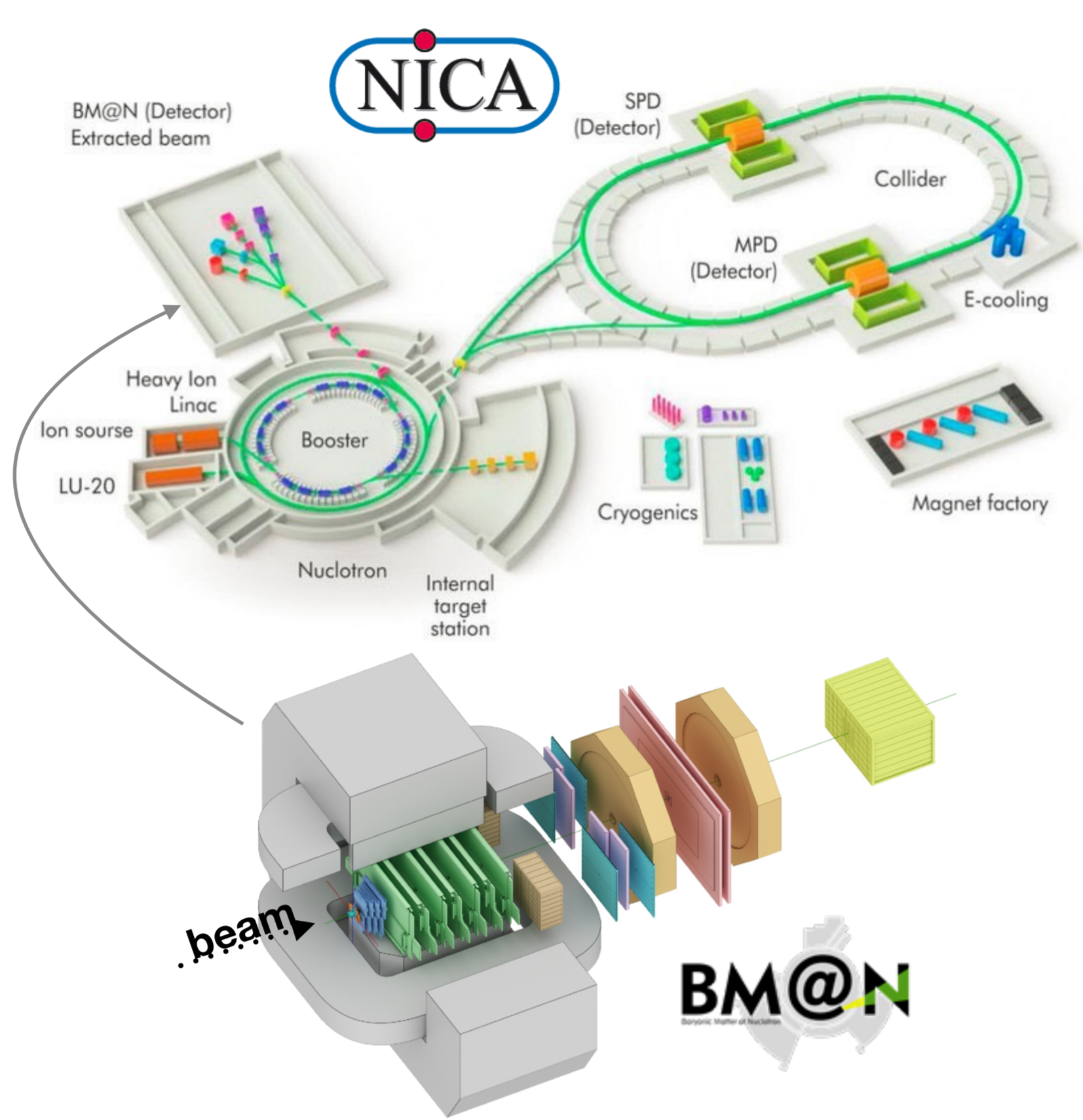


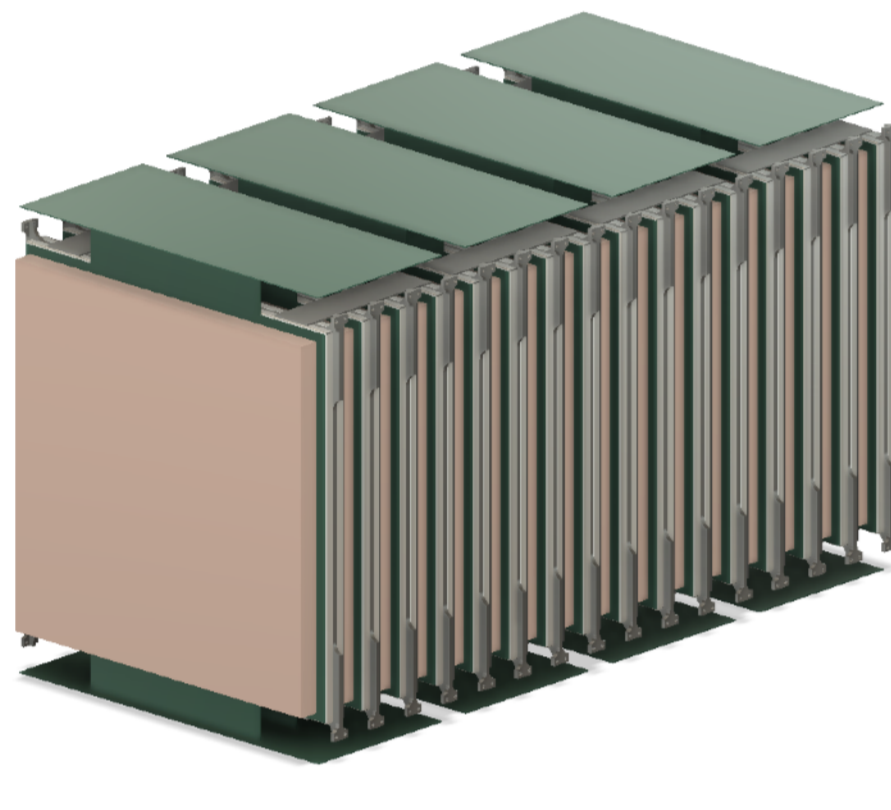
BM@N experiment

Studies of **Baryonic Matter** at the Nuclotron (NICA, JINR)

- Fixed target experiment
- Heavy-ion beam with energies up to 4.5 GeV/nucleon
- investigate the equation-of-state (EOS) of dense nuclear matter which plays a central role for the dynamics of core collapse supernovae and for the stability of neutron stars.
