



Muon identification with Deep Neural Network in the Belle II K-Long and Muon detector

Based on [arXiv:2503.11351](https://arxiv.org/abs/2503.11351)
Accepted by NIM-A

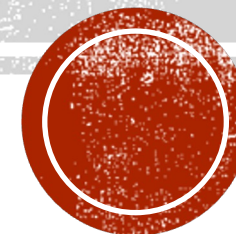
WANG Zihan 王子涵

The University of Tokyo

On behalf of authors

EPS-HEP 2025

2025.07.09



SuperKEKB and Belle II Experiment

SuperKEKB

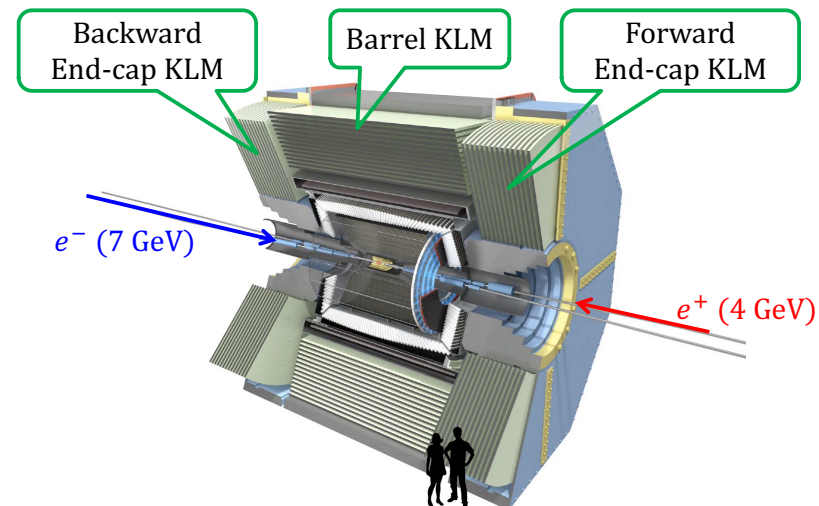
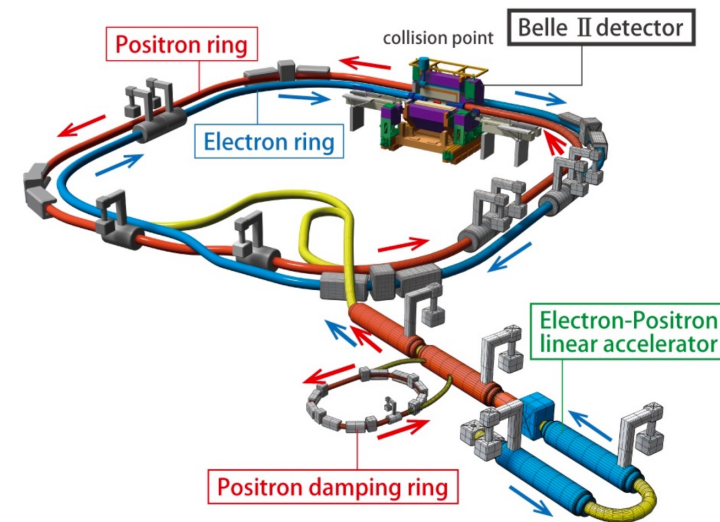
- Asymmetric e^+e^- collision at $\Upsilon(4S)$ resonance

	$\int \mathcal{L} dt$	\mathcal{L}
Target	50 ab ⁻¹	$6 \times 10^{35} \text{ cm}^{-2} \text{ s}^{-1}$
Achieved	575 fb ⁻¹	$5.1 \times 10^{34} \text{ cm}^{-2} \text{ s}^{-1}$

New world record!

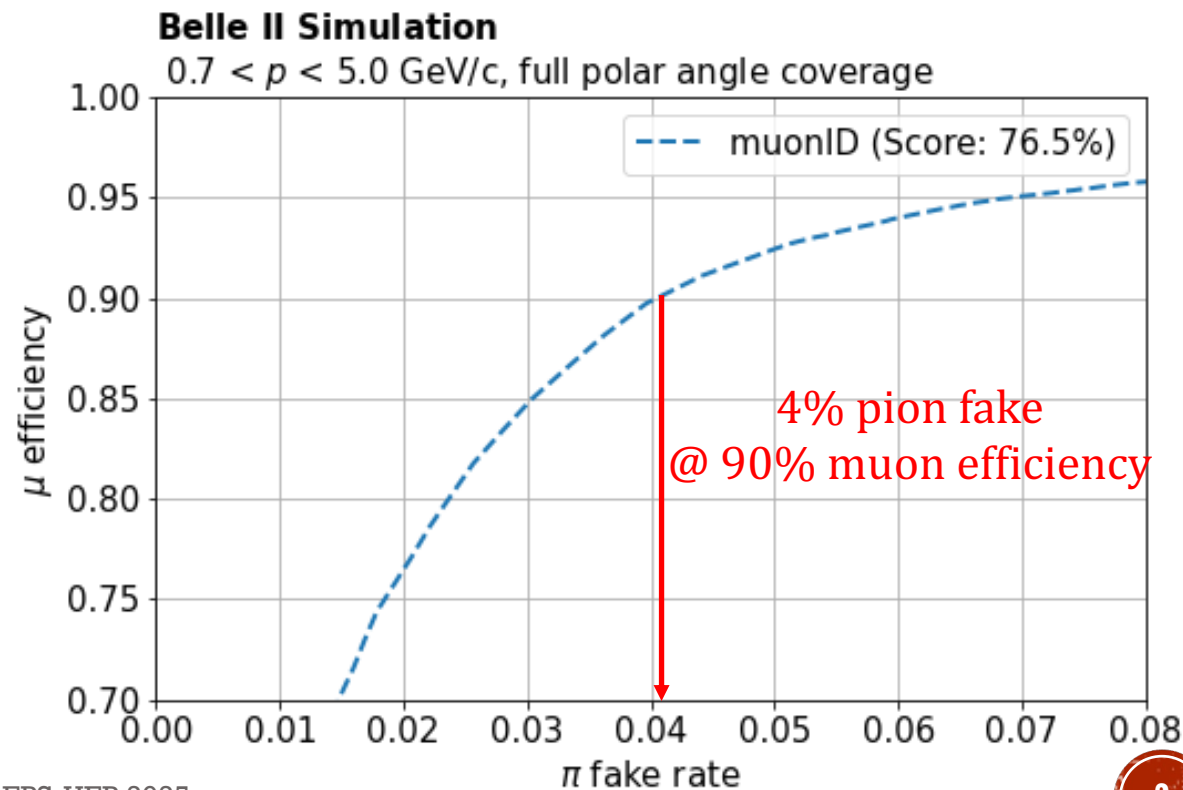
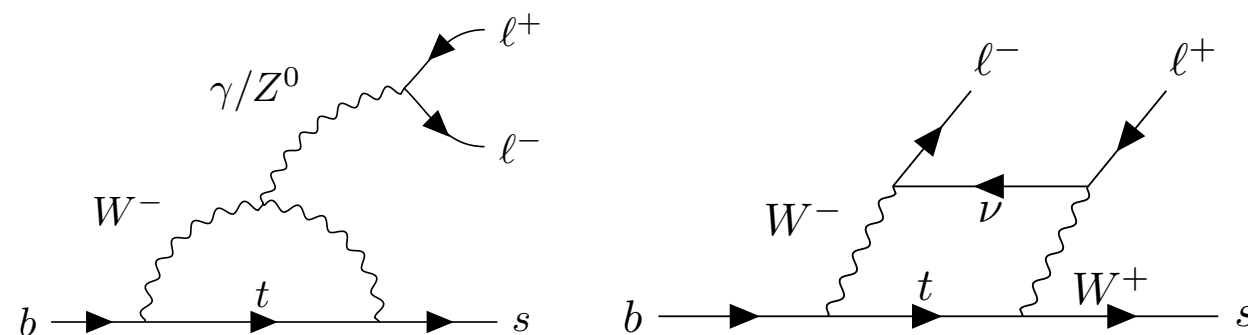
Belle II experiment

