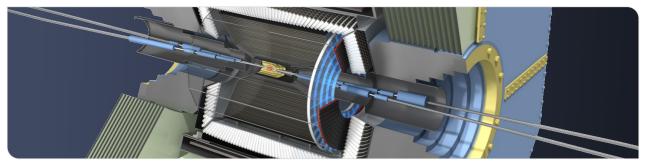




GNN-based Track and Vertex Finding at Belle II

26. Deutsche Physikerinnentagung: Physics Talks - Particle Physics Experiment

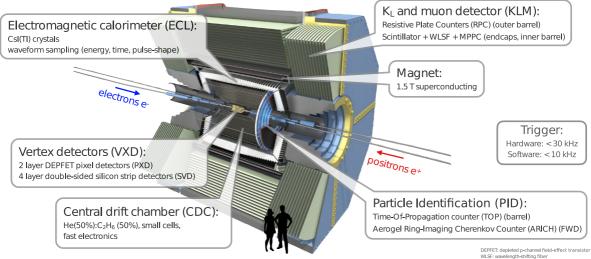
Lea Reuter - lea.reuter@kit.edu | 27th November 2022



www.kit.edu

Belle II Detector

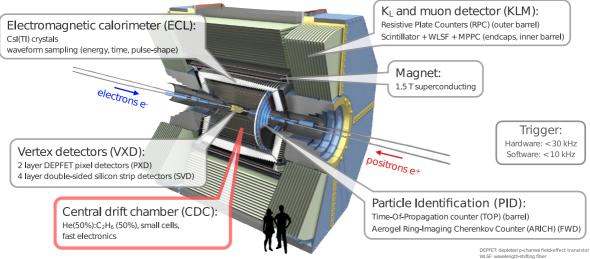




MPPC: multi-pixel photon counter

Belle II Main Tracking





WLSF: wavelength-shifting fiber MPPC: multi-pixel photon counter





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(a) An axial wire layer - sense wires are parallel to the beamline

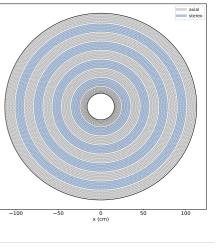


(b) A stereo wire layer - sense wires are skewed to the beamline (exaggerated)

3/11

- Main tracking detector of Belle II
- Sense wires are arranged parallel to the the beamline and the magnetic field (z-axis) to measure charged particles
- Information in z-direction is gathered from angled stereo layers

 \rightarrow Focus on track reconstruction using the CDC hits



CDC x-y view

100

50

-50

-100

0 (Ĵ

arXiv:2003.12466



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(b) A stereo wire layer - sense wires are skewed to the beamline (exaggerated)

3/11

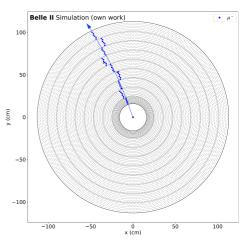
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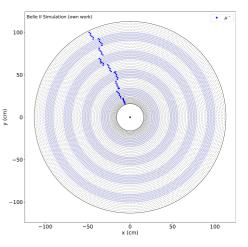
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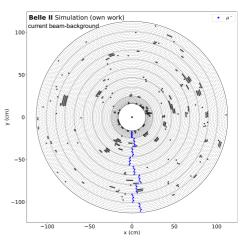


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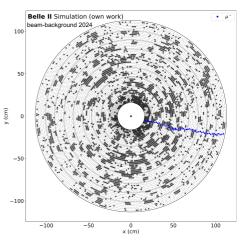


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3/11

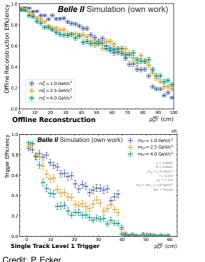
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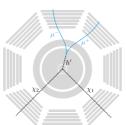
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arXiv:2003.12466

GNN-based Track and Vertex Finding







Motivation:

Displaced vertices important signature in searches for new physics a

Challenge:

- Real-time (< 2 μs) trigger and offline reconstruction ^b efficiency decreases depending on displacement (K⁰₀, Λ⁰, Dark Sector searches)
- Tracks with radial displacement larger than 40 cm are currently not saved by real-time single-track selection
- New track finding methods must be developed for expected high beam-background in 2024

\rightarrow Improve both real-time reconstruction and offline reconstruction and develop real-time vertex finder with Graph Neural Networks (GNN)

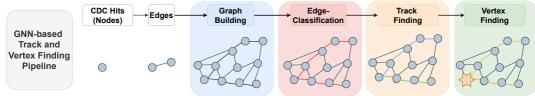
^bhttps://github.com/belle2/basf2

^aLong-lived Dark Higgs and Inelastic Dark Matter at Belle II (arXiv:2012.08595)

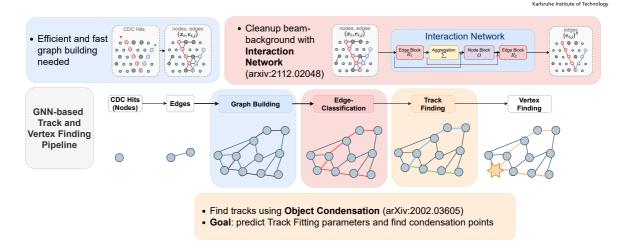
Approach with Graph Neural Networks



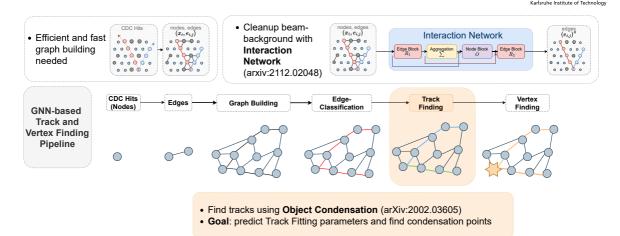
Variable number of CDC hits \rightarrow utilize Graphs and Graph Neural Networks (GNN)



