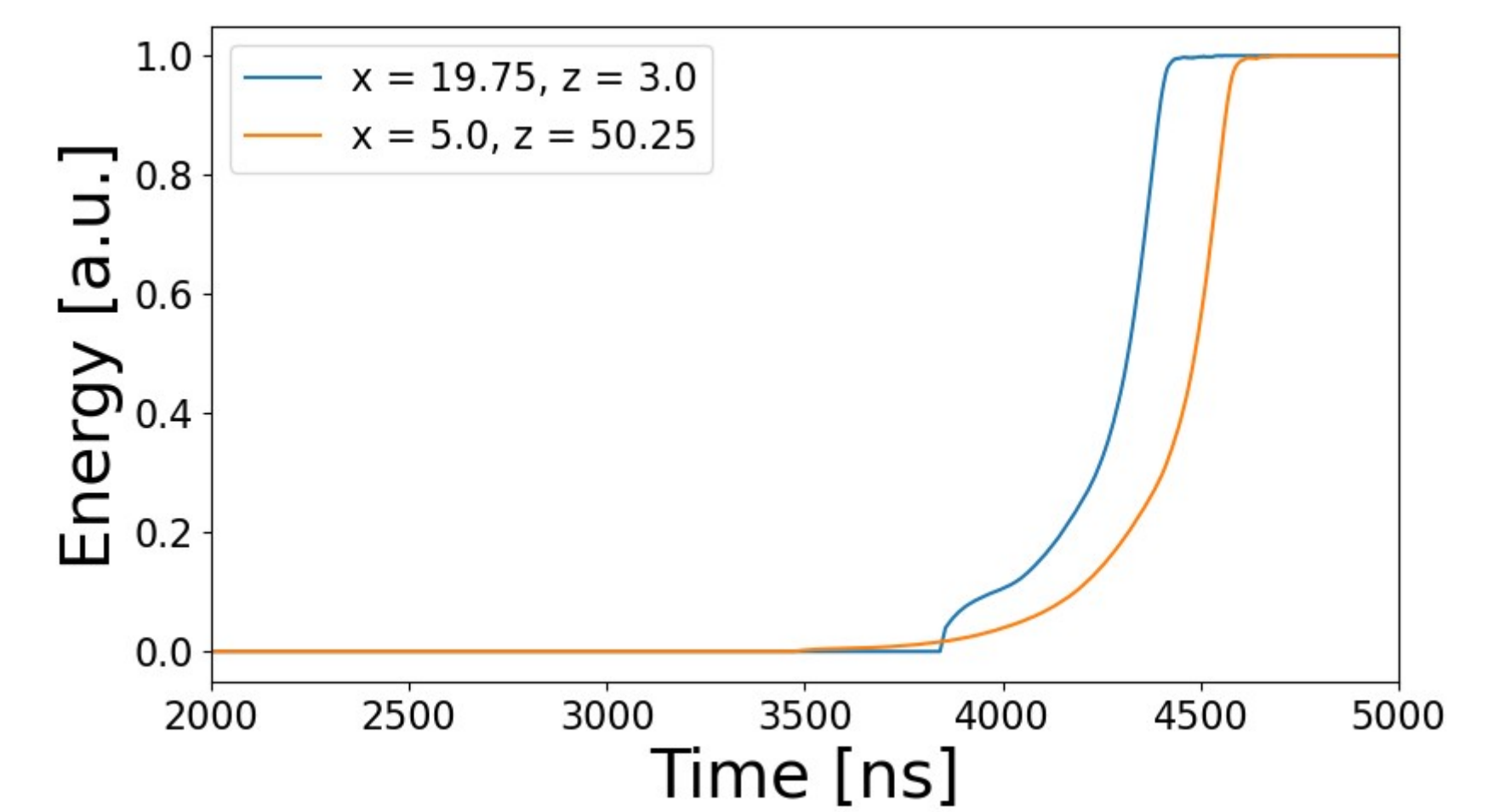
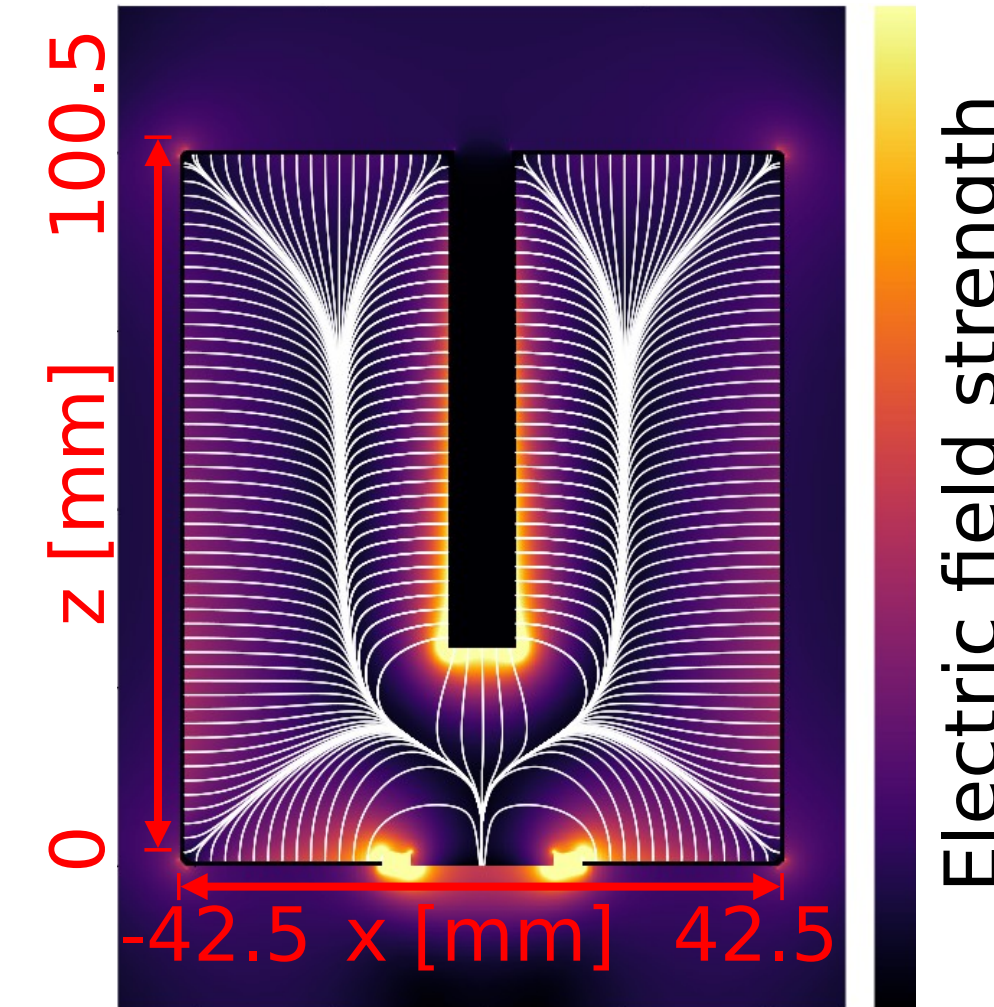
Germanium detectors and pulse shape simulation