- neutron azimuthal flow** - new tool for EOS studies

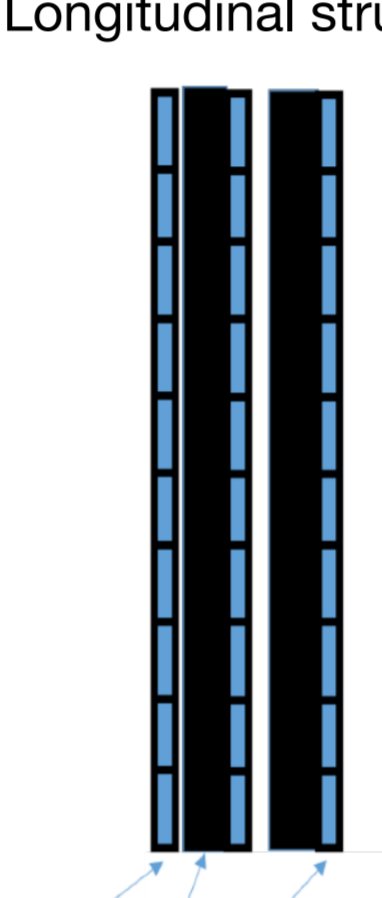


Highly granular time-of-flight neutron detector (HGN)

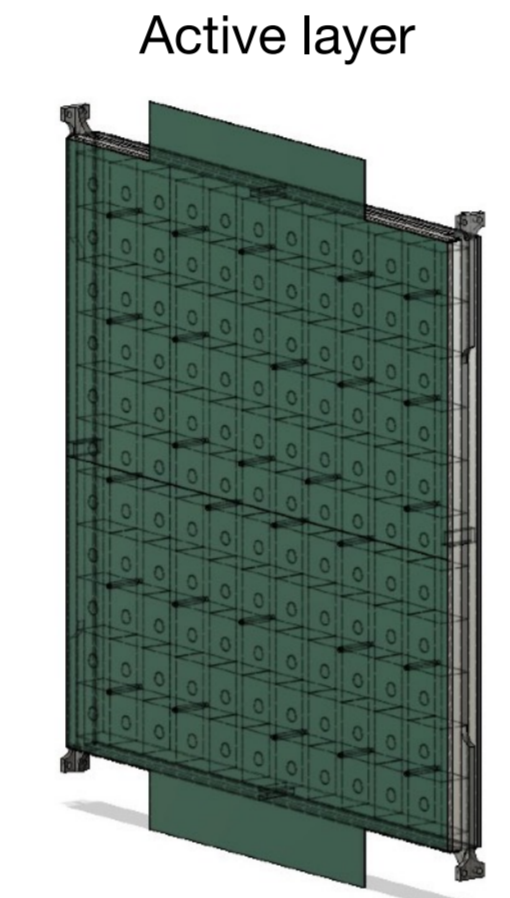


- Total length: **~1m** (~3 λ_n)
- Transverse size: **44x44 cm²**
- 16 layers: **3 cm Cu (absorber) + 2.5cm Scintillator + 0.5cm PCB**
- 11x11 scintillator cell grid**