- New physics search at the luminosity frontier
- Rich muon-related physics topics like $b \rightarrow s\mu^+\mu^-$ transition, $b \rightarrow c\tau\nu$ with $\tau \rightarrow \mu\nu\bar{\nu}$
- Muon detection by K-Long and Muon (KLM) detector at the outer most of Belle II

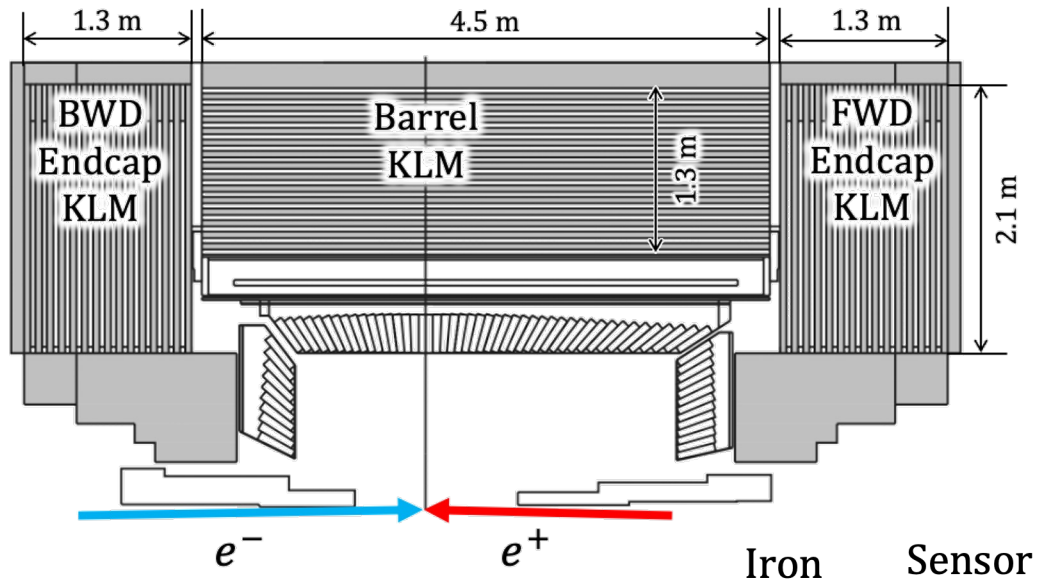


Physics motivation

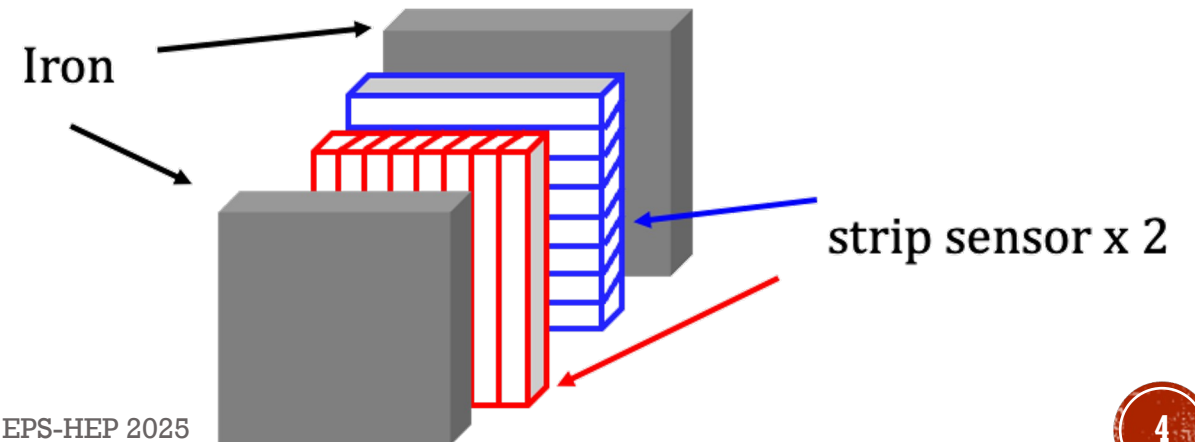
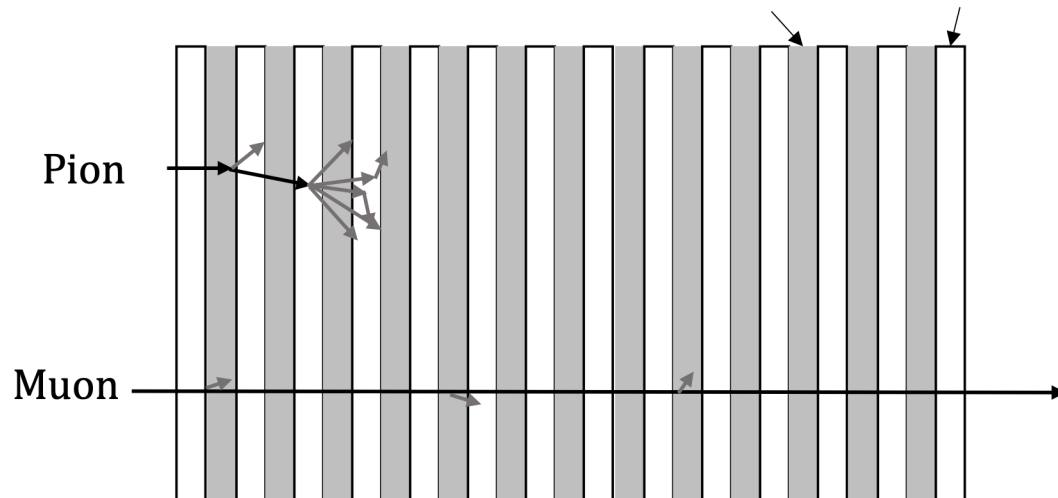
- Electro-weak penguin process $b \rightarrow s\mu^+\mu^-$ is suppressed in SM, and sensitive to new physics
- $\pi^\pm \rightarrow \mu^\pm$ mis-identification from $B \rightarrow X\pi^+\pi^-$ decay is one of the main background
 - $\mathcal{B}(B \rightarrow X\pi^+\pi^-)/\mathcal{B}(B \rightarrow X_s\mu^+\mu^-) = \mathcal{O}(1000)$
- π fake $< 2\%$ is required
- Target: Suppress pion fake rate below 2% @ 90% muon efficiency



K-Long and Muon Detector (KLM)



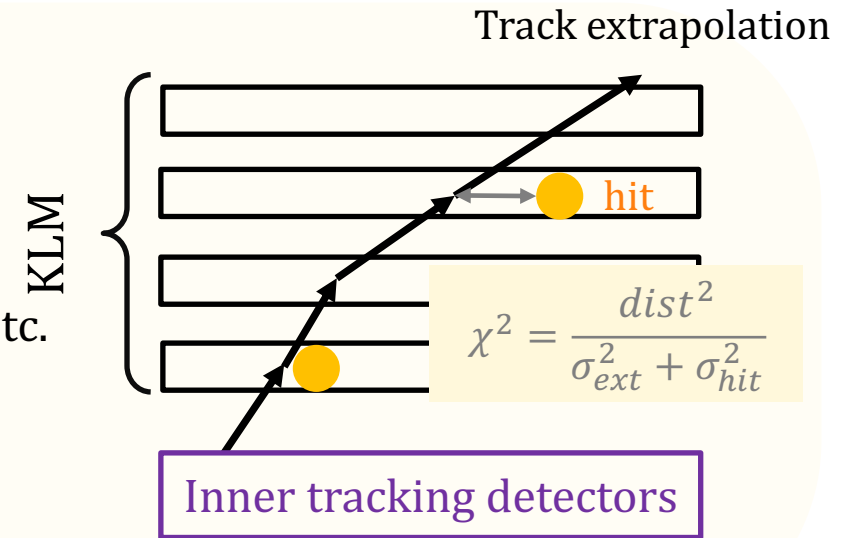
- Sandwich structure of iron plate and sensor
 - Barrel: 15 layers, Endcap: 14/12 layers for FWD/BWD
 - Sensor: Resistive Plate Chamber or Plastic scintillator
- Muon identification based on penetration ability
 - > 3.9 interaction length
- Detect 2-dimentional spatial information with 2 orthogonally placed strip sensors in each layer



