

# Drift chamber for CEPC 4<sup>th</sup> conceptual detector

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on behalf of DC PID Group

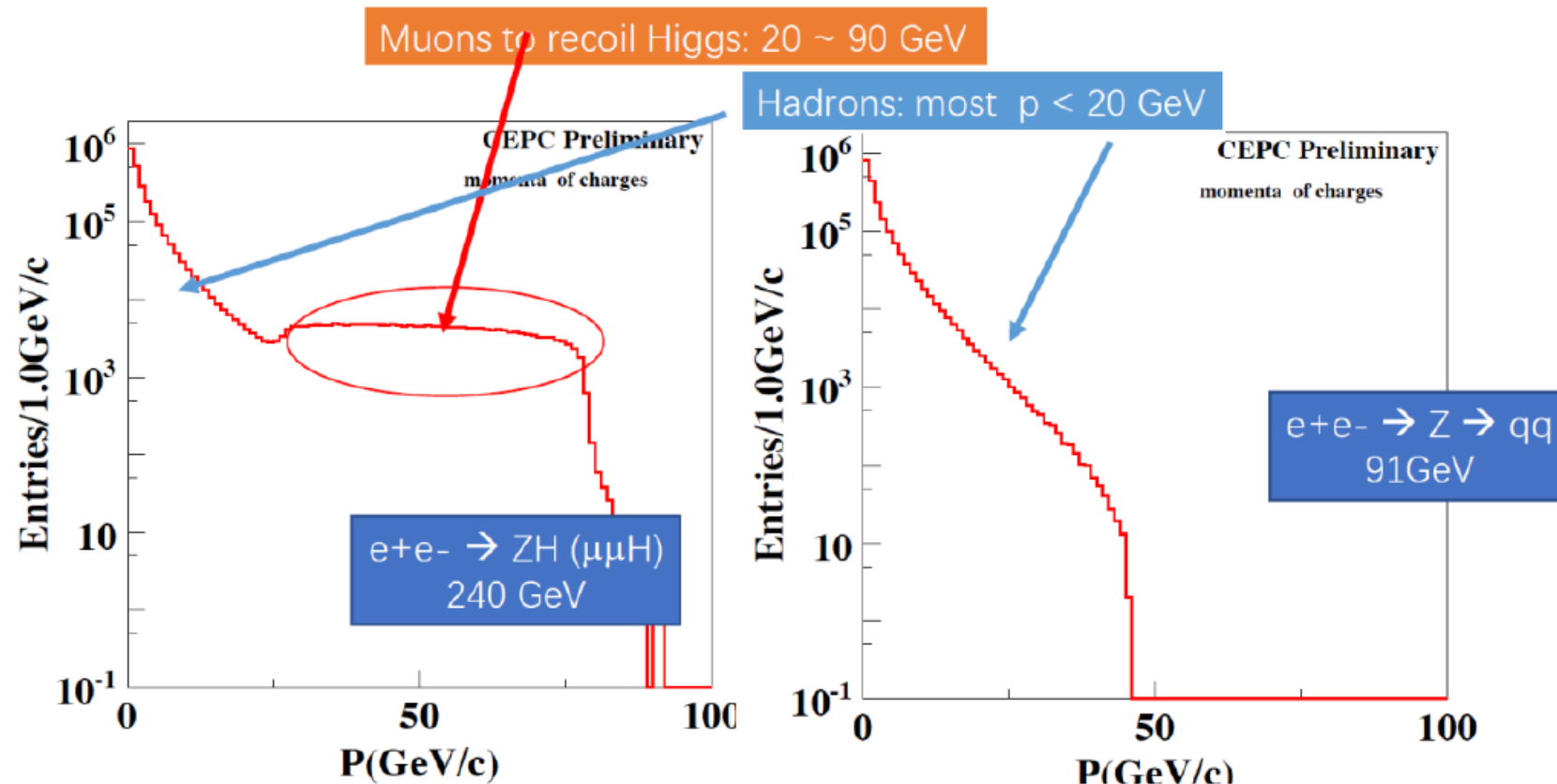
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# Outline

- Introduction of drift chamber for CEPC 4<sup>th</sup> conceptual detector
- Simulation study and detector preliminary parameters
- Cluster reconstruction algorithm with machine learning
- Summary and outlook

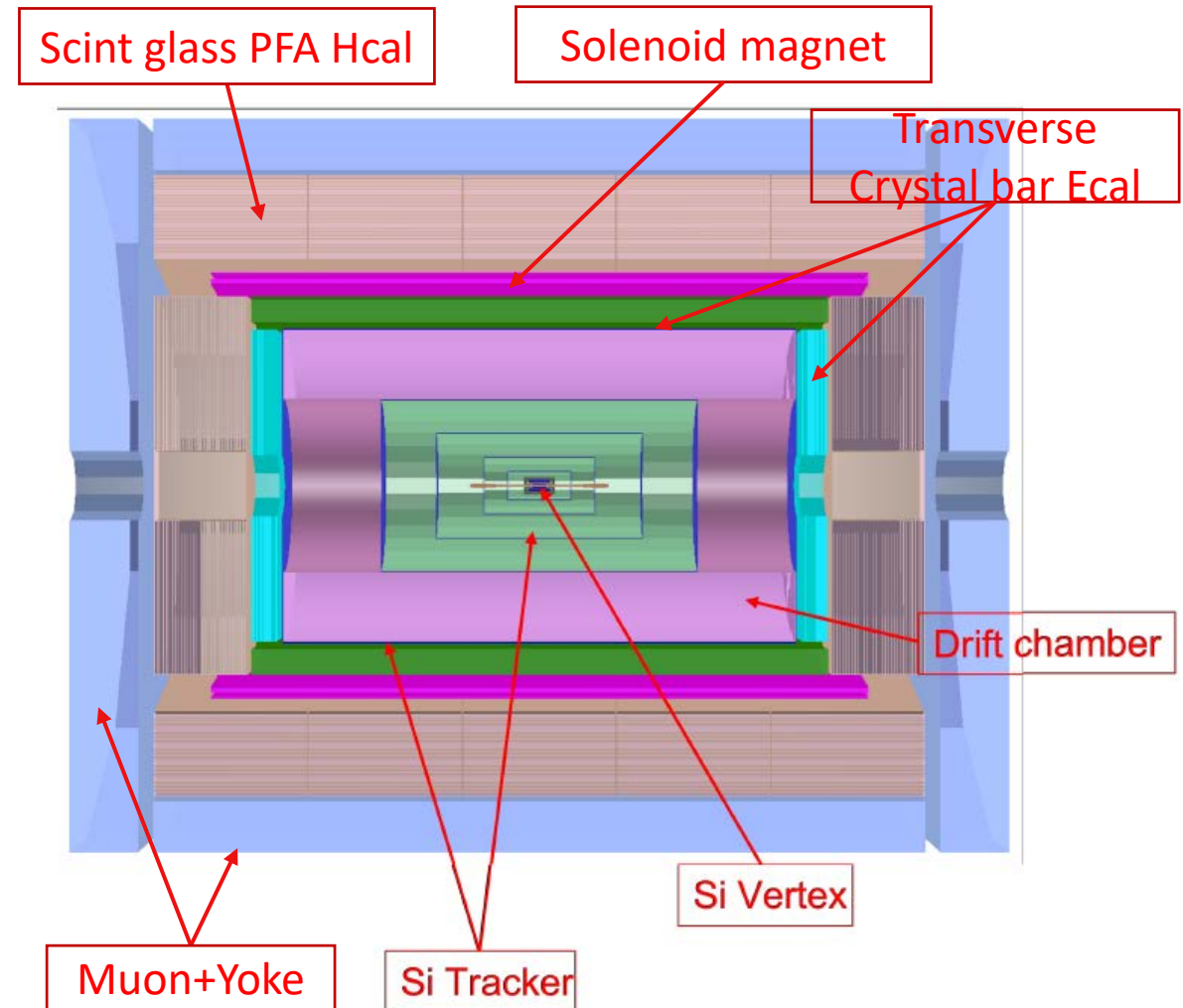
# Motivation

- CEPC will produce  $10^{11}$  -  $10^{12}$  Z bosons, offer great opportunity for flavor physics and jet study
- Most charged hadrons are concentrated on low momentum region (below 20 GeV/c)
- Particle identification (PID) would be critical

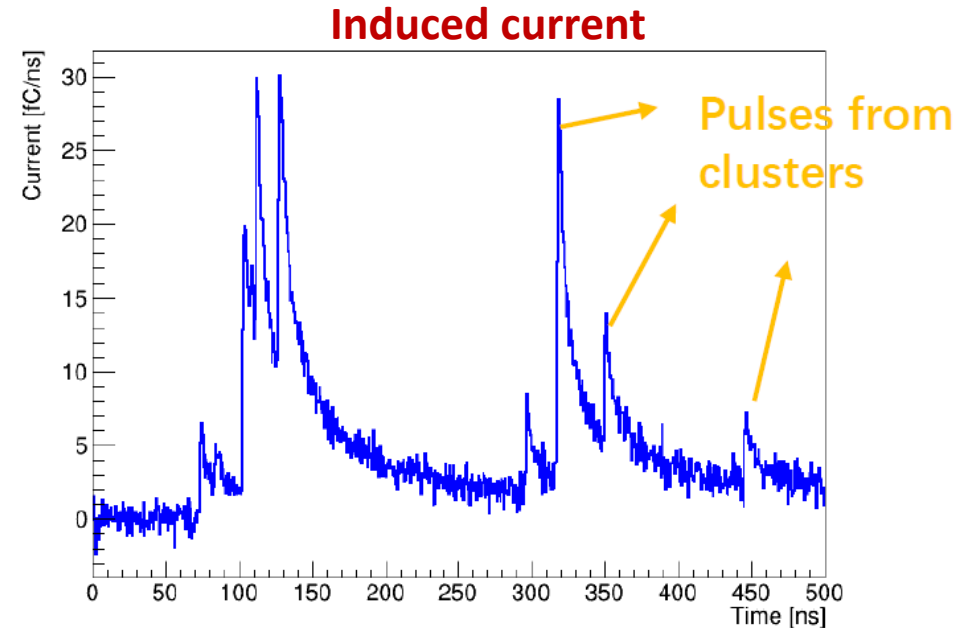
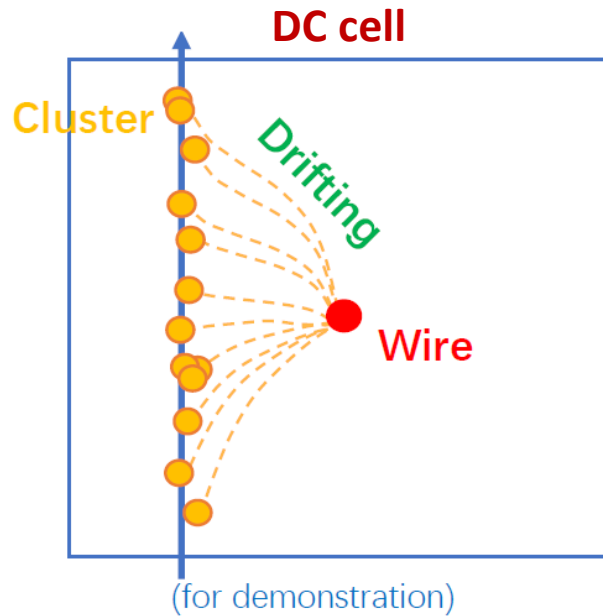


# Drift Chamber for CEPC 4<sup>th</sup> detector design

- Drift Chamber is proposed in CEPC 4<sup>th</sup> conceptual detector for particle identification (PID)
  - Inserted between Si inner tracker (SIT) and Si external tracker (SET)
  - **Mainly provides PID capability** by using cluster counting technique while keeping a reasonable detector size
    - $3\sigma$  separation of K/ $\pi$  with momentum up to  $\sim 20$  GeV/c
  - Could also benefit tracking and momentum measurement



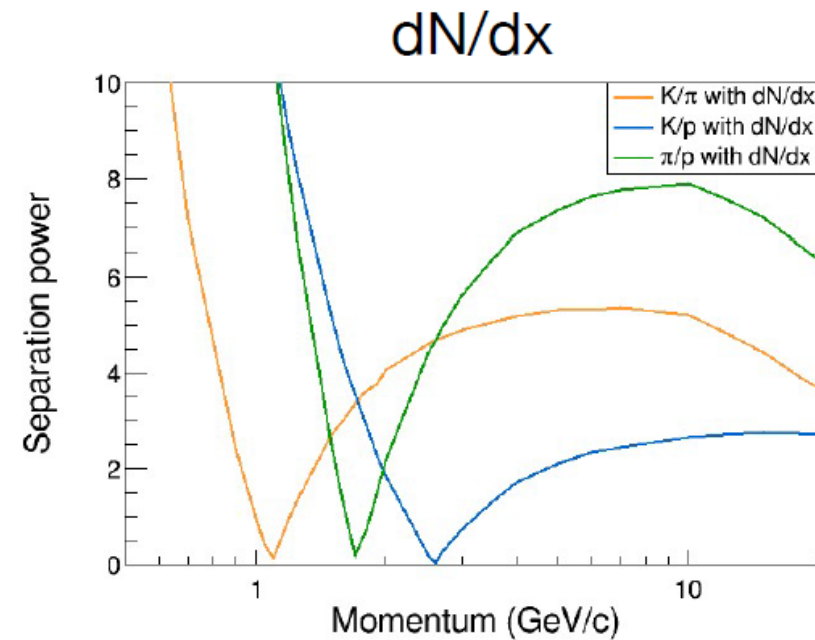
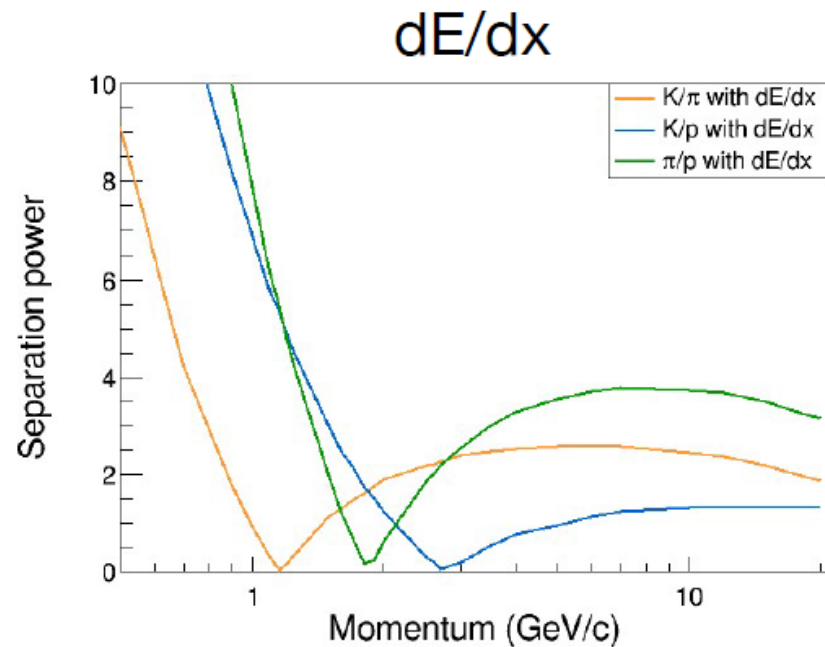
# Ionization measurement with $dN/dx$



- Cluster counting ( $dN/dx$ ): Measure number of clusters over the track
- Yield of primary ionization is Poisson distributed:  $P(\bar{N}_p, k) = \frac{\bar{N}_p^k}{k!} e^{-\bar{N}_p}$
- Less sensitive to Landau tails than  $dE/dx$

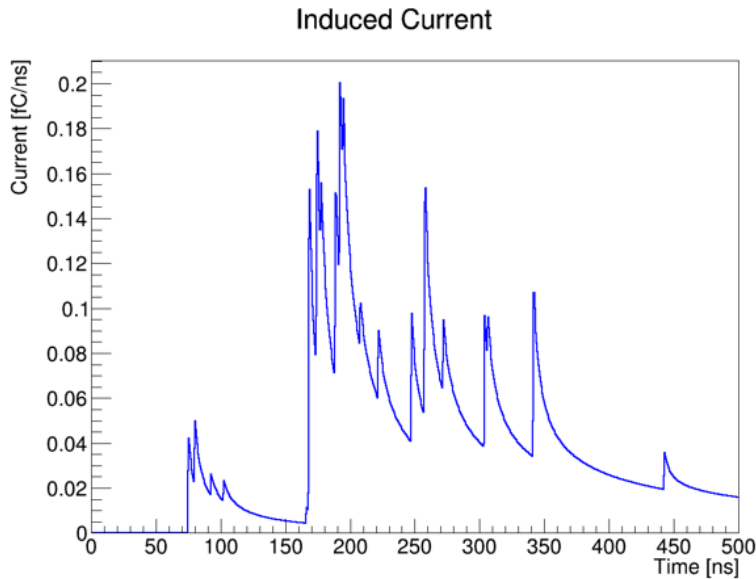
# PID with ionization simulation

- The theoretical results are obtained from Garfield++ simulation.
- $dN/dx$  shows better PID performance than  $dE/dx$ . (a factor of  $>2$  )

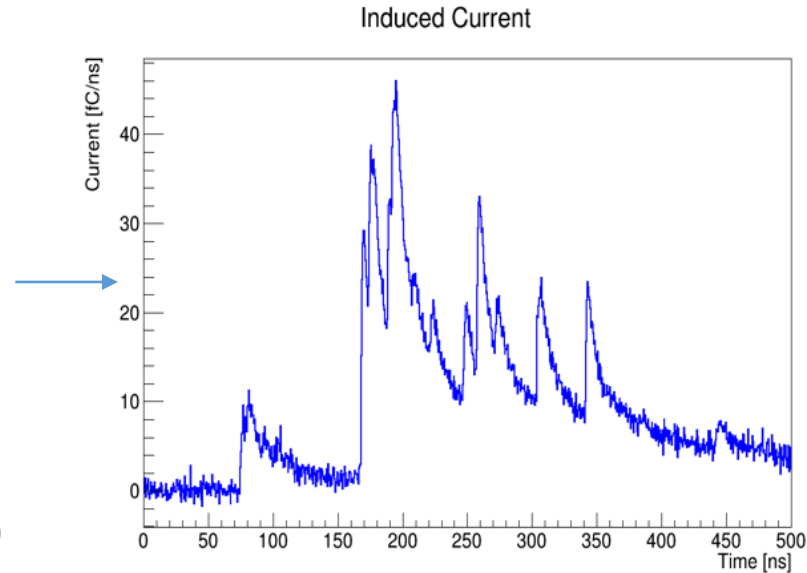


- The effect of electronics, noise should be taken into account, and efficiency of counting algorithm is important
- To study/control those factors, a waveform-based simulation is performed.

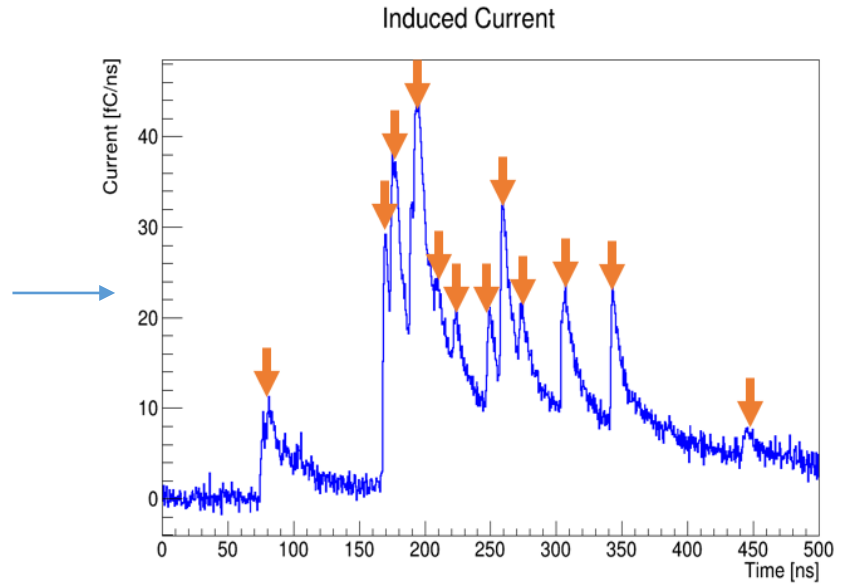
# Waveform-based simulation



**Simulation**



**Digitization**



**Reconstruction**

## Signal generator (Garfield++)

- Heed: ionization process
- Magboltz: gas properties (drift/diffusion)

## Electronics:

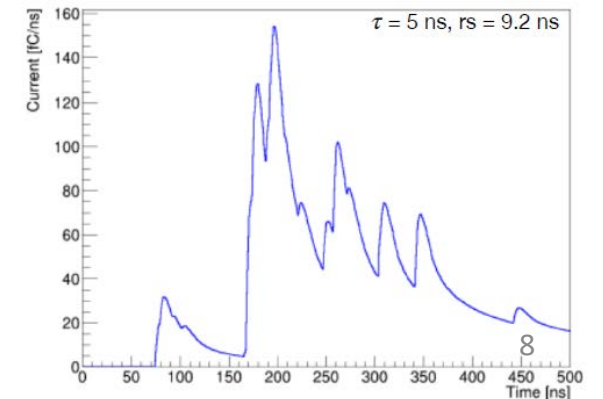
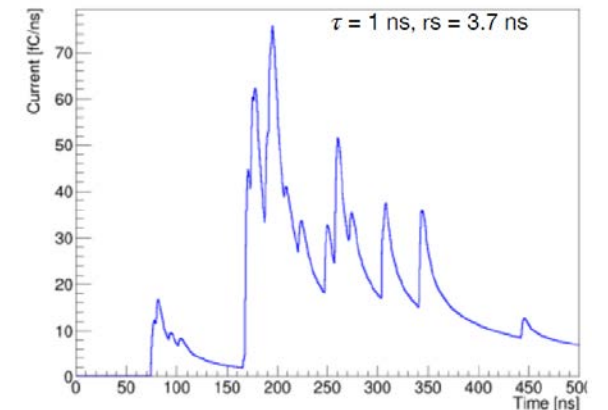
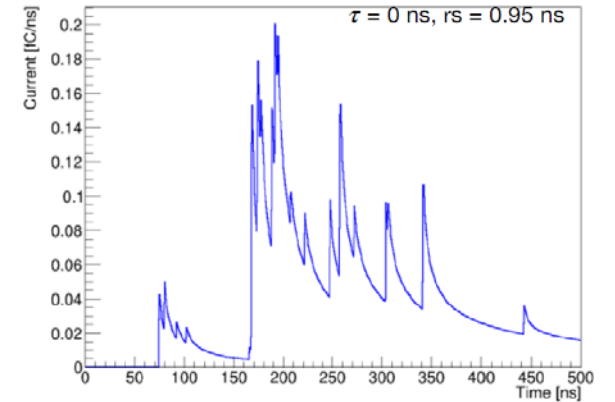
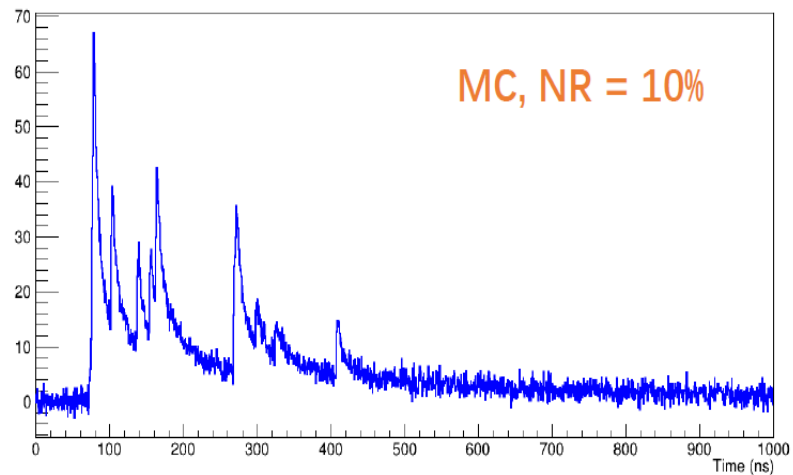
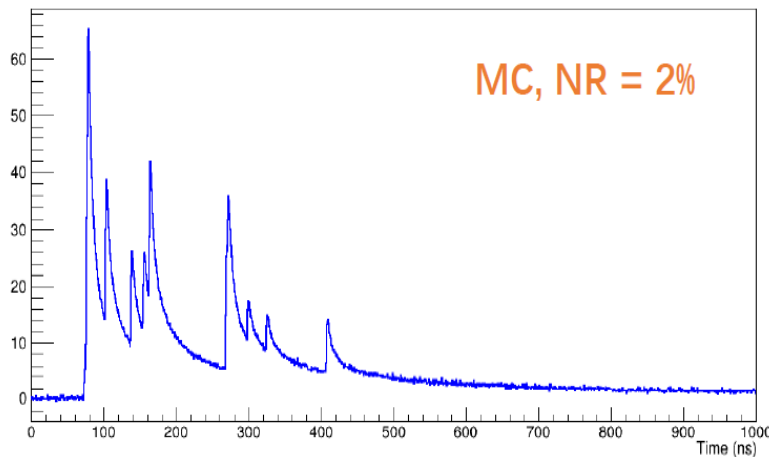
- Preamplifier
- Noise
- ADC sampling rate

## Peak finding algorithm

- Second derivative
- Machine Learning

# Digitization

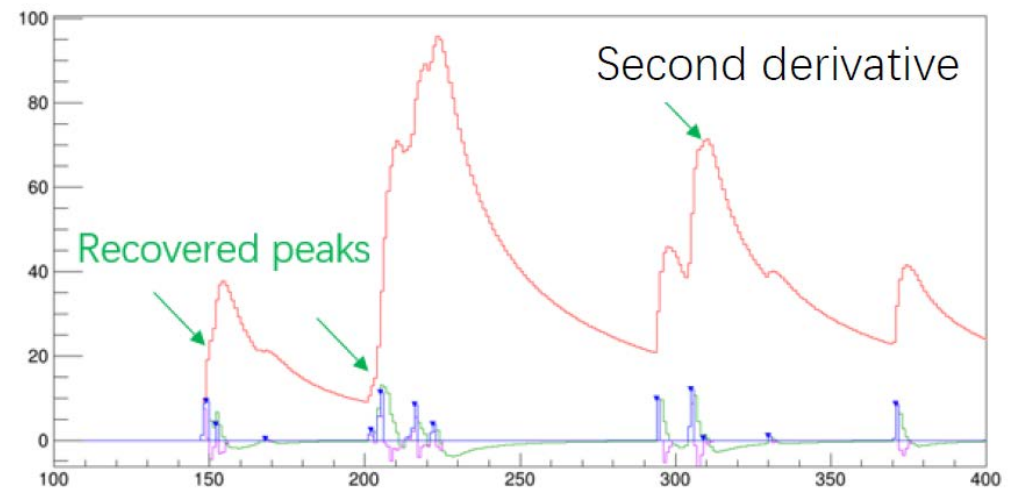
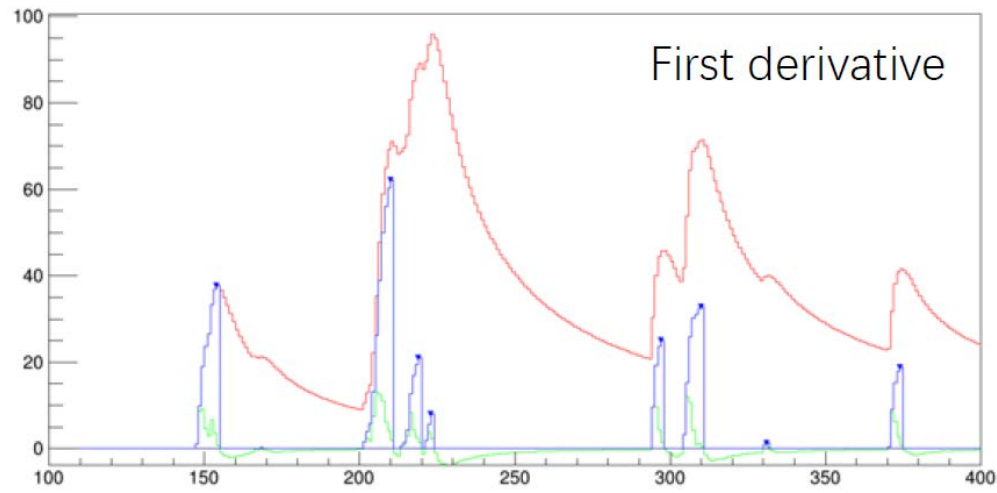
- Impulse response: parameterized by the amplification factor  $A$  and the time constant  $\tau$
- Noise ratio definition:  $\sigma_{Noise} / A_{signal}$ 
  - $A_{signal}$ : Averaged single-pulse amplitude
  - $\sigma_{Noise}$ : Noise RMS





# Reconstruction

- 1<sup>st</sup> and 2<sup>nd</sup> order derivatives
  - Fast and efficient peak finding algorithms
  - Not sensitive to baseline
  - Can be easy to implement in hardware
- More efficient reconstruction algorithm: machine learning

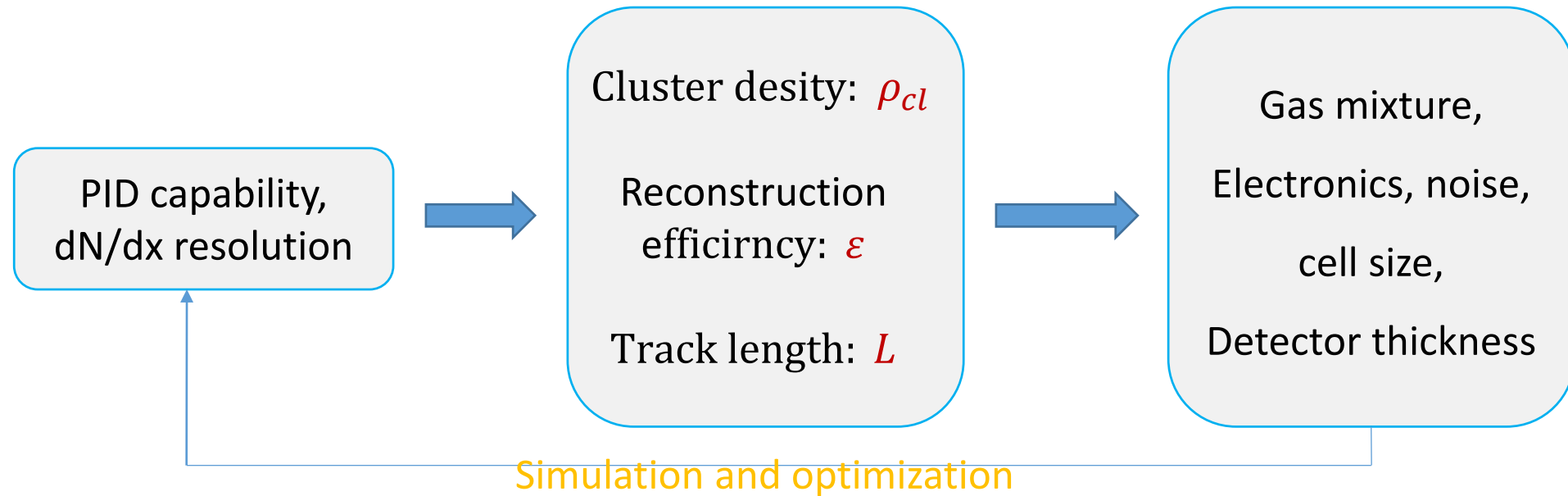


# Detector simulation and optimization

Physics requirement and detector performance

Impact factors

Detector design parameters

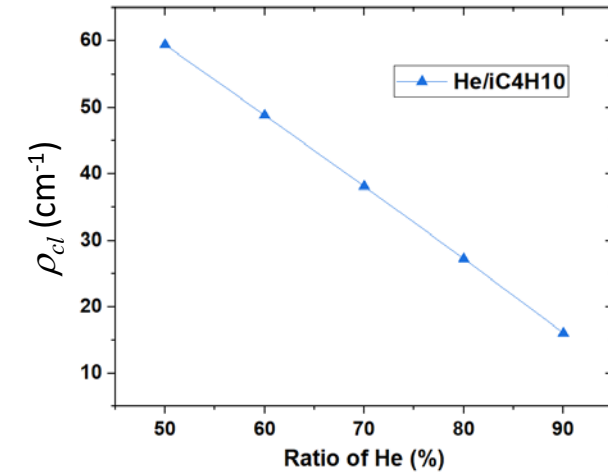


- Parameters in simulation:
  - Track direction:  $\theta=90^\circ$
  - Impact parameter of track w.r.t. sense wire: 0.2 cm

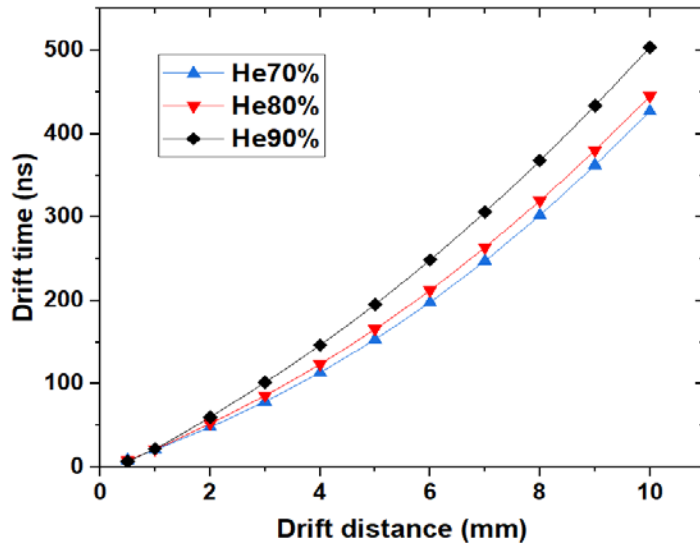
# Gas mixture

- High cluster density  $\rho_{cl}$  compatibly with cluster counting efficiency  $\varepsilon$
- Low drift velocity helps to identify clusters in time
- Small longitudinal diffusion benefits both dN/dx measurement and spatial resolution
- Prefer smaller cluster density, slower drift velocity, smaller longitudinal diffusion. He/iC<sub>4</sub>H<sub>10</sub> =90/10 is good

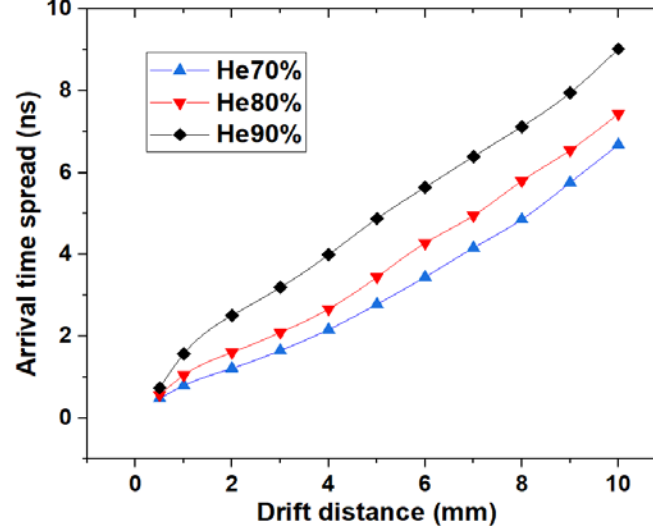
Cluster density vs ratio of He



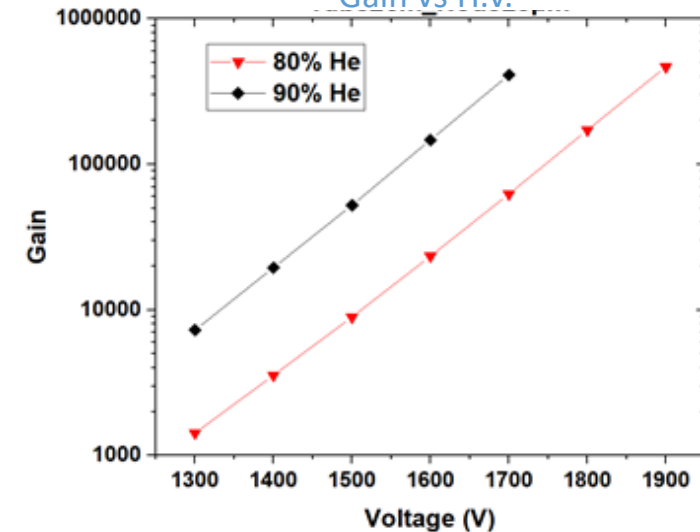
Drift time vs drift distance



Diffusion effect vs drift distance

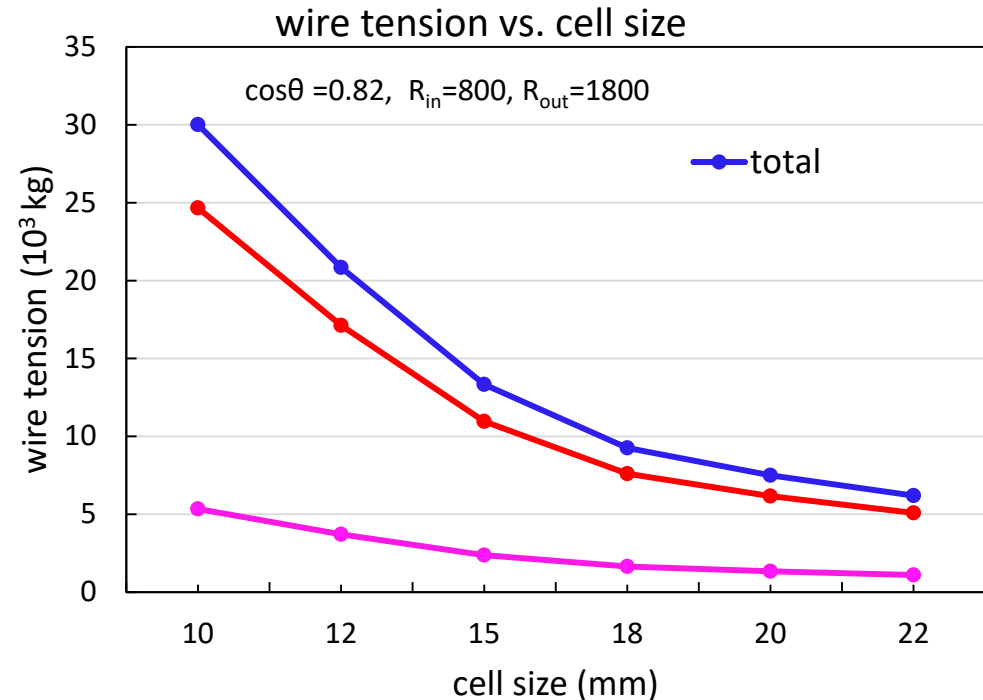
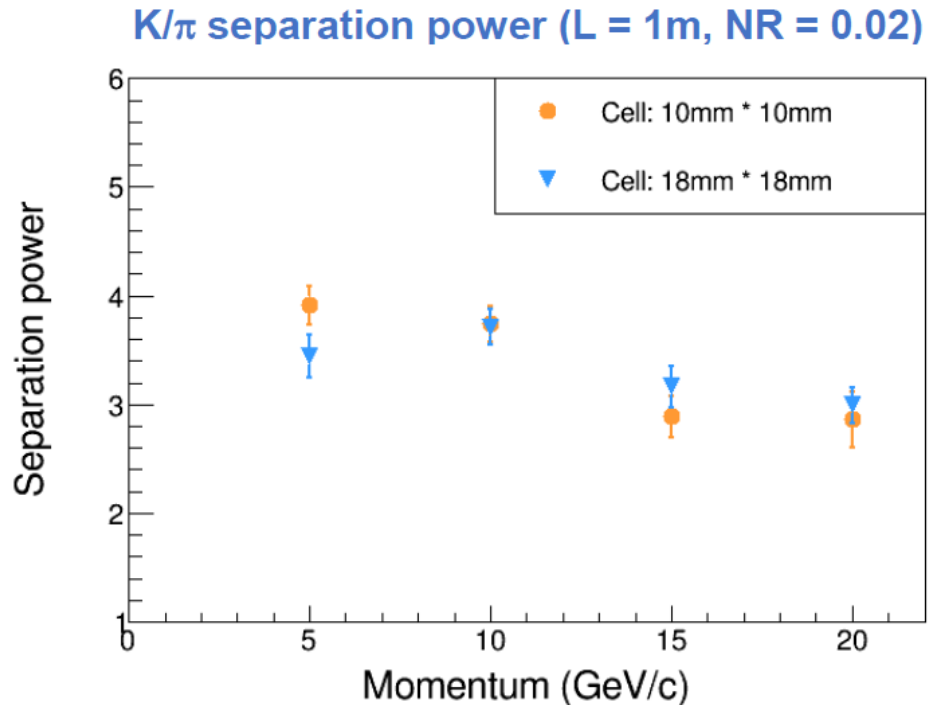


Gain vs H.V.



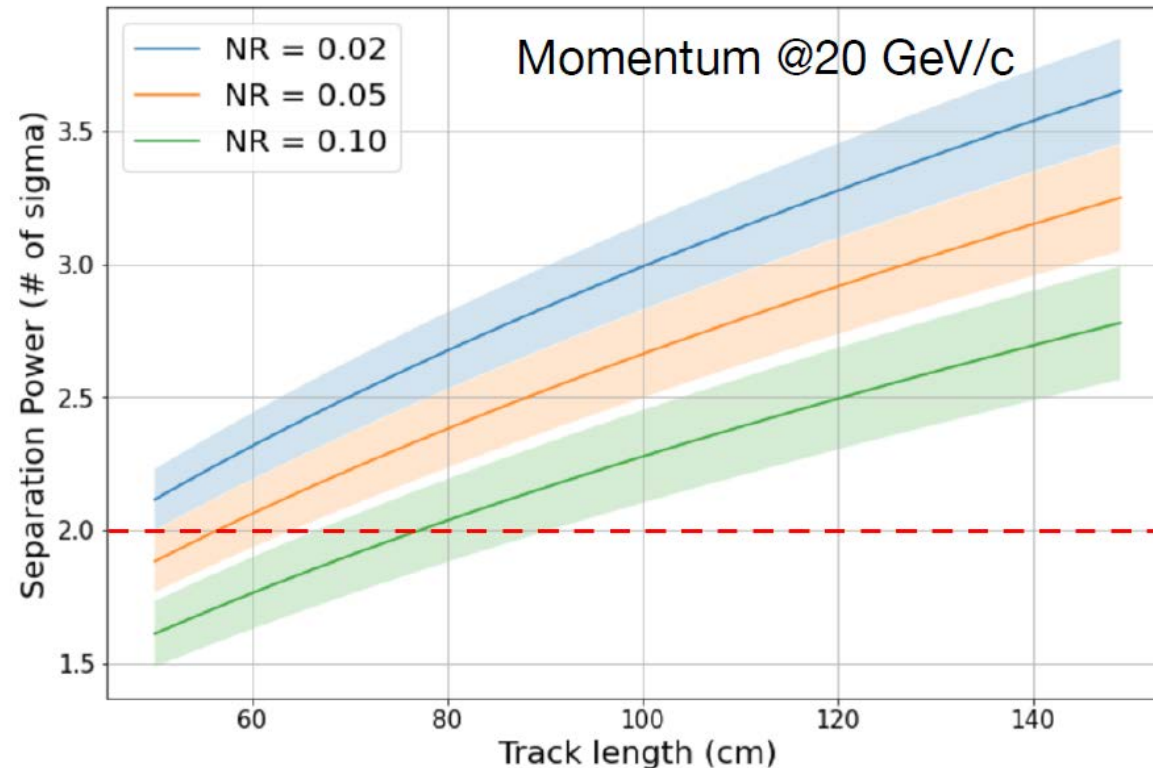
# Cell size

- Only total track length affects the PID with cluster counting technique, not the granularity
- The cell size impacts on engineering
- Prefer less wire tension, reduced number of wires => larger cell

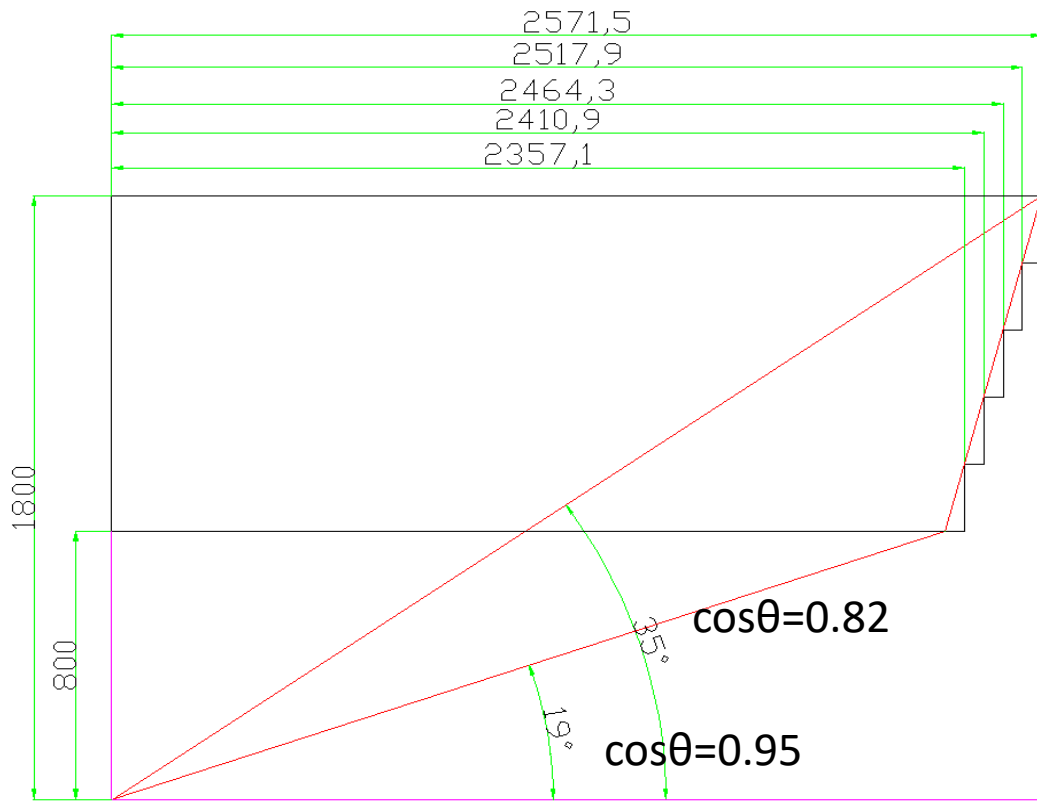


# K/ $\pi$ separation vs. track length

- Small radial thickness (while keeping sufficient PID performance) will make the CEPC detector more compact and reduce the cost of other detectors (e.g., SiTrk, Calorimeters)
- Better than  $2\sigma$  separation at 20 GeV/c is achieved within 1m track length



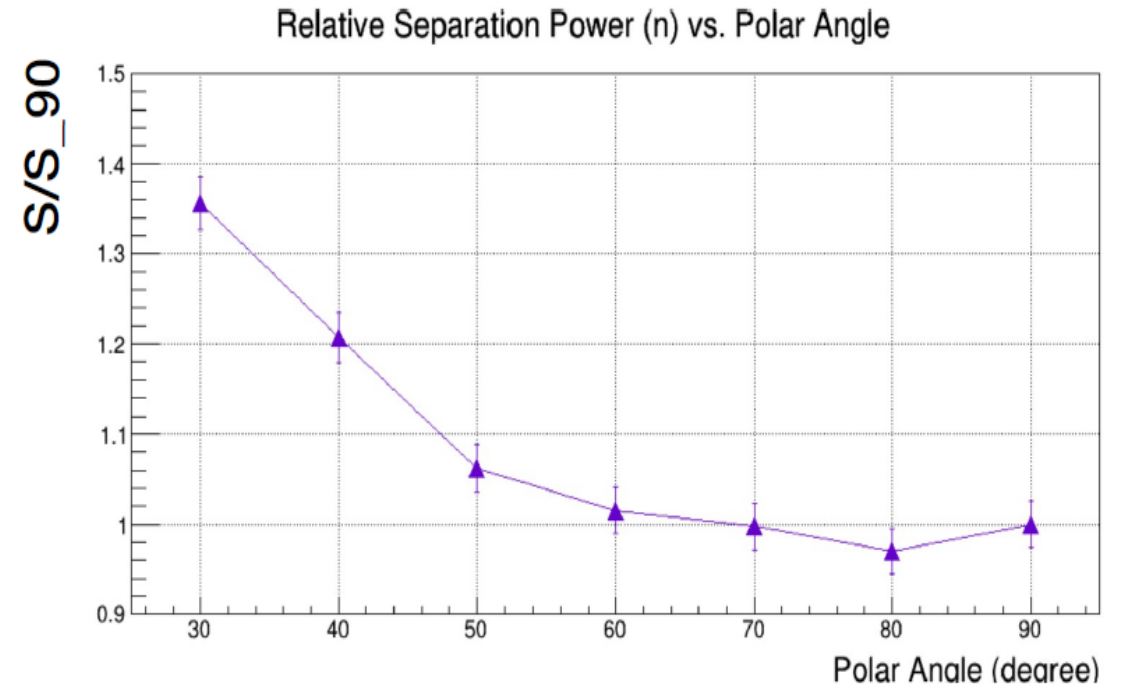
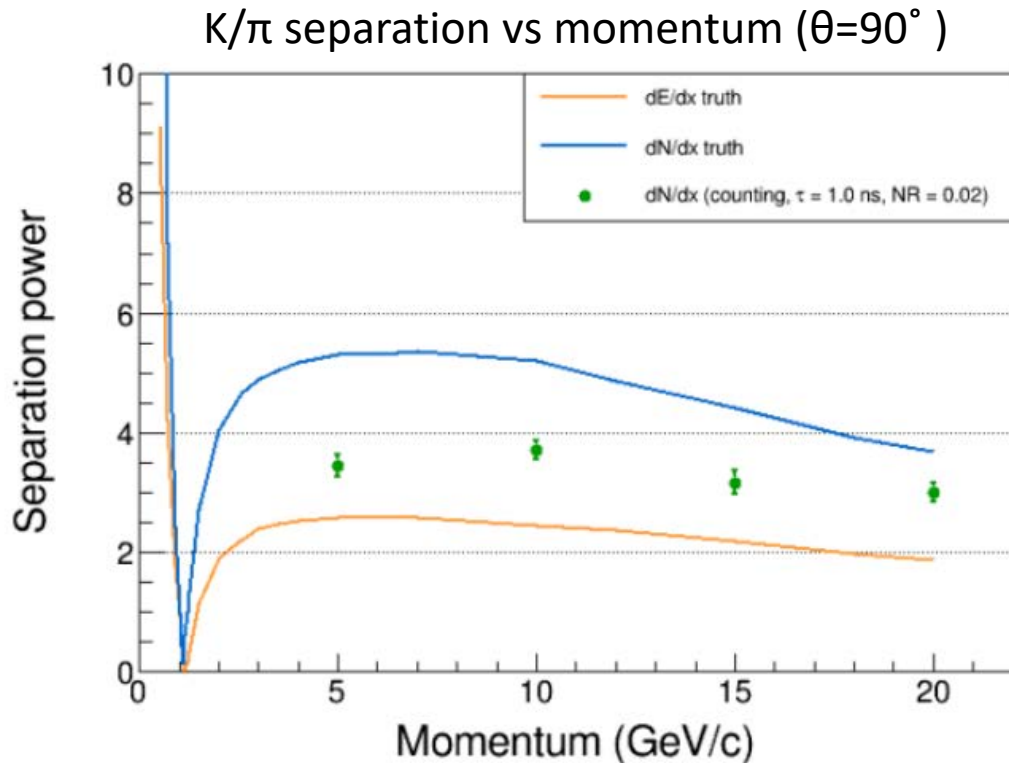
# Preliminary parameters



Radius extension	800-1800mm
Length of outermost wires ( $\cos\theta=0.82$ )	5143mm
Thickness of inner CF cylinder:	200 $\mu$ m
Outer CF frame structure:	Equivalent CF thickness: 1.63mm
Thickness of end Al plate	35mm
Cell size:	$\sim 18 \text{ mm} \times 18 \text{ mm}$
Number of cell	24766
Ratio of field wires to sense wires	3:1
Gas mixture	He/ <i>i</i> C <sub>4</sub> H <sub>10</sub> =90:10

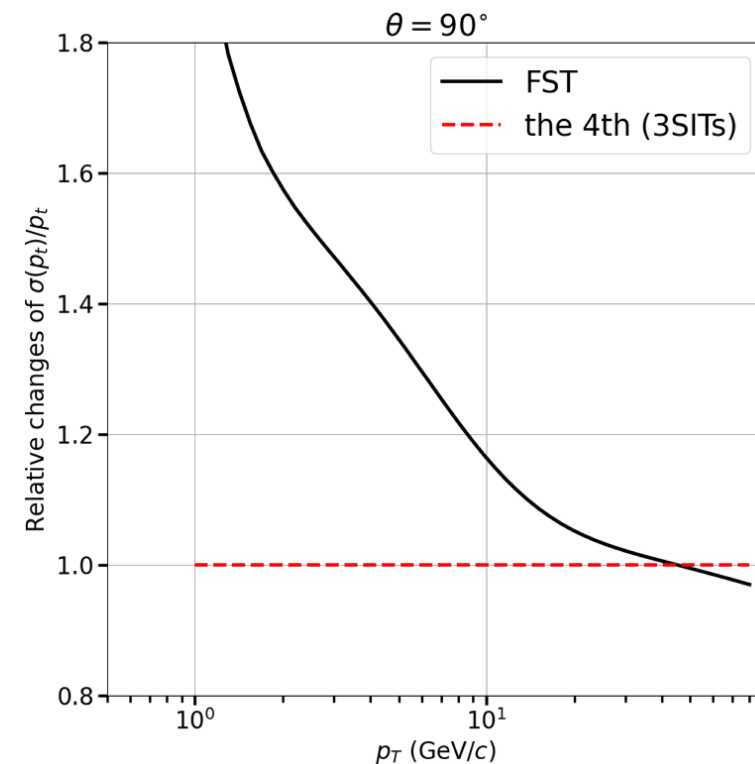
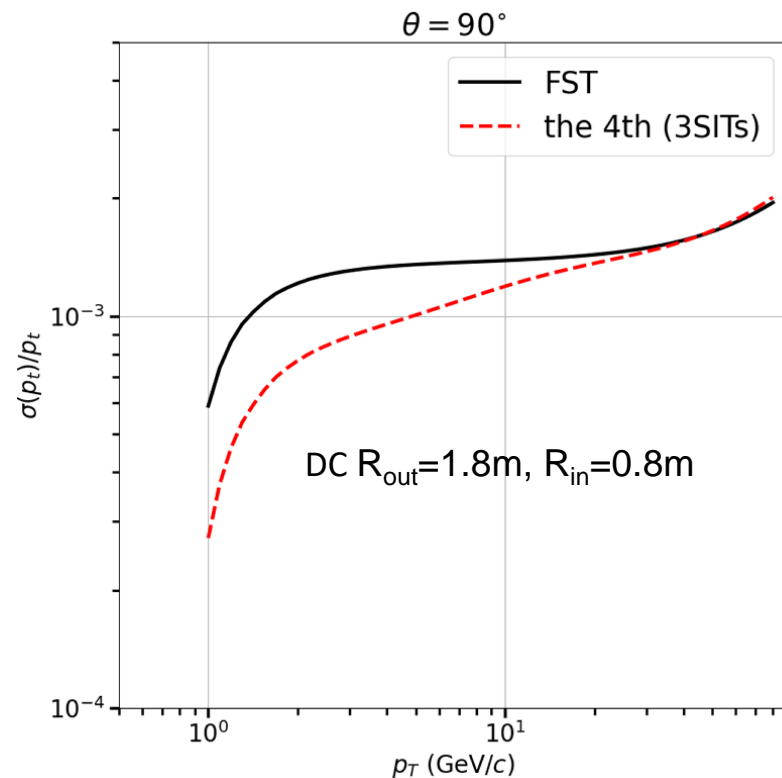
# K/ $\pi$ Separation power

- $3\sigma$  K/ $\pi$  separation at 20GeV/c, 1.5 better than dE/dx truth ( $\theta=90^\circ$ , NR = 0.02)
- Polar angle scan: long track length allows better separation power
- Studies with physics channels are ongoing



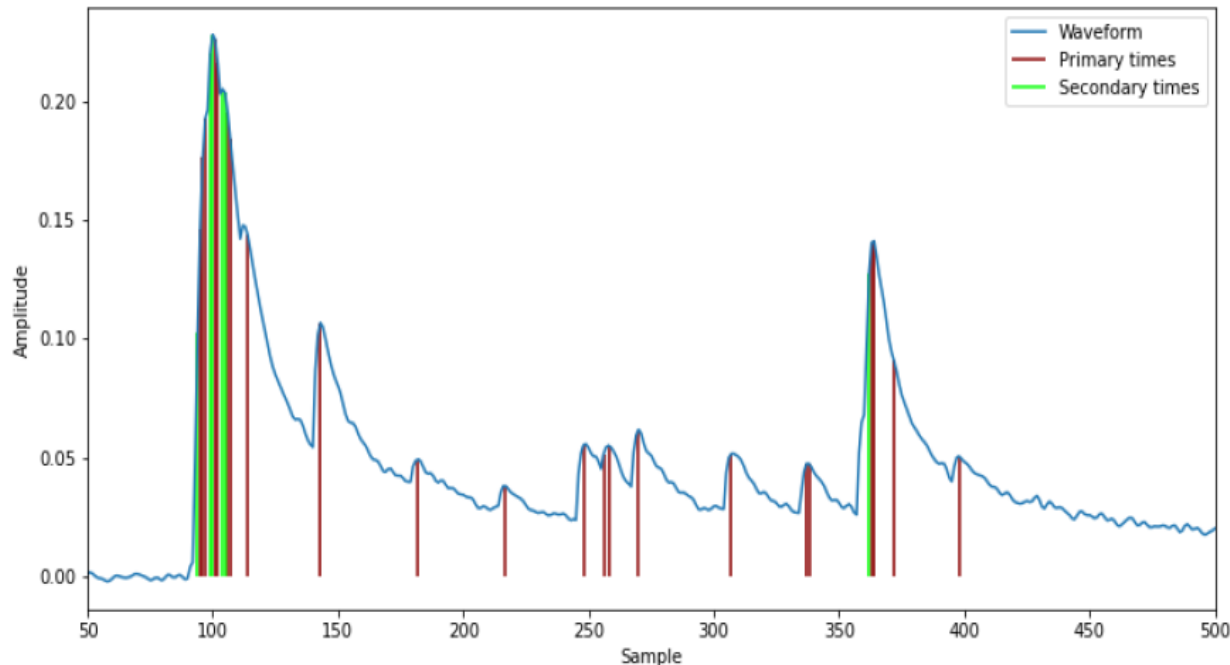
# Impact on momentum measurement

- Compared with full silicon tracker (FST),  $P_T$  resolution of the hybrid tracker system with drift chamber
  - Improved significantly in momentum range of 0-20 GeV/c
  - Almost no degradation with momentum up to 80 GeV/c





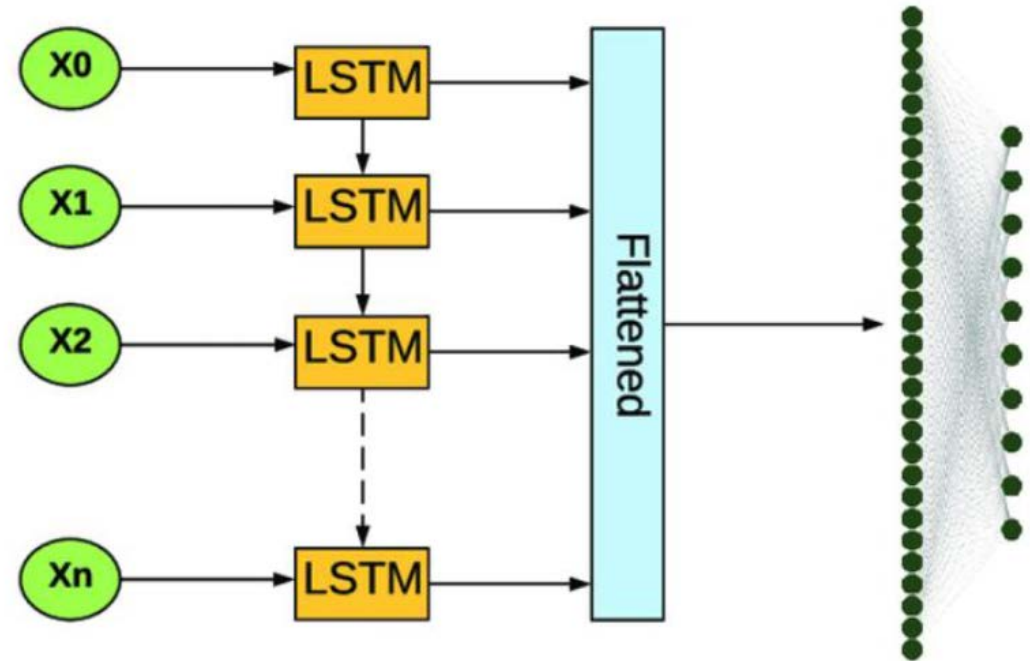