Longitudinal structure

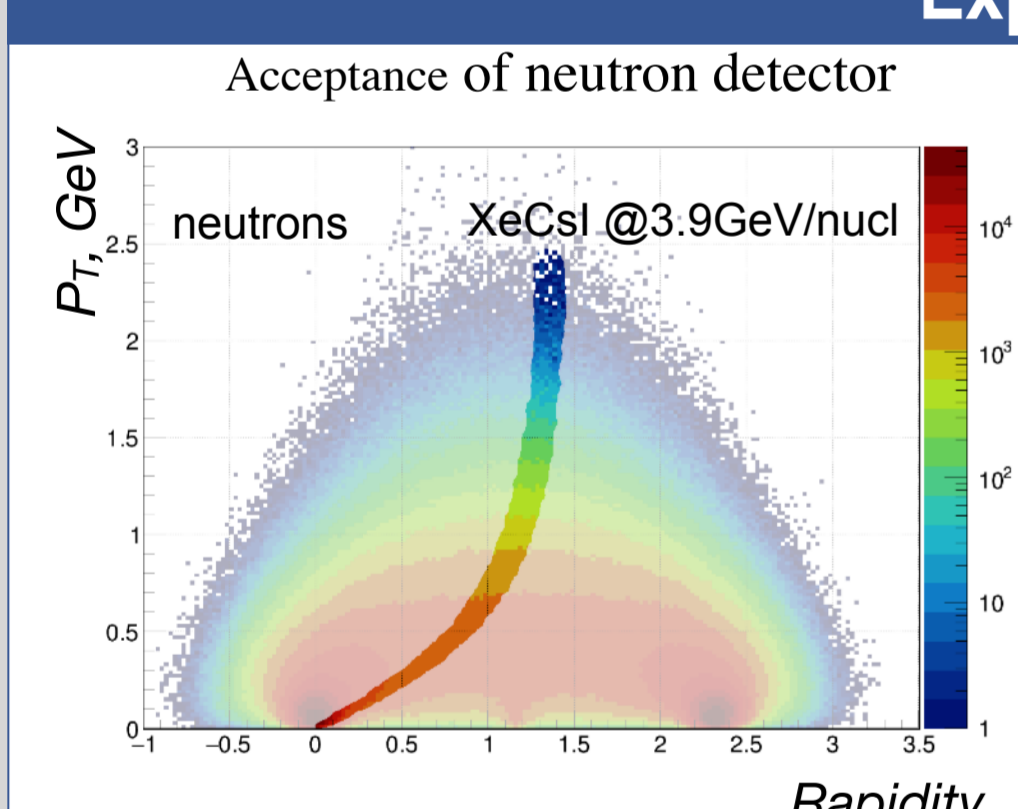


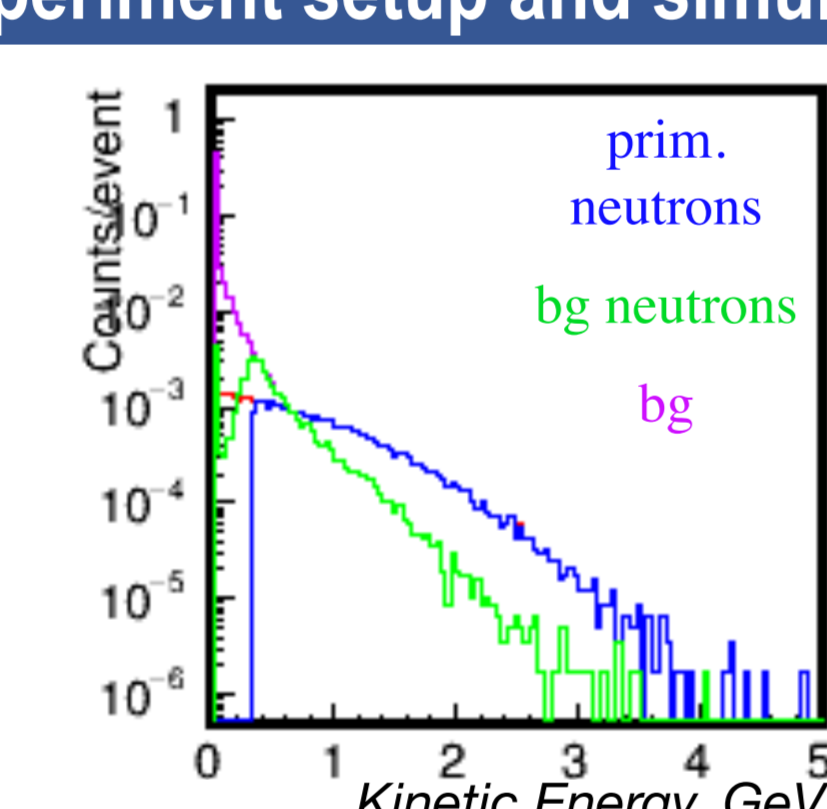
Active layer

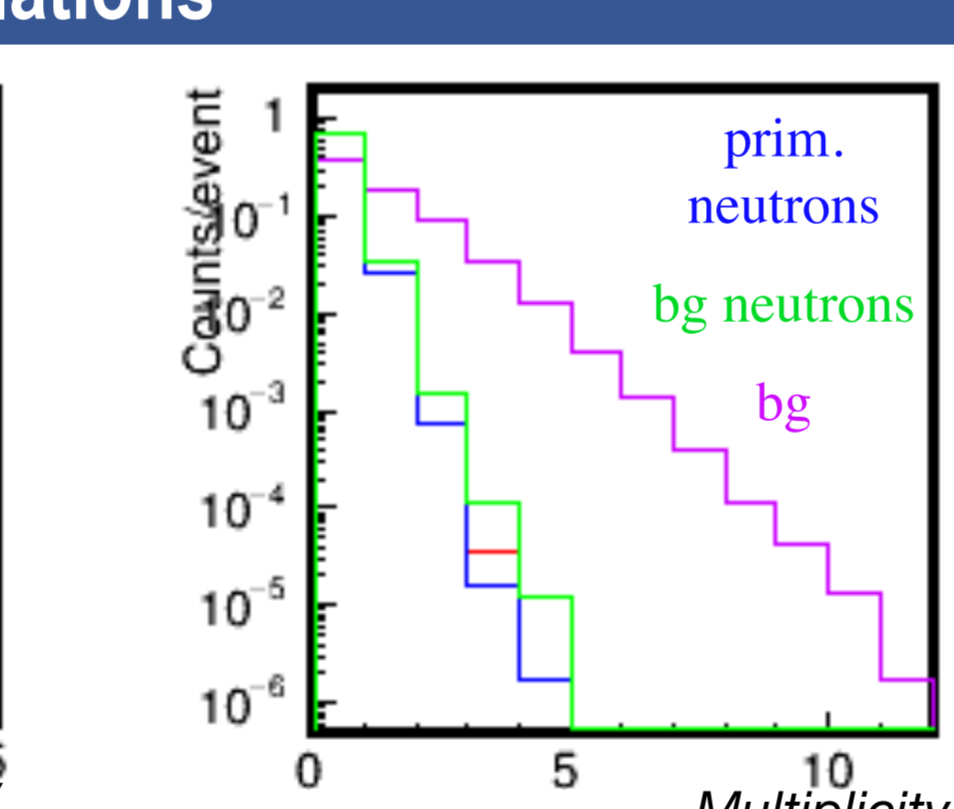


- scintillator cells:
 - size: 4x4x2.5 cm³,
 - total number of cells: 1936
 - light readout: SiPM,
 - expected time resolution: ~150 ps

Experiment setup and simulations







Neutron detector is located at **27° to the beam axis at ~6m from the target**

- DCM-SMM event generator + Geant4