LEGEND-1000 mainly uses inverted coaxial point contact (ICPC) detectors (left)

- Can be produced with comparatively high masses ($\sim 3\text{ kg}$), pulse shape discrimination works well

Pulse shape simulation:

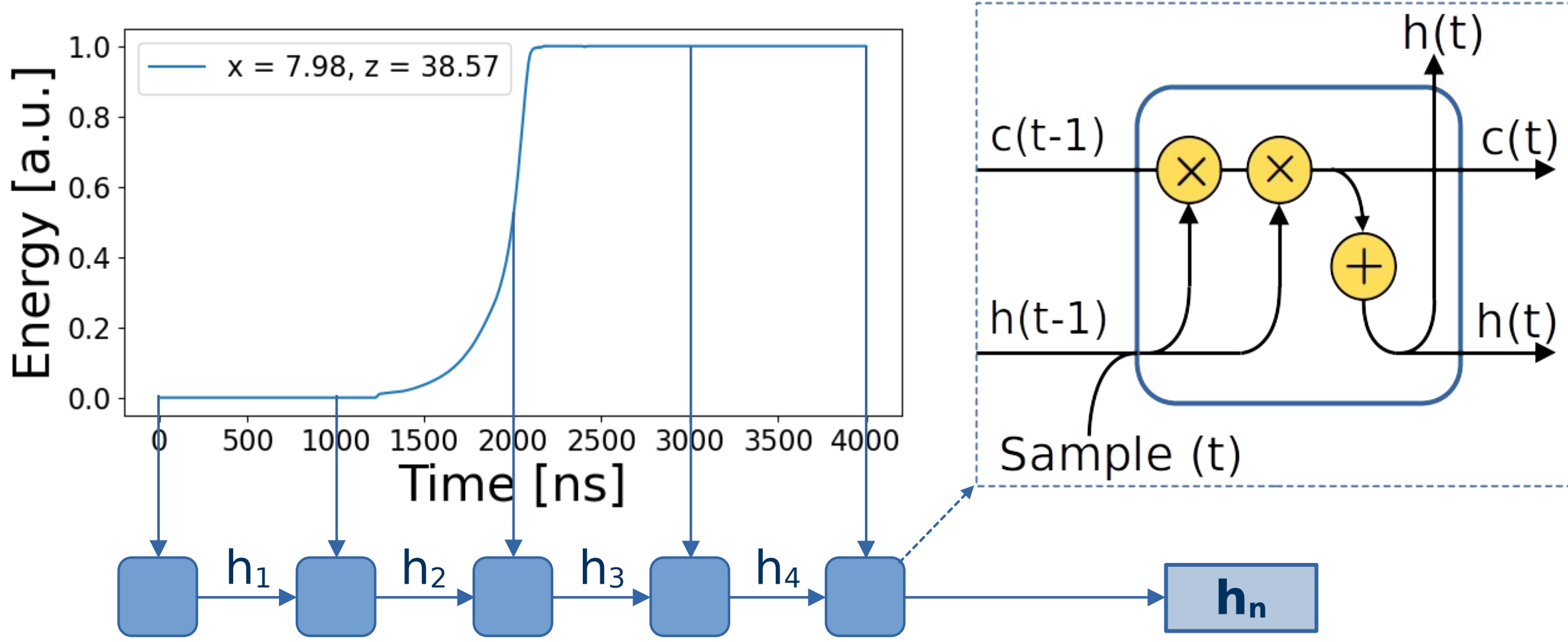
Simulate detector geometry and electric field \rightarrow Energy deposition creates charge cloud \rightarrow Charge cloud measured at point contact as waveform, which depends on event position