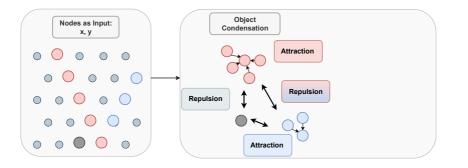
Approach with Graph Neural Networks



Approach with Graph Neural Networks: Track Finding



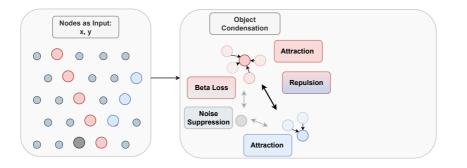




 \rightarrow Use nodes as input to Object Condensation (arXiv:2002.03605)

 \rightarrow Goal: Find unknown and variable number of Tracks and simultaneously predict track fitting parameters (momentum, displacement...)

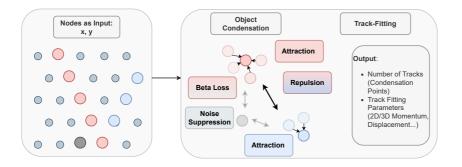




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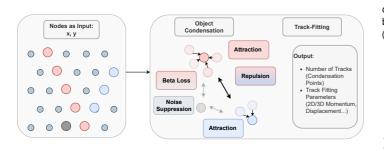




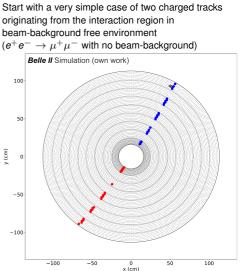
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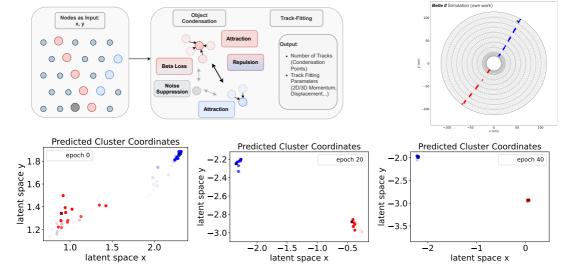




 \rightarrow Use nodes as input to Object Condensation (arXiv:2002.03605) \rightarrow **Goal:** Find **unknown** and **variable** number of Tracks and simultaneously predict track fitting parameters (momentum, displacement...)







Object Condensation Track Finding



More realistic scenario:

Simulated samples:

- Between 1 to 3 μ^- and μ^+
- Starting with current (moderate) beam background conditions
- Not displaced
- θ = [30, 120], so within CDC
- Transverse momentum p_t = [0.3, 5] GeV

Object Condensation Track Finding



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Training

- Trained on 12 000 samples
- Input features per node: Only use 2D-information CDC hit x, y position
- Predicting: unknown and variable number of tracks *n*tracks and 3D-Momentum *px*, *py*, *pz*

Object Condensation Track Finding

More realistic scenario:

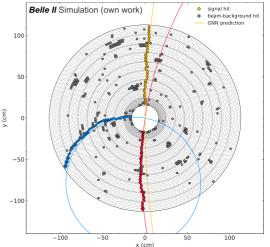
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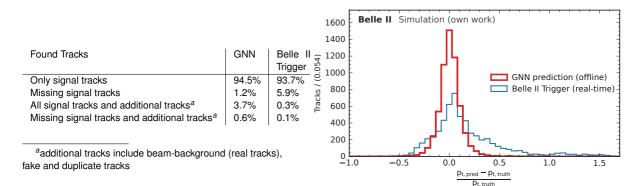
Assumption: Tracks starting at the Interaction Point, currently only momentum is predicted





Track Finding: Comparison with Online L1 Trigger

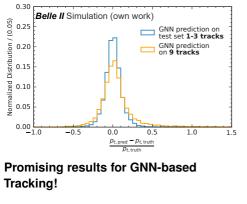


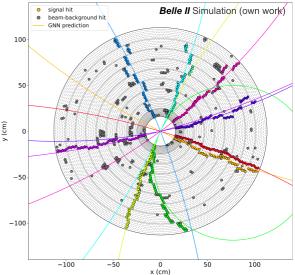


Track Finding: High Multiplicity Events



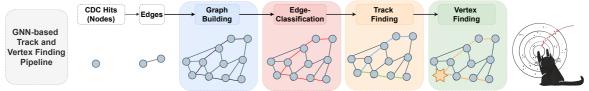
Object Condensation model is good in **generalizing**: Evaluation of previously trained model on high multiplicity events (9 particles)





Summary and Outlook





Current Status

- Developing GNN-based Track and Vertex Finding Pipeline
- Performance measurement of different graph building methods and edge-classification
- Implemented Object Condensation for Drift Chamber Track Finding and testing on samples with current beam-background conditions

\rightarrow Have first full Track Finding and Fitting GNN model working

Outlook

- Working together with ITIV (Department of Electrical Engineering and Information Technology at KIT) to implement Graph Building, Edge Classification Network and Object Condensation Model on FPGA for real-time application
- Extend Object Condensation to predict displaced tracks offline and online and develop Vertex Finding
- Evaluate Object Condensation on Data/MC for displaced vertex Example

$$^+e^- o \Phi \gamma, \ \Phi o {K_{
m S^0}} {K_{
m L^0}}$$