- Neutron rate in acceptance $\sim 2.6 \cdot 10^7$ n / month
- Reconstruction goal: identify neutrons and reconstruct energy on event basis
- Particle multiplicity $\approx 1 \Rightarrow$ **event classification approach**

Observables per hit:

- $(x, y, z)_{hit}$
- $E_{dep} (> 3 \text{ MeV})$
- $T_{hit} + \mathcal{N}(0, \sigma = 150 \text{ ps})$

Dataset

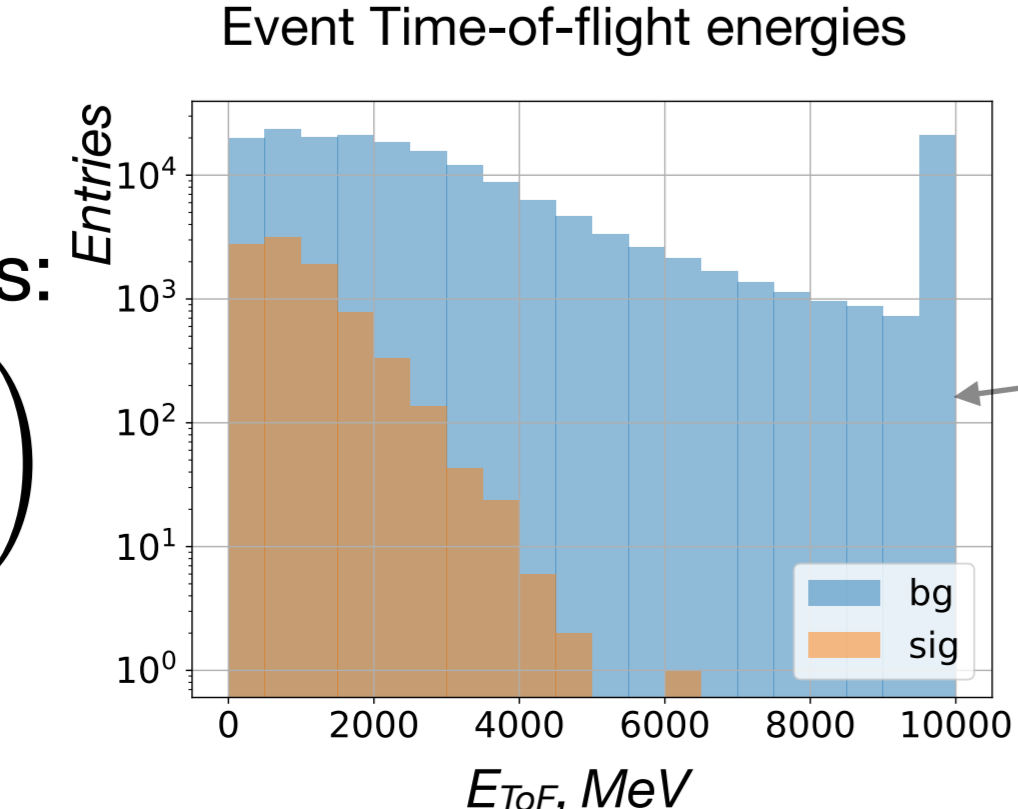
	E_{dep} in acceptance	n in acceptance	signal
N_{events}	206K	~25K	~9K