LSTM with attention mechanism

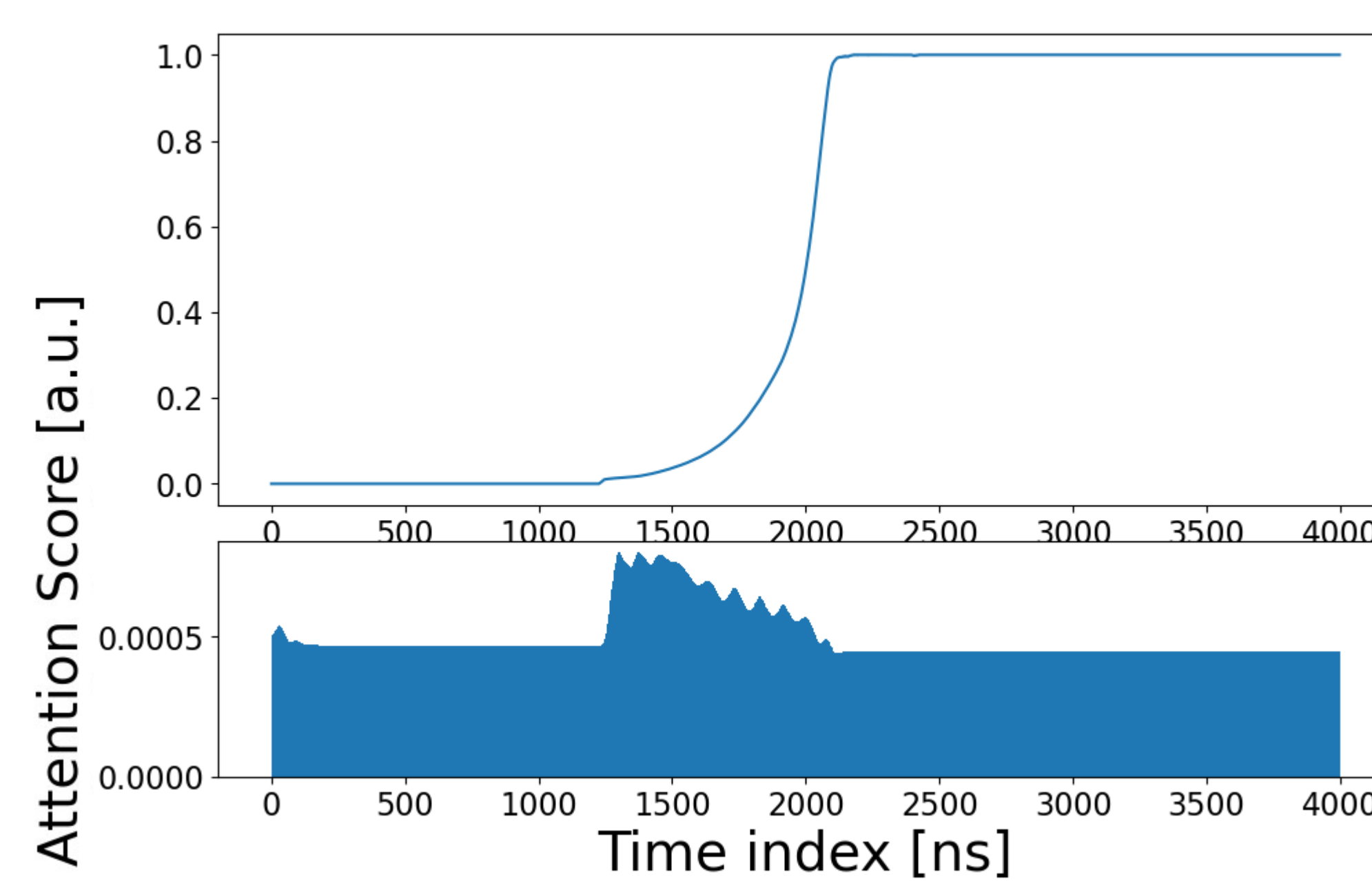
Long-Short-Term-Memory (LSTM):