Current muonID algorithm

STEP 1 Track extrapolation & Hit association

- Measure track position/momentum with inner detectors
- Use Geant4E to estimate penetration depth inside KLM
 - Assuming tracks are muon, considering multiple-scattering effects etc.
- Associate one hit with smallest χ^2 to target track at each layer
 - Also require $\chi^2 < (3.5)^2$
- Use Kalman-Filter to correct extrapolation



STEP 2

muonID calculation

- Use penetration ability and extrapolation quality to calculate muon/pion likelihood ($\mathcal{L}_\mu/\mathcal{L}_\pi$)
- muonID definition:

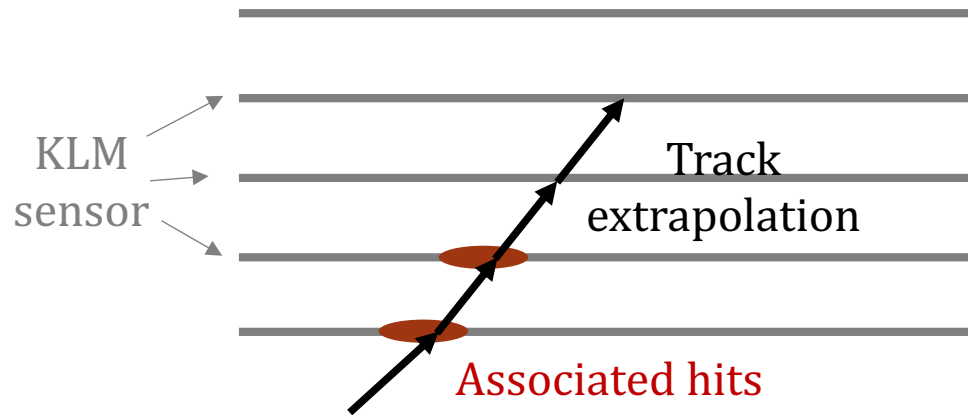
$$\text{muonID}_{\text{KLM}} = \frac{\mathcal{L}_\mu}{\mathcal{L}_\mu + \mathcal{L}_\pi}$$

Likelihood extraction on penetration ability

- Assign a probability to each layer depending on whether a hit is associated or not
- Calculate probability by direct product of layer probability

$$\mathcal{L}_t^{pene} = \prod_{n=1}^{N_{ext}} L_{t,n} \quad t \in \{\mu, \pi\}$$

- For tracks extrapolated to stop at layer 4 ($N_{ext} = 4$) and only hits associated at the first two layers ($N_{hit} = 2$):



$$\begin{aligned} L_{t,4} &= 1 - P_{t,4} \\ L_{t,3} &= 1 - P_{t,3} \\ L_{t,2} &= P_{t,2} \\ L_{t,1} &= P_{t,1} \end{aligned}$$

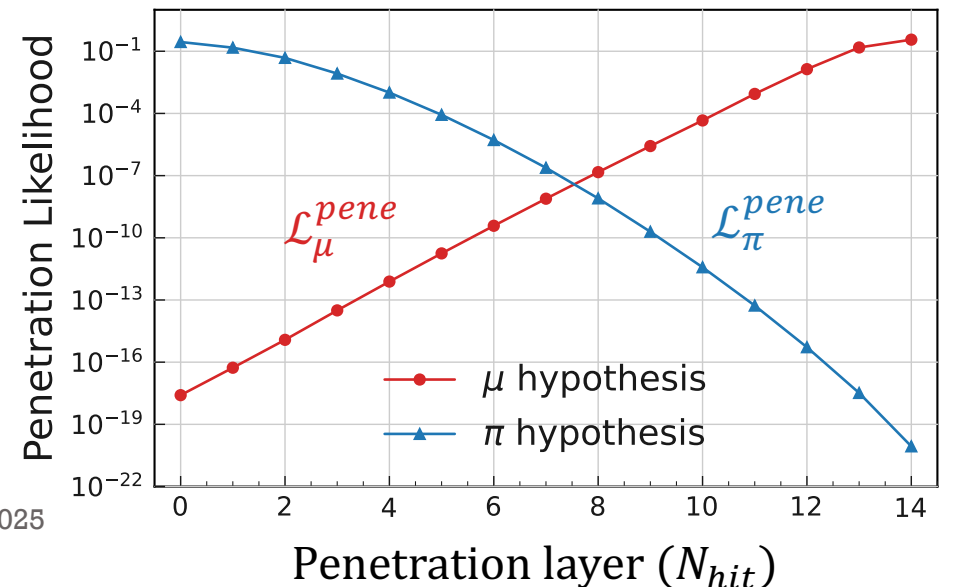
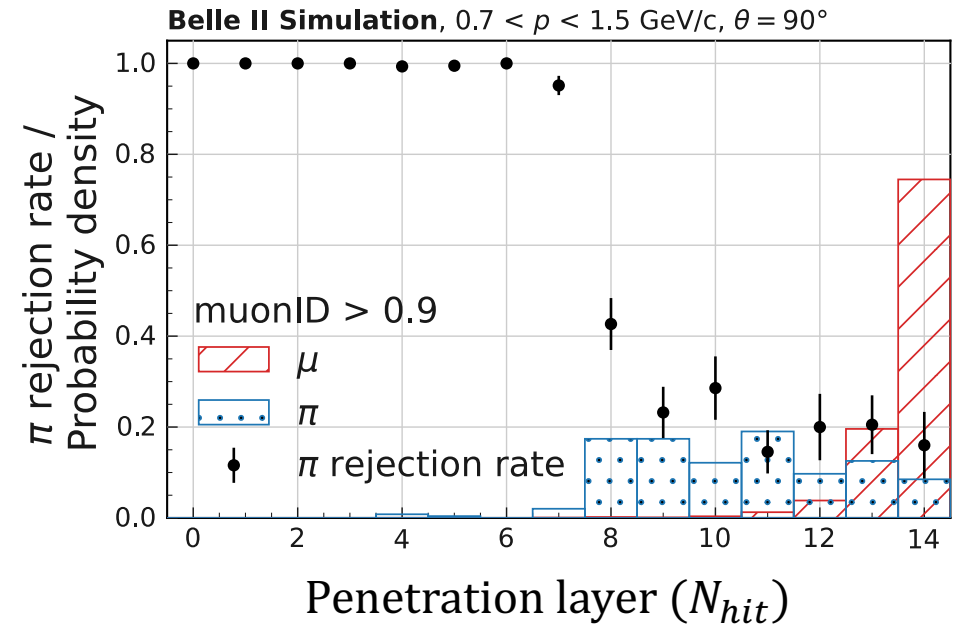


$$\begin{aligned} \mathcal{L}_t^{pene} &= \prod_{n=1}^{N_{ext}=4} L_{t,n} \\ &= P_{t,1} P_{t,2} (1 - P_{t,3})(1 - P_{t,4}) \end{aligned}$$

Discussion on penetration likelihood

- For tracks extrapolated to layer 14 ($N_{ext} = 14$), muonID only rejects pions with penetration (stopping) layer $N_{hit} < 8$
- Still, quite a fraction of remaining pions exists in $8 \leq N_{hit} \leq 10$, which rarely happens for muon
- MuonID failed to reject pions in $8 \leq N_{hit} \leq 10$ due to imperfect modeling of penetration likelihood
 - \mathcal{L}_{μ}^{pene} larger than \mathcal{L}_{π}^{pene} from $N_{hit} > 8$
 - Because **correlations between layers** are not properly considered

Better modeling of penetration information can improve performance



Strategy

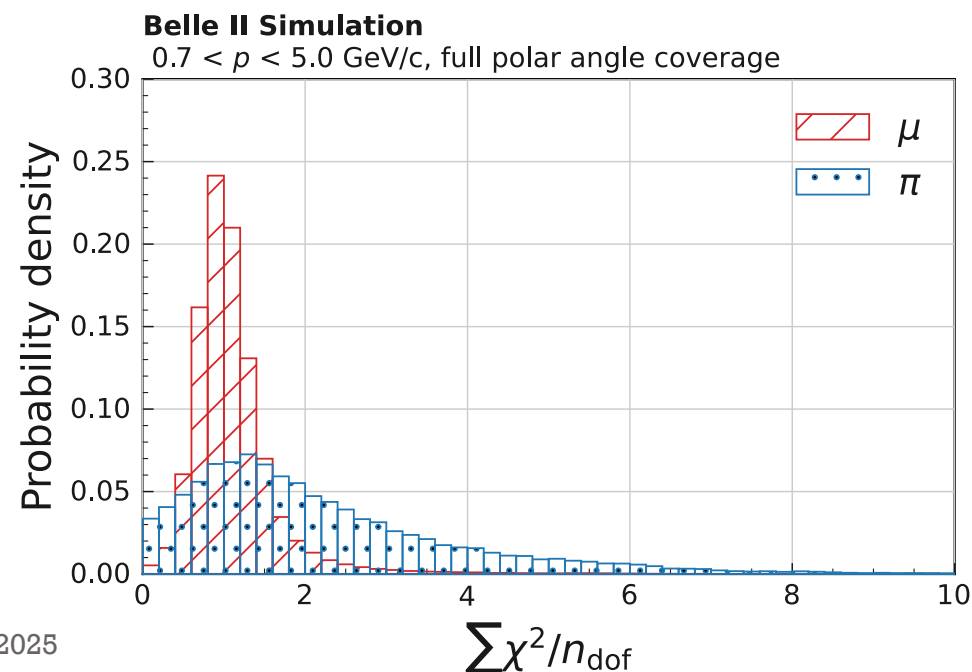
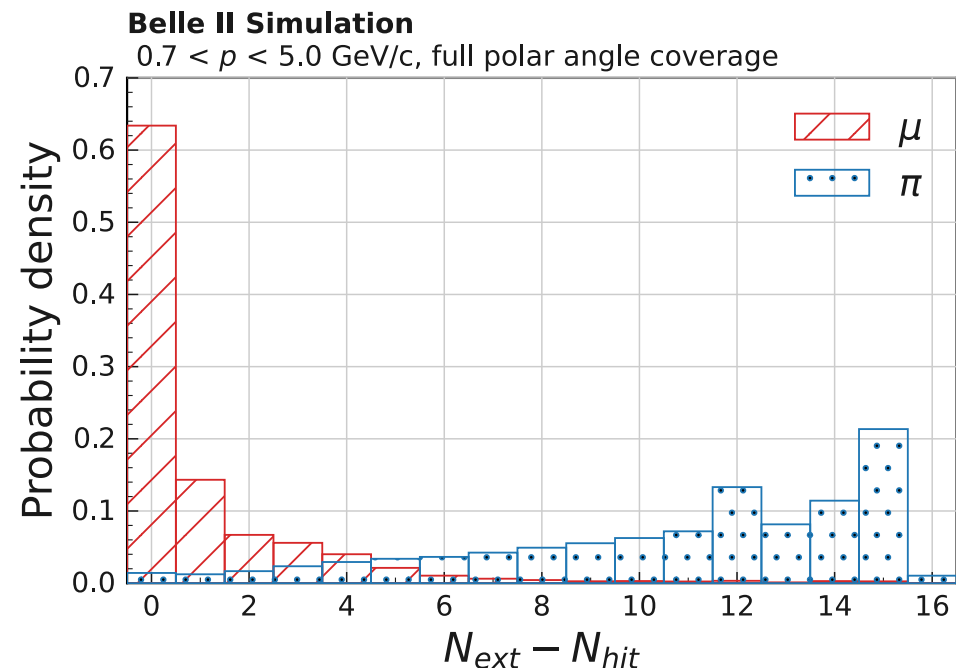
- New Deep Neural Network (DNN)-based algorithm
- Consider correlations between input variables

Input variables

1. Global variables
 - Track-level variables like penetration layers
2. Hit pattern variables
 - Detailed hit-level information
 - New for muon identification

Global variables

- Difference in extrapolation & penetration layer:
 $N_{ext} - N_{hit}$
- Extrapolation layer
 - Penetration layer distribution is different for different extrapolation layer
- Extrapolation quality ($\sum\chi^2$ and n_{dof})
 - Muons extrapolated with muon hypothesis and thus peak at 1
 - Pion has larger value due to multiple scattering
- Transverse momentum

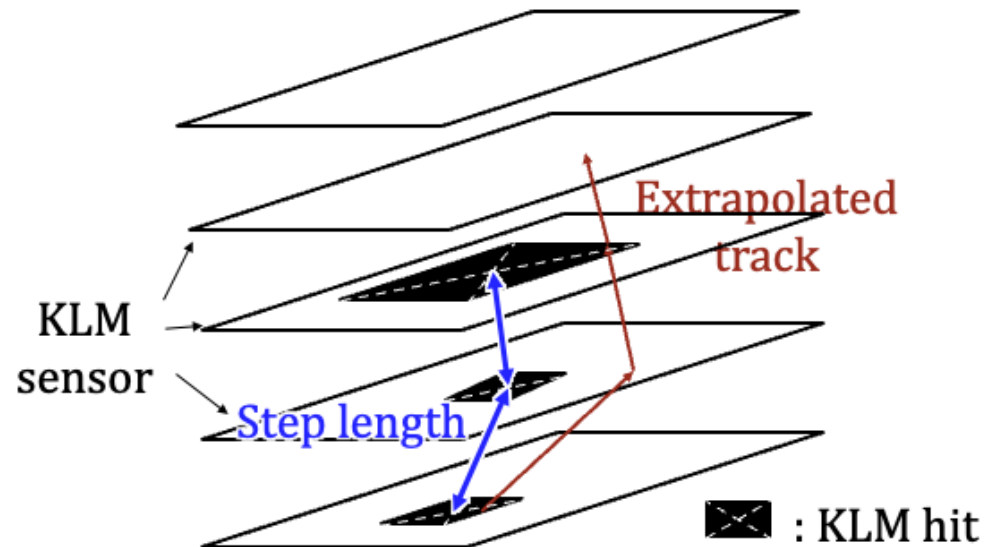


Hit pattern variable: Step length

- **Step length**: distance to hit in previous layer

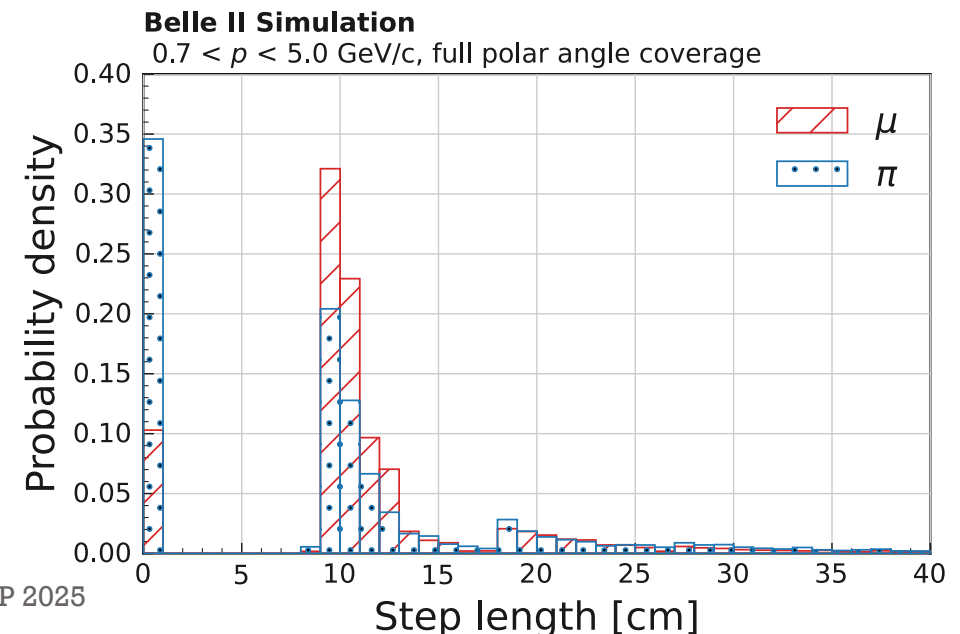
Strength:

- Detailed 3D penetration information compared to penetration layer, which only reflects penetration projected in normal direction of detector plane
- Penetration depth (sum of step length) of muon is larger than pion



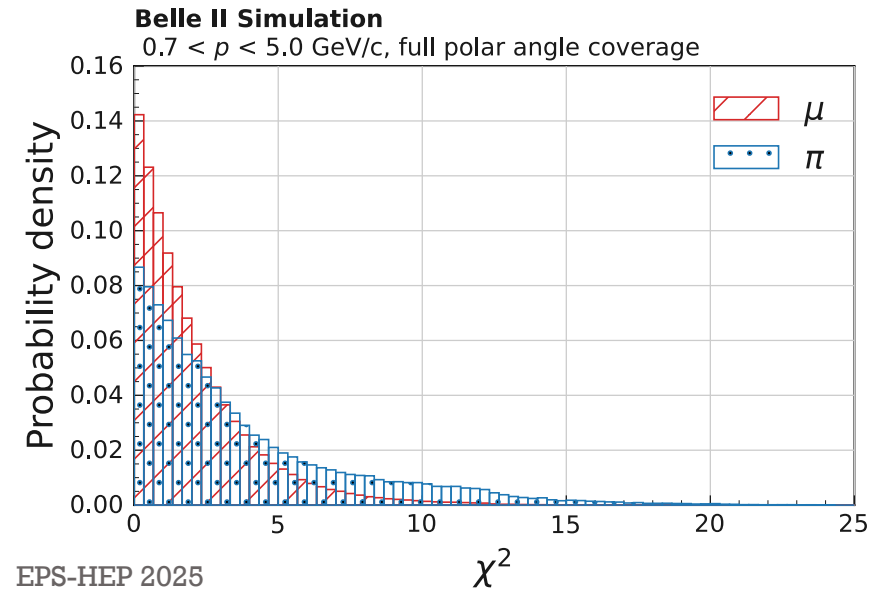
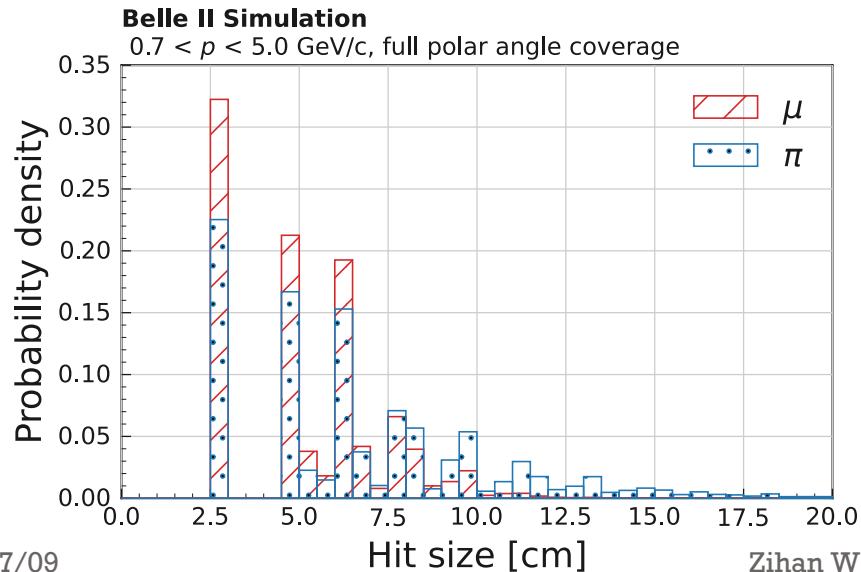
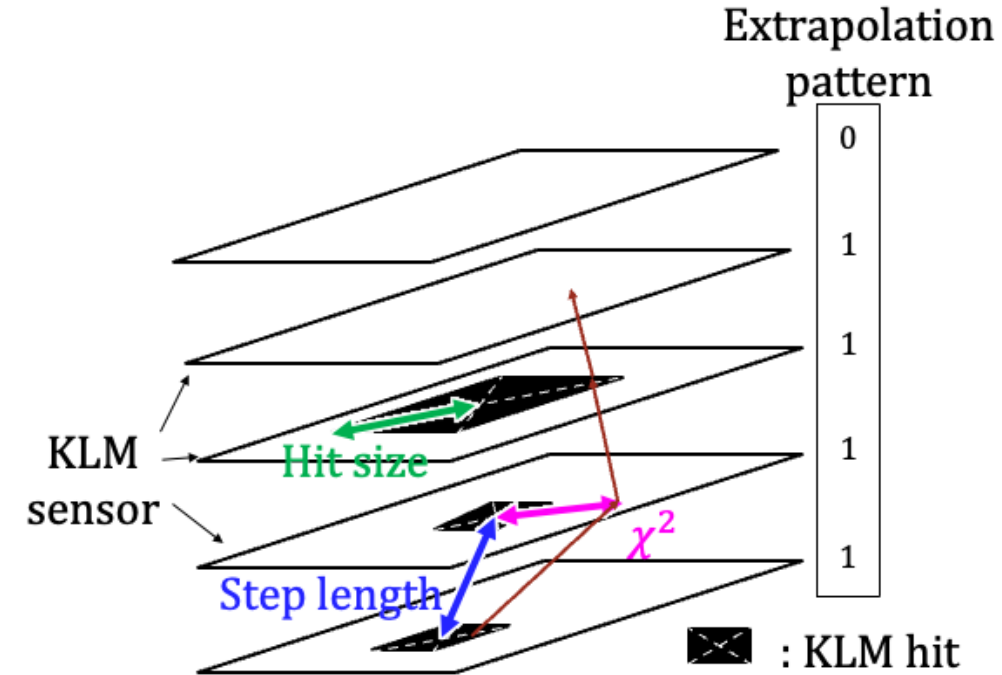
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Other hit pattern variables

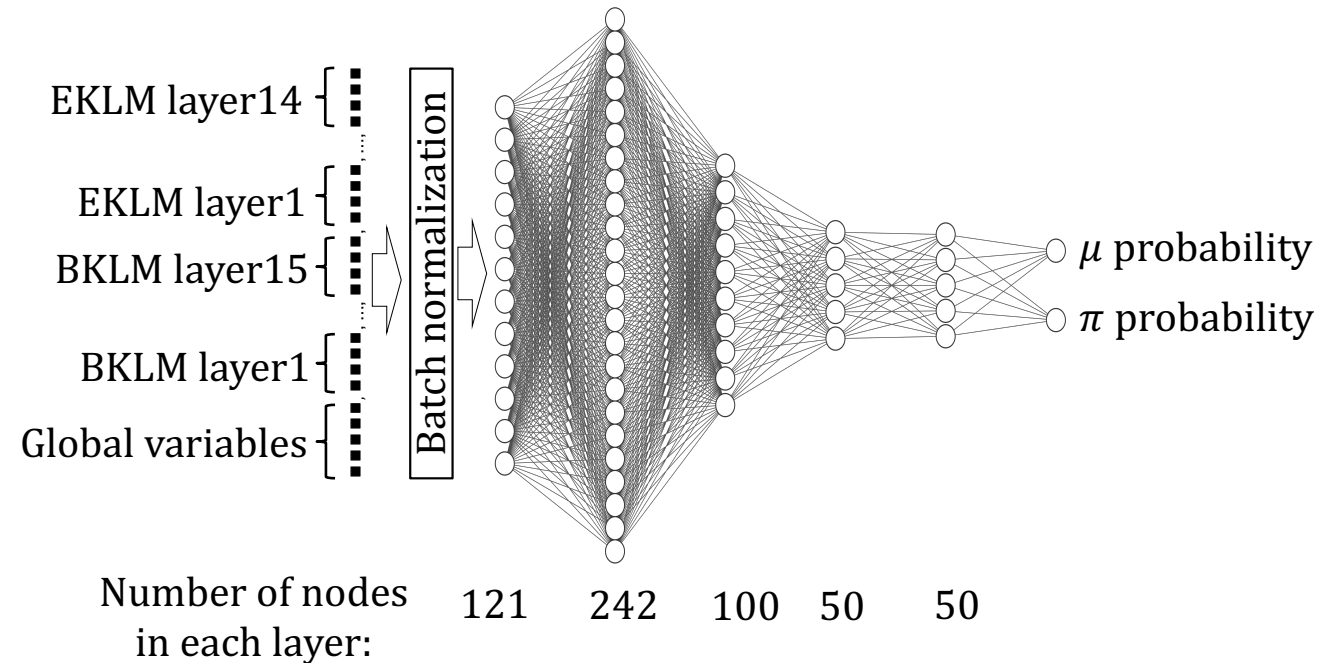
1. **Hit size**: Half of the diagonal length of the rectangular shape of a hit
2. χ^2 : Deviation from hit to extrapolation position
3. **Extrapolation pattern**: whether extrapolation has reached this layer
 - To compare extrapolation and penetration pattern



Network structure & training

Network structure

- 121 input variables
 - 5 global variables
 - 4 hit pattern variables per layer, 29 layers in total
- Fully connected Network
 - 64k trainable parameters
- LeakyReLU for activation function
- Softmax for probability output



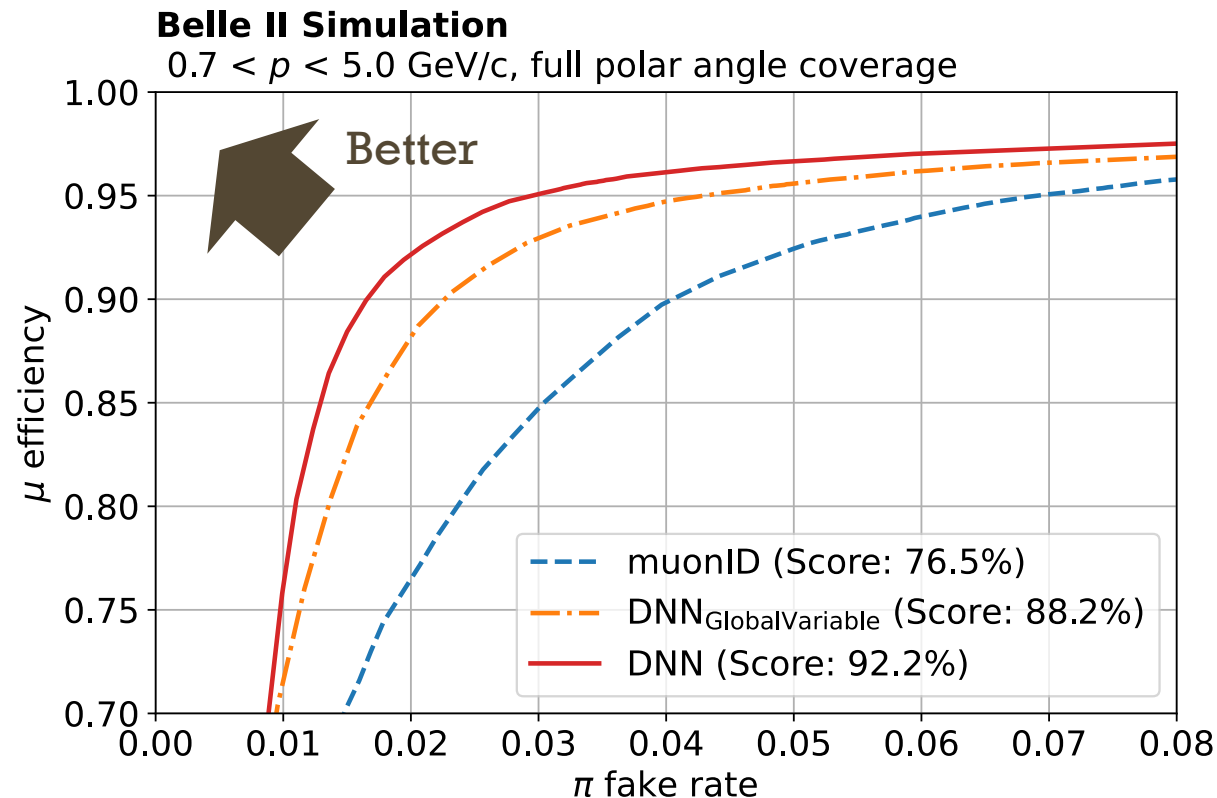
Training

- Event generated with 4 to 16 tracks for simulating different event multiplicity
 - Simulated beam-induced background overlayed
- Uniform phase space distribution covers full KLM geometric acceptance; 0.5M pions and muons
- No overtraining

Performance

- Also trained DNN with global variables only, showing the importance of hit pattern variables
 - Almost same network structure

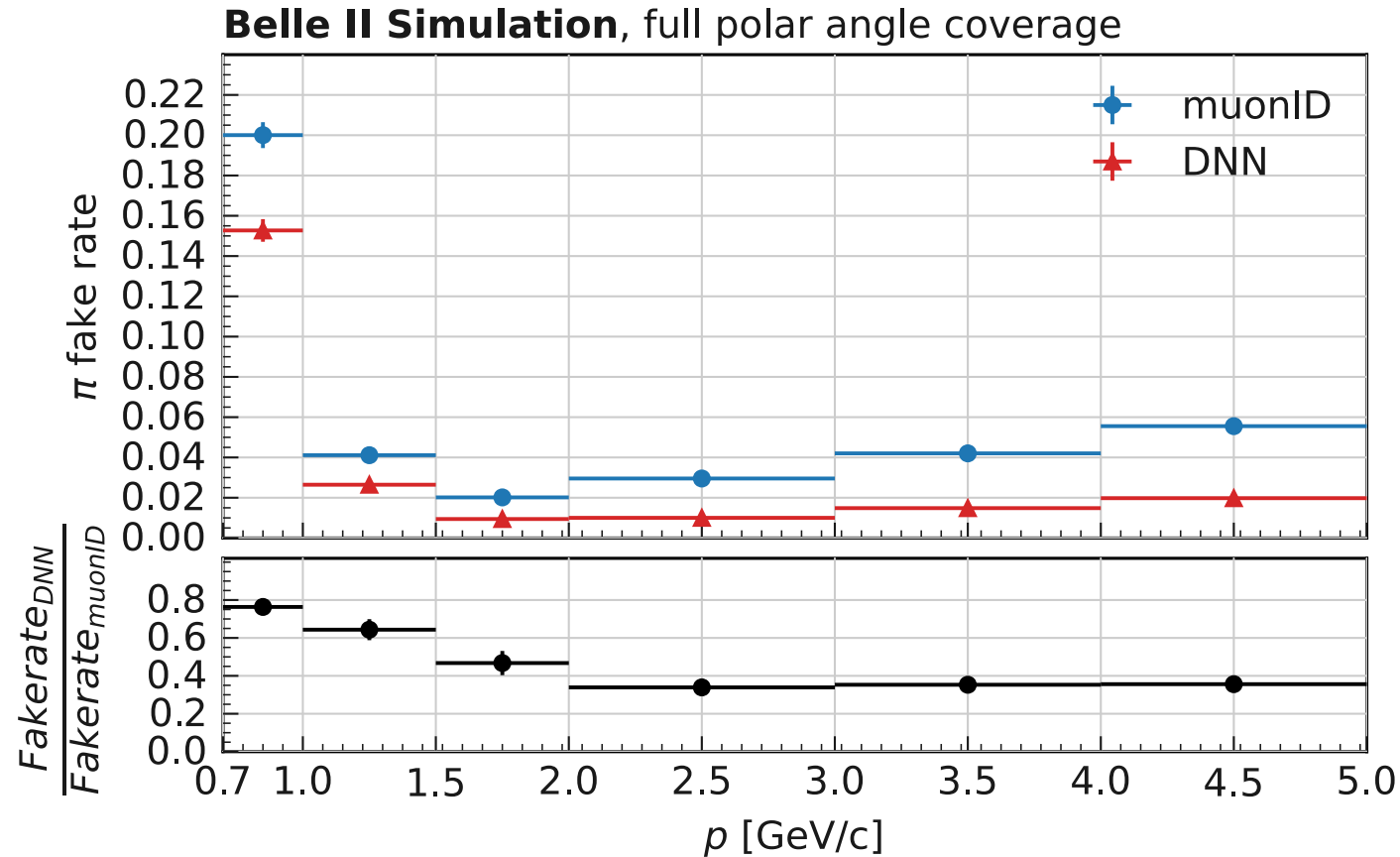
	Fake rate @90%muon effi	Muon effi @ 2% fake rate
muonID	4.1%	76.5%
DNN _{GlobalVariable}	2.3%	88.2%
DNN	1.6%	92.4%



DNN with hit pattern variables

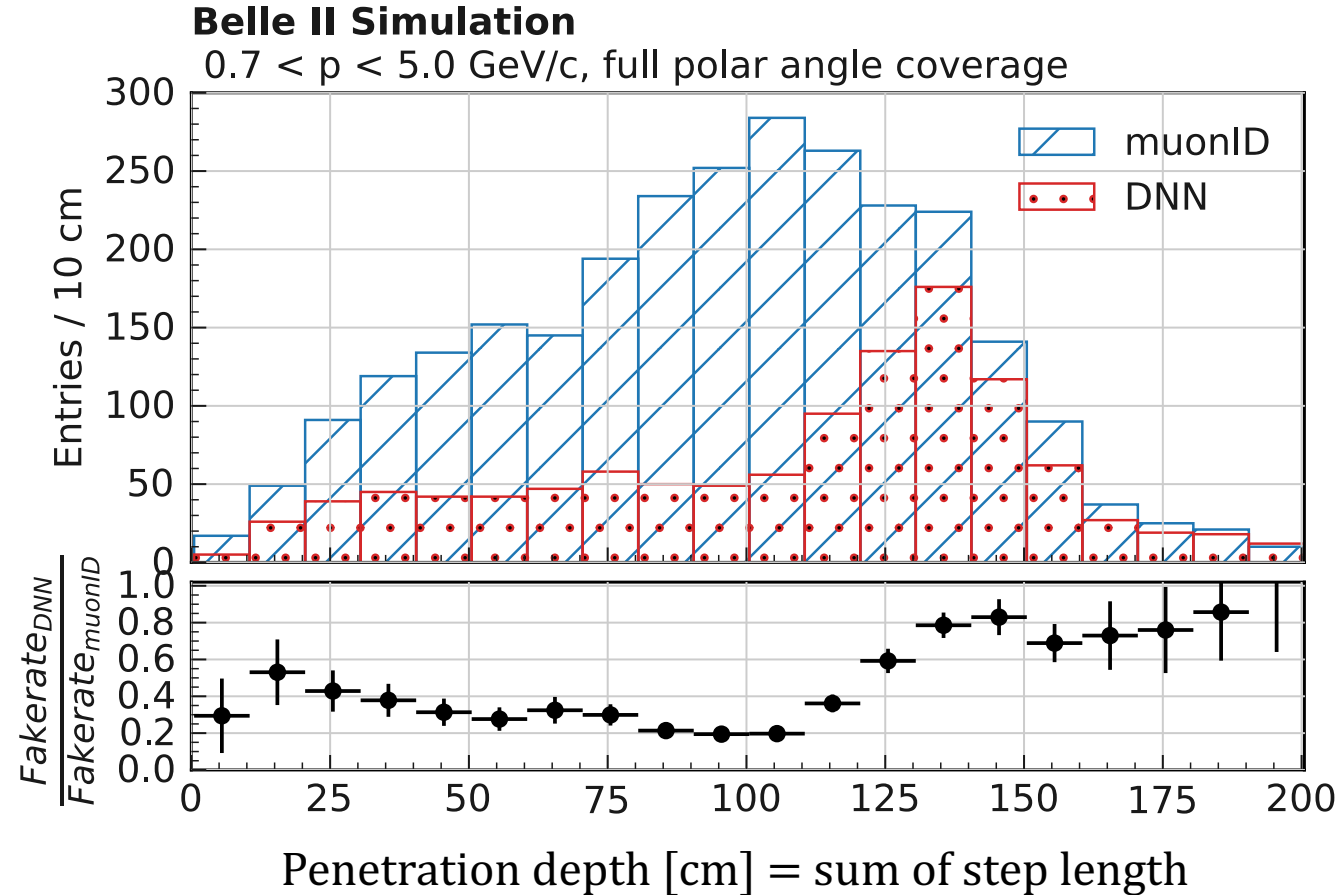
suppresses pion fake rate below 2% @ 90% muon efficiency

Performance



- Fix muon efficiency to 90% at each momentum bin.
- Pion fake rate improved over 60% at high momentum range ($p > 2.0 \text{ GeV}/c$).
- Improvement less significant at low momentum range where muons cannot traverse KLM entirely.

Performance



- Above plot shows penetration depth distributions of pions faking muons at 90% muon efficiency
- DNN rejected about 60% of deeply penetrated pions up to a depth of 125 cm
 - Rise of fake rate at 125 cm penetration depth similar KLM detector thickness of 130 cm
 - For tracks escaping from KLM, μ/π separation by penetration ability is impossible

Summary

- Muon identification algorithm with high performance is necessary for precise measurement of flavor physics at Belle II
- Current muonID cannot reject deeply penetrating pion
- We developed a new algorithm using DNN, employing additional hit-pattern variables
- We learned that this new algorithm suppresses pion fake rate below 2% at 90% muon efficiency in a simulation sample
 - Satisfied the requirement in $b \rightarrow s\mu^+\mu^-$ measurement
- This algorithm is implemented into Belle II software framework and validations on full dataset is expected in the future
- Check [arXiv:2503.11351](https://arxiv.org/abs/2503.11351) for more details!

BACK UP

Likelihood calculations for muonID

$$\mathcal{L}_\mu = \mathcal{L}_\mu^{pene} \times \mathcal{L}_\mu^{in-plane}$$

- Extract the probability of track t leaving hits in layer n , $P_{t,n}$ from simulation, and assign probability to each layer:

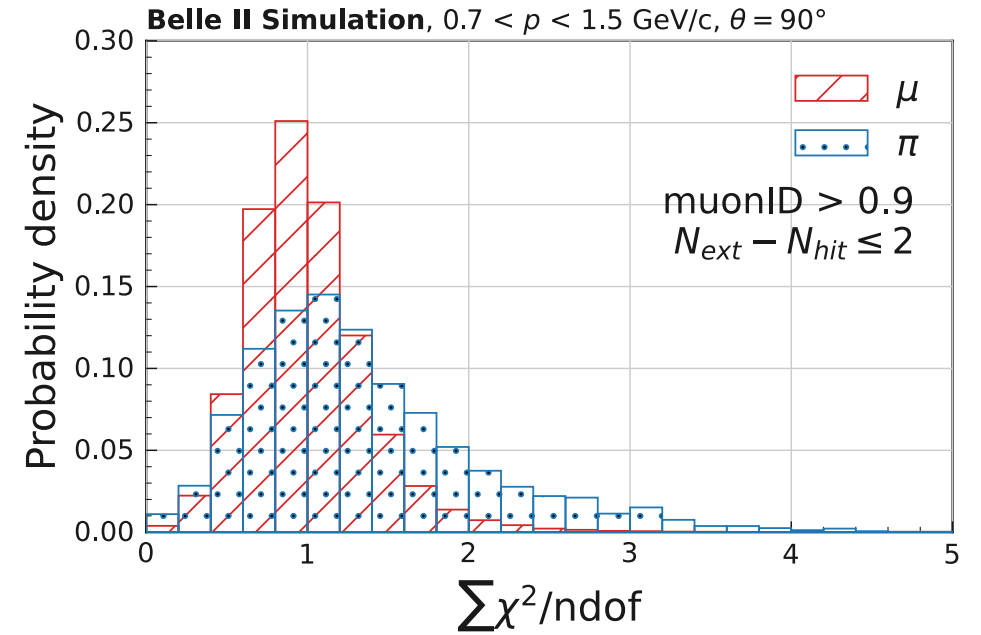
$$L_{t,n} = \begin{cases} P_{t,n} \cdot \varepsilon_n, & \text{with hit} \\ 1 - P_{t,n} \cdot \varepsilon_n, & \text{without hit} \end{cases}, \quad t \in \{\mu, \pi\}$$

- Typically, a 96% hit detection efficiency per layer (ε_n) is achieved in operations
- Calculate probability by direct product of layer probability

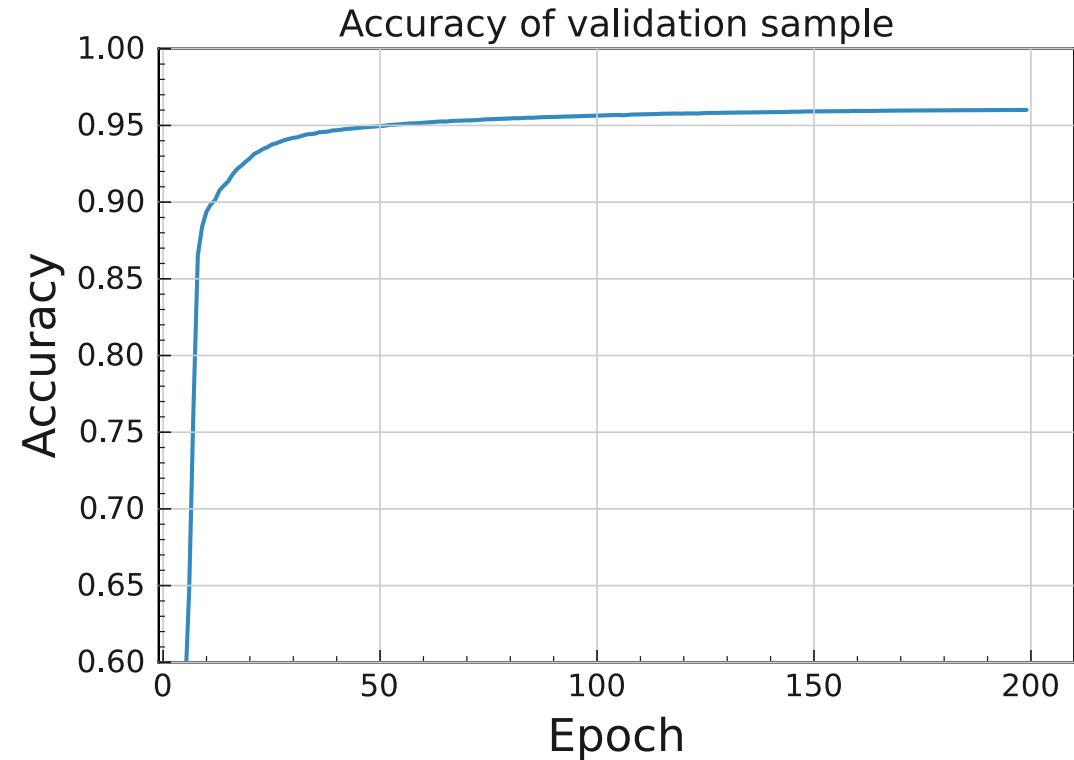
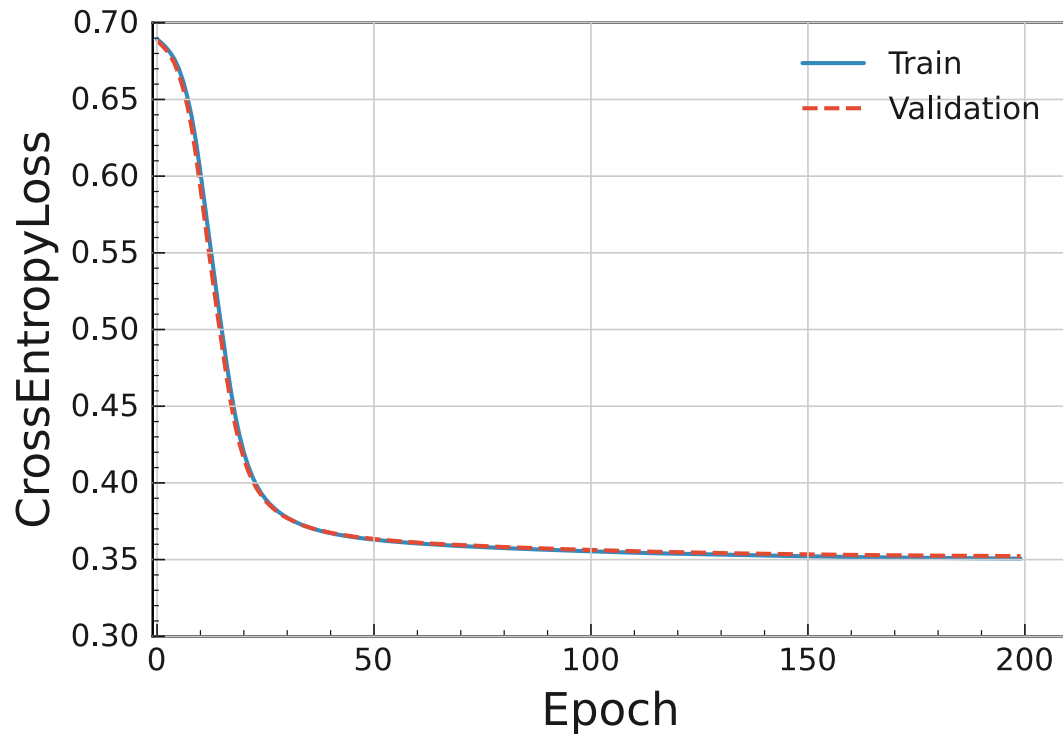
$$\mathcal{L}_t^{pene} = \prod_{n=1}^{N_{ext}} L_{t,n}$$

In-plane information

- In-plane likelihood derived from $\sum\chi^2/n_{dof}$ distribution
 - Extrapolated with muon hypothesis and thus peak at 1
 - Pion has larger value due to multiple scattering
- Still, some remaining pions can be rejected by requiring $\sum\chi^2/n_{dof} < 2$
- Failed to do so because of different scale of \mathcal{L}^{pene} and $\mathcal{L}^{in-plane}$.
- Specifically, $\mathcal{L}_\mu^{in-plane} / \mathcal{L}_\pi^{in-plane} \approx 10^{-1}$
- , $\mathcal{L}_\mu^{pene} / \mathcal{L}_\pi^{pene} \approx 10^{13}$
- The relationship $\text{muonID} > 0.9$ can hardly be influenced by the in-plane likelihood.



Training details



- Adam optimizer; learning rate of 10^{-5} ; batch size = 10000; CrossEntropy loss
- Early stop when accuracy does not increase for 10 epochs
- Device: RTX 3090 for 30 min training.

Robustness test and implementation

Robustness test

- Background:
 - Sample overlayed with beam background simulated at Belle II nominal luminosity (x3 of training sample)
 - Pion fake rate degrades to 2.4%, probably because of changes in hit pattern
 - Still DNN performs better than muonID, and re-training can be expected when machine status evolves
- Detector inefficiency
 - Set hit detection efficiency 10% below typical efficiency
 - Pion fake rate slightly degrades to 2.0%
- Extrapolation uncertainty from magnetic field modeling
 - Verified B field by 10%, no significant performance degradation

Implementation into Belle II Analysis Software Framework (basf2)

- Tested inference latency is 1.81 ± 0.78 ms per $e^+e^- \rightarrow B\bar{B}$ event on KEK CPU cluster (KEKCC)
- Implemented in offline reconstruction phase of Belle II data processing