Signal event selection:

- Single neutron,
- $E_{kin} > 100 \text{ MeV}$,
- Angle to detector axis $< 10^\circ$
- $\delta(E_{ToF}) < 40\%$
- ~36% neutrons selected

Time-of-flight (ToF) energy for n hypothesis:

$$E_{ToF} = m_n \left(\frac{1}{\sqrt{1 - \beta^2}} - 1 \right)$$



Distribution of E_{ToF} has visible separation power however it is our **target variable** and it depends on simulations - possible **source of bias**

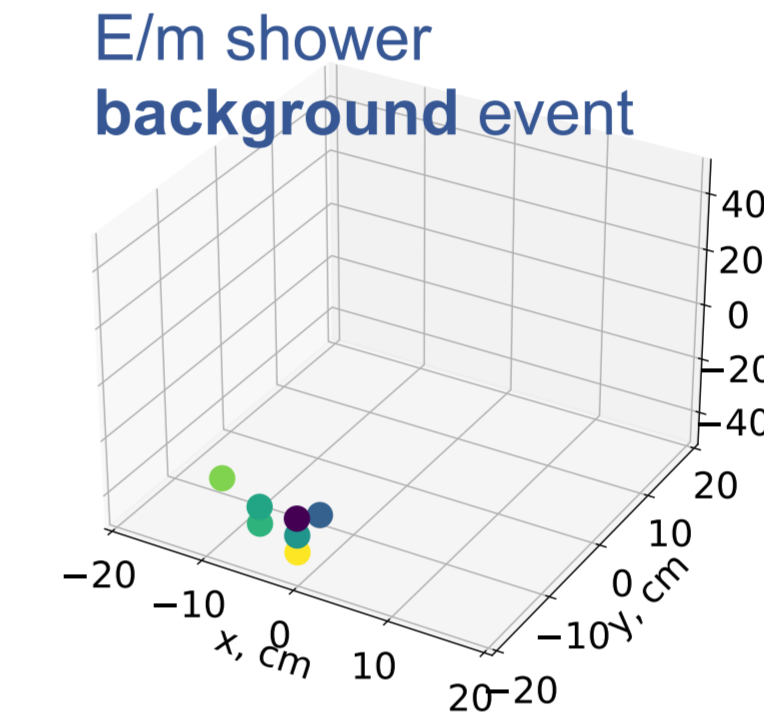
Conclusions

- Event structure-based **GNN compared with BDT** which performs well in various event classification problems in HEP
- BDT learns feature distributions
- GNN learns event structures
- Similar performance using target feature E_{ToF}**
- Excluding E_{ToF} variable increases significance of event topologies for events with $N_{hits} > 1 \Rightarrow$ slight increase of GNN performance compared to BDT**
- Possible limit of GNN performance:
 - Large fraction of single hit events and irregular event signatures for given dataset
 - \Rightarrow GNN can be more beneficial at higher energies and higher detector granularities

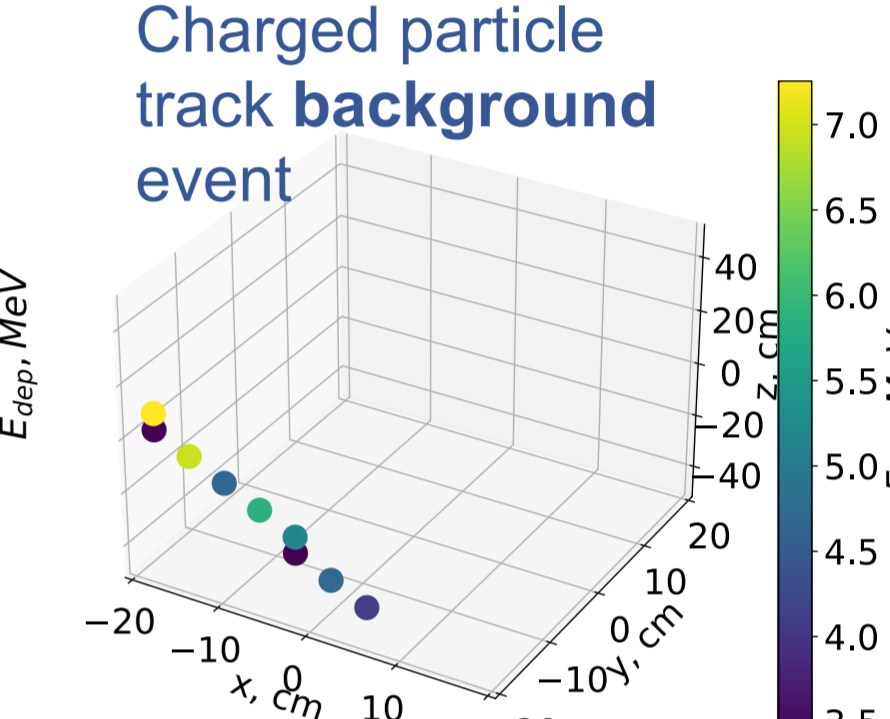
Imaging capabilities of the HGN detector

Event type signatures:

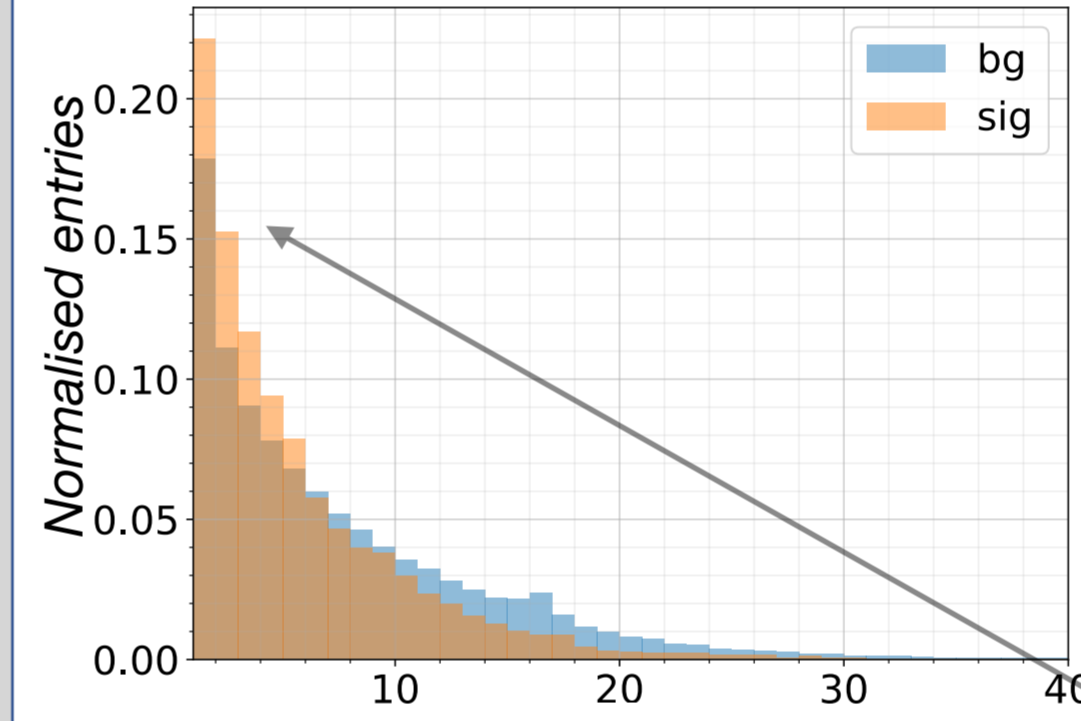
- tracks of charged particles
- compact **electromagnetic showers**
- sparse and irregular **hadronic showers**
 - no upstream track for neutral hadrons



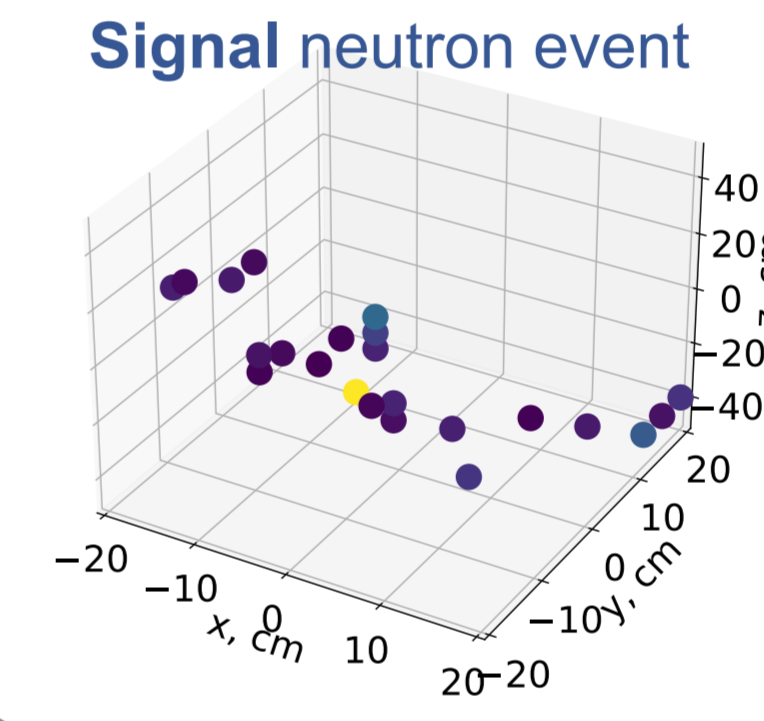
E/m shower background event



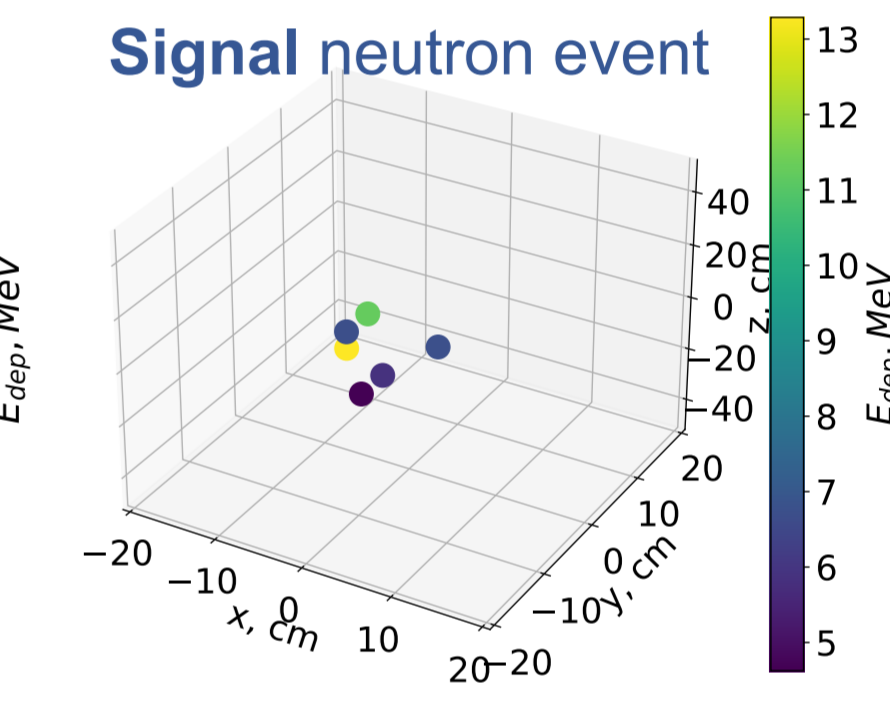
Charged particle track background event



Normalised entries vs N_{hits}



Signal neutron event



Signal neutron event

Significant fraction of single hit events

Classification models

Event structure model

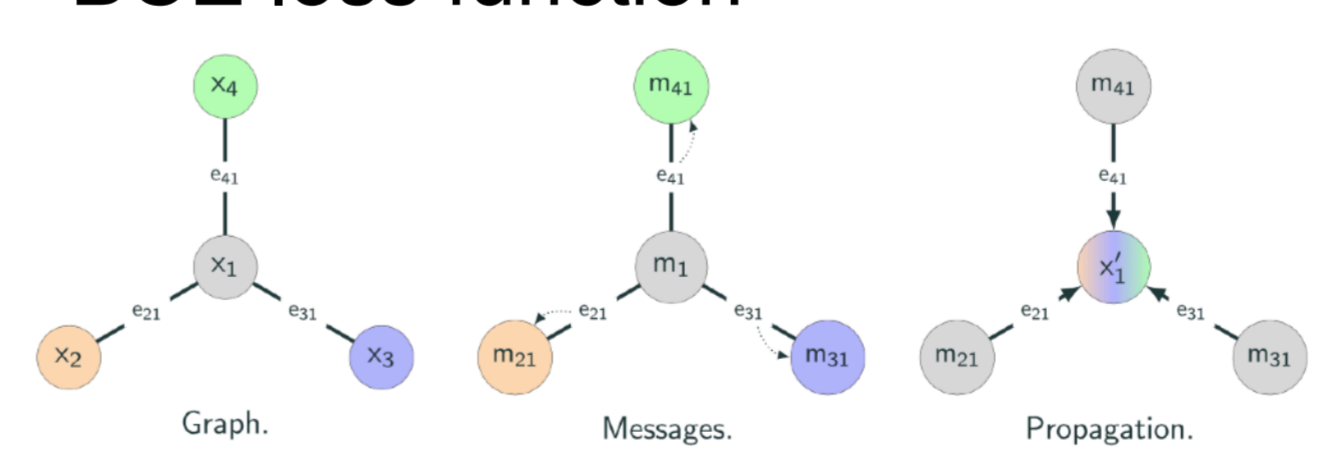
Graph neural network (GNN)

- $(x, y, z), E_{dep}, T_{hit}$ (after first hit) + E_{ToF} (optional)
- Fully connected hit graphs
- 2 GraphSage layers with 32 hidden channels \rightarrow Self-attention pooling layer \rightarrow MLP
- BCE loss function



VS.

First principle model

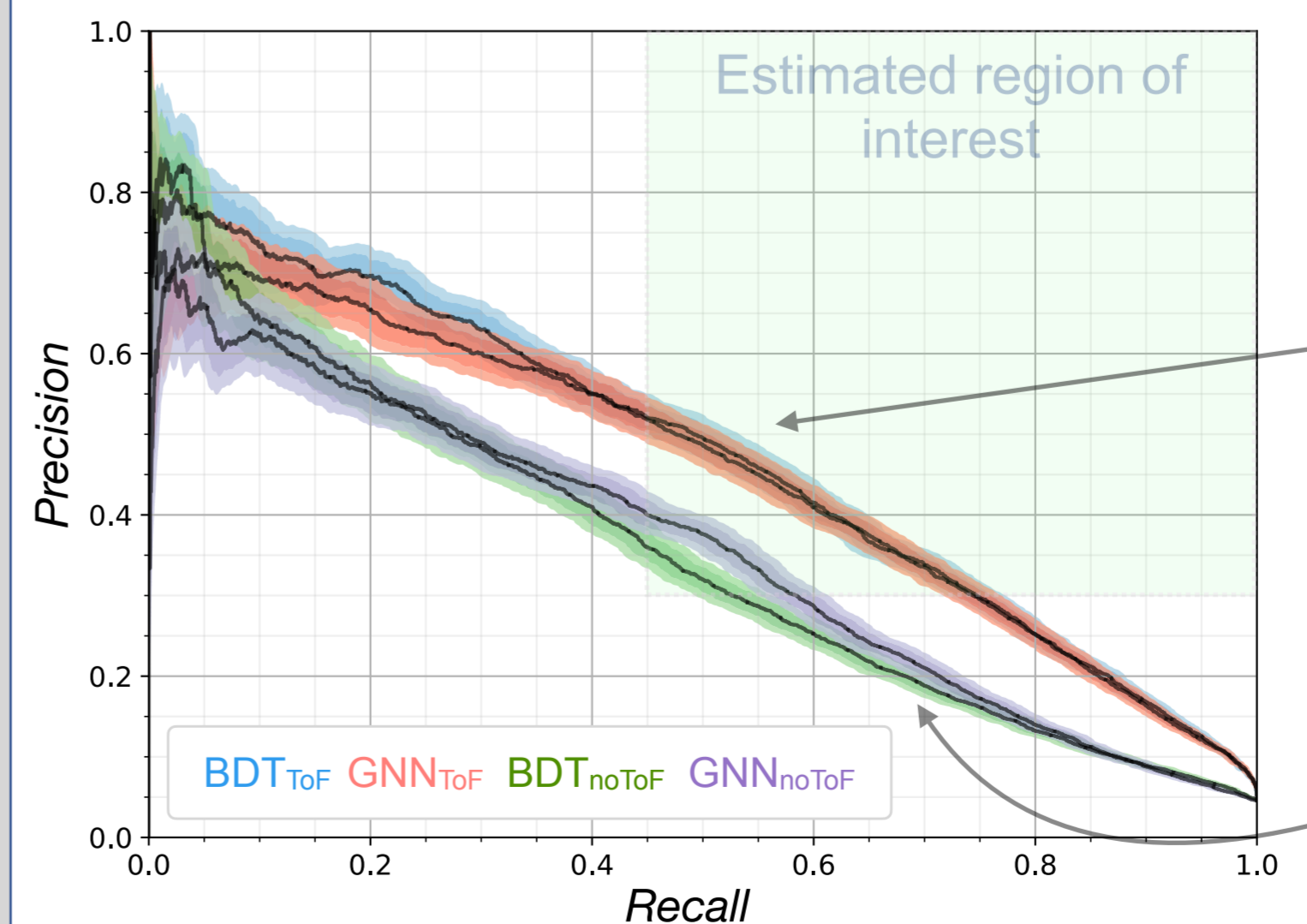
Boosted Decision Tree (BDT) model with **first-principle feature set** based on global event properties and parameters of most informative hits.



J. Gilmer et al., "Neural message passing for quantum chemistry," 2017.

Results



2 sets of GNN and BDT models:

- Using E_{ToF} feature for classification
 - Biased to the parameters of simulations
- No time-of-arrival information is used**
 - Less dependent on simulation

Region of interest:

- ~ Precision threshold - exclude flat neutron flow hypothesis
- ~ Recall threshold - covers most of neutron E_{kin} spectrum

$Precision = \frac{TP}{TP + FP}$

$Recall = \frac{TP}{TP + FN}$

[MMU package](#) for PR-uncertainties