- Iterates an LSTM unit over each waveform sample
- Hidden state $h(t)$ stores short-term memory, cell state $c(t)$ stores long-term memory



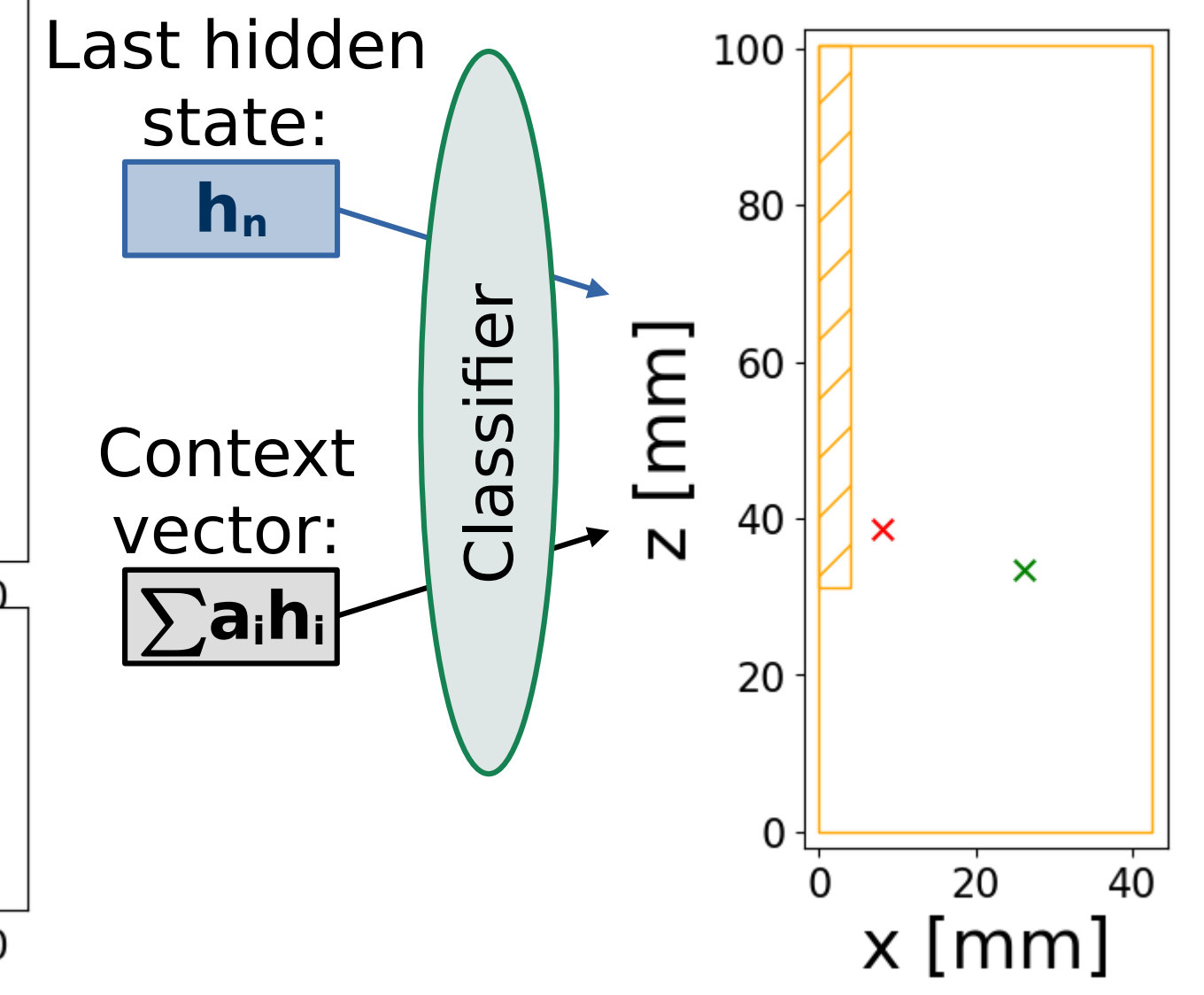
Attention mechanism:

- Highlight important hidden states to optimize reconstruction



Result:

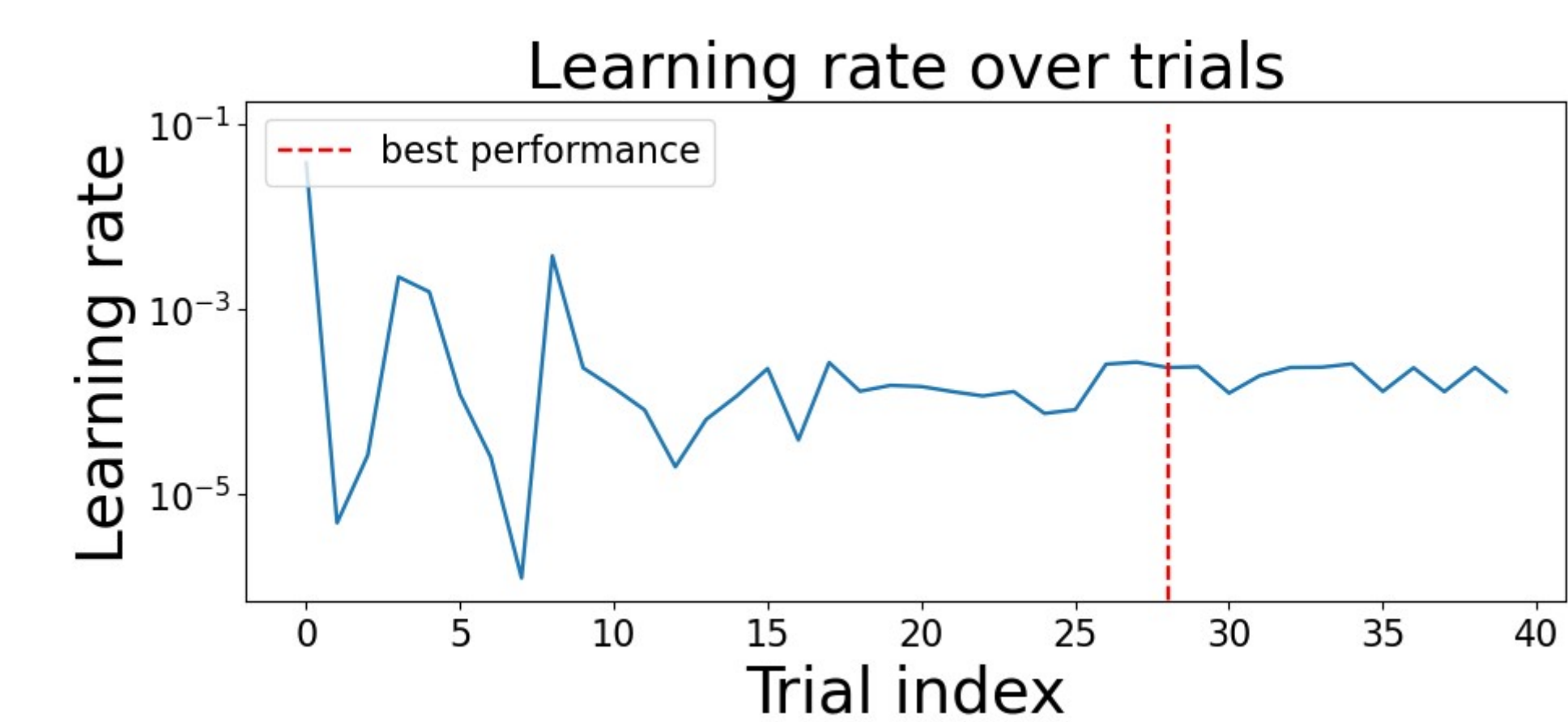
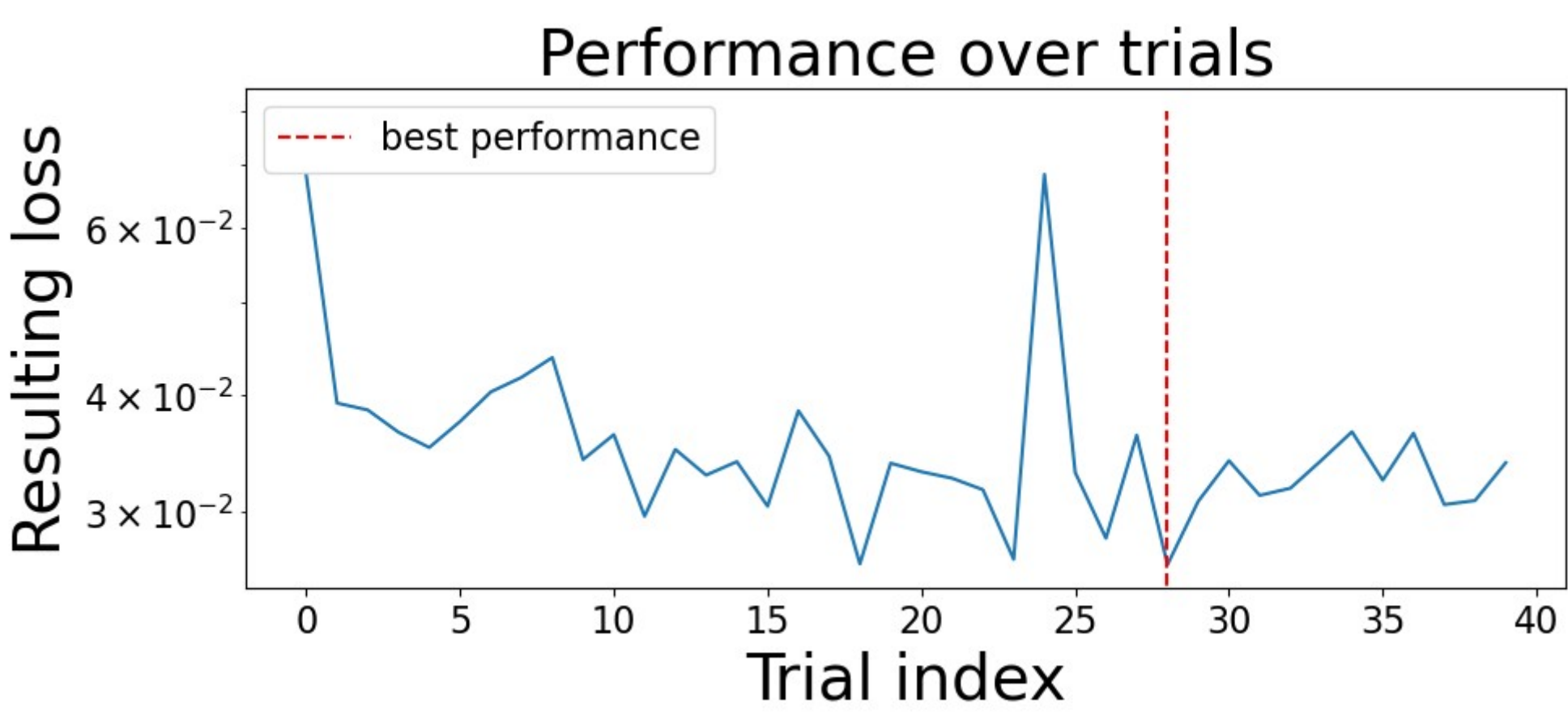
- \times reconstructed position
- \times original position



Results: Training the LSTM on simulated pulses

Optimization of the neural network:

- Precision of a neural network depends on various fixed parameters
- Bayesian optimization: Choose parameters of the next training based on previous trainings \rightarrow Efficient optimization
- Optimization with 120000 simulated events and 50 iterations over the dataset (epochs)

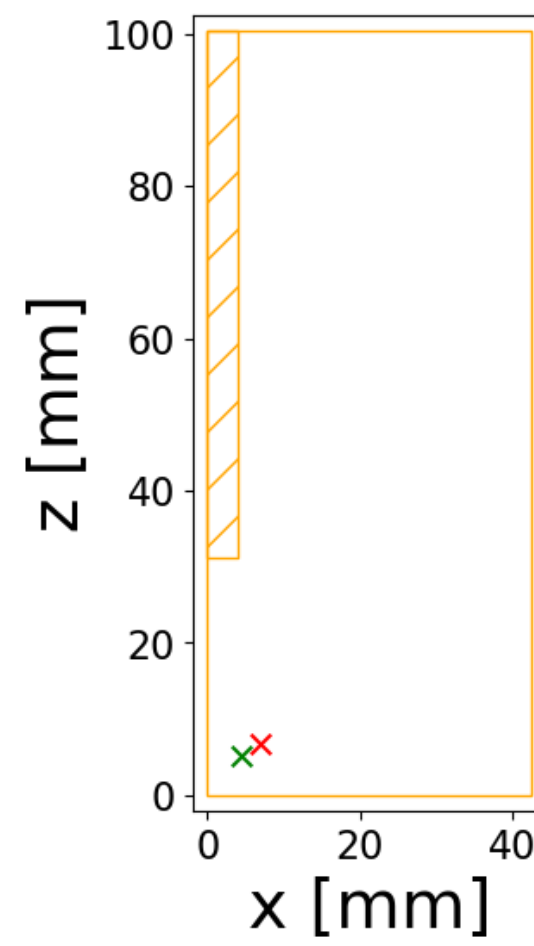


Extended training with optimized parameters

- 240000 events, 100 epochs \rightarrow Higher stability

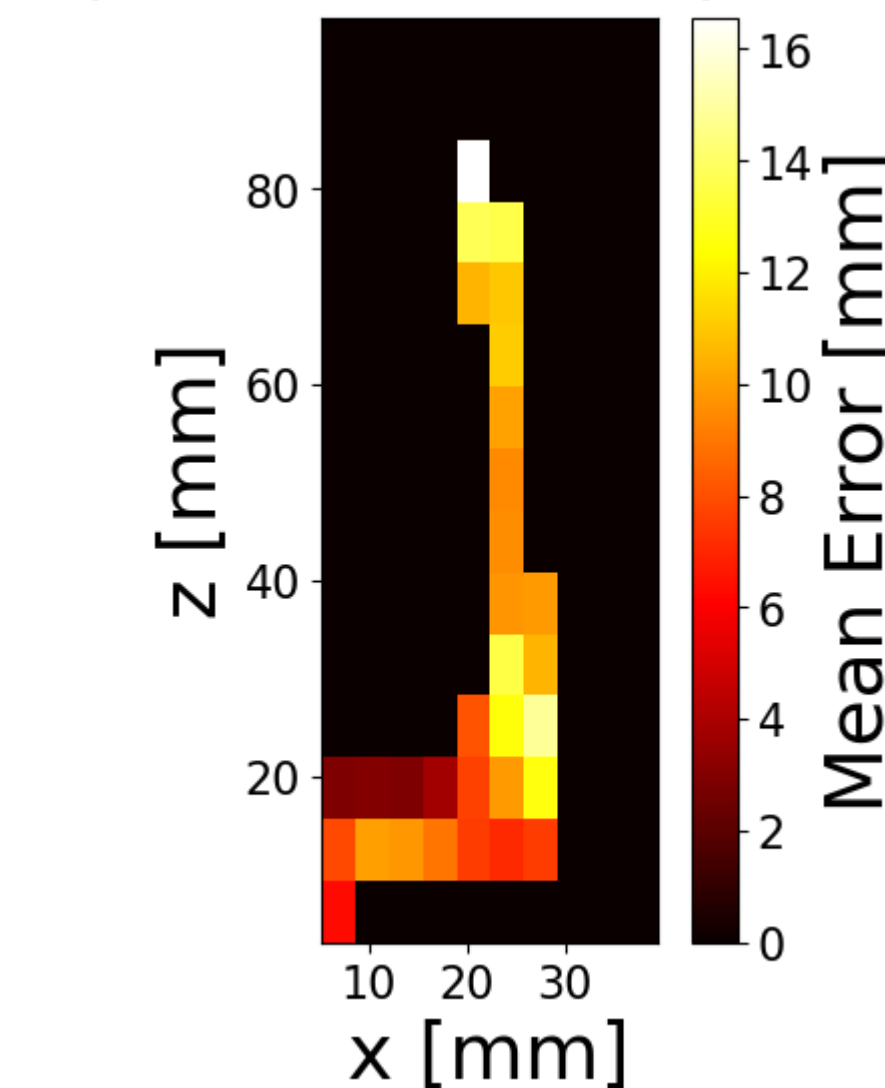
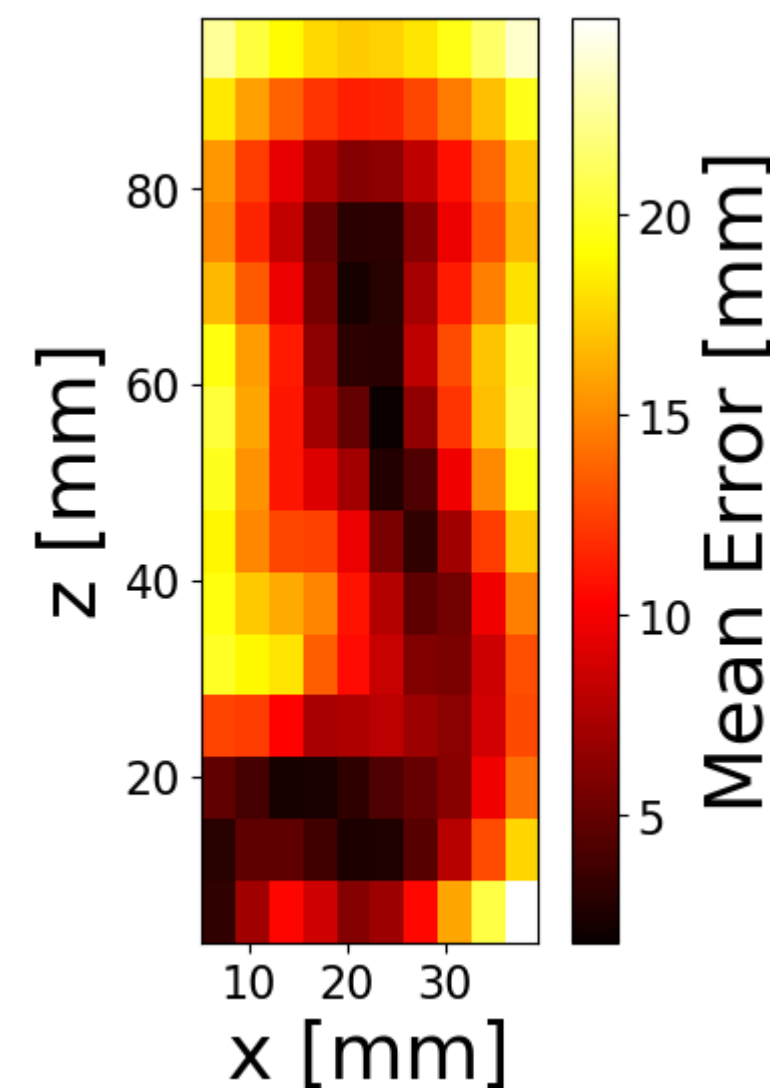
Average errors:

	Absolute error [mm]	Relative to detector size [%]
dx	7.44 ± 0.02	17.5 ± 0.1
dz	7.05 ± 0.02	7.01 ± 0.02
dr	11.00 ± 0.02	19.68 ± 0.05



Error heatmaps:

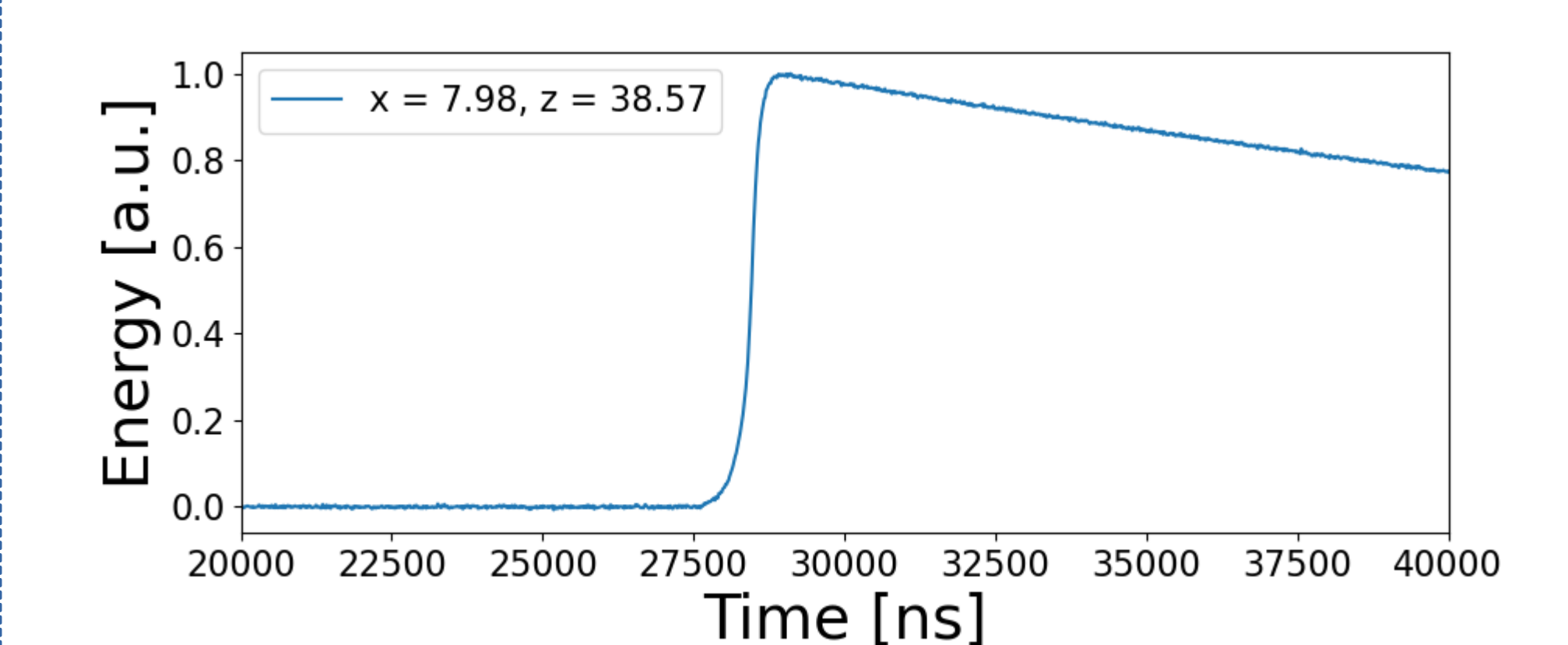
by original position: by reconstructed position



- x-reconstruction only possible close to contact
- (Wanted) property of ICPC detectors
- z-reconstruction works better (7% average error)

Position reconstruction of realistic waveforms

- Adding features of the data acquisition (electronic noise, preamplifier) to mimic real data waveforms



Train network to reconstruct z-coordinate of events:

Results:

- **Average Error [mm]:**

15.95 ± 0.05

- z-reconstruction good for near-point-contact events
- z-reconstruction difficult for events far from contact
- Such events have waveforms with a long rising edge
- The start of the rising edge is "hidden" by noise

- **Reconstruction limit of approximately 70 mm**

Error heatmap:

by original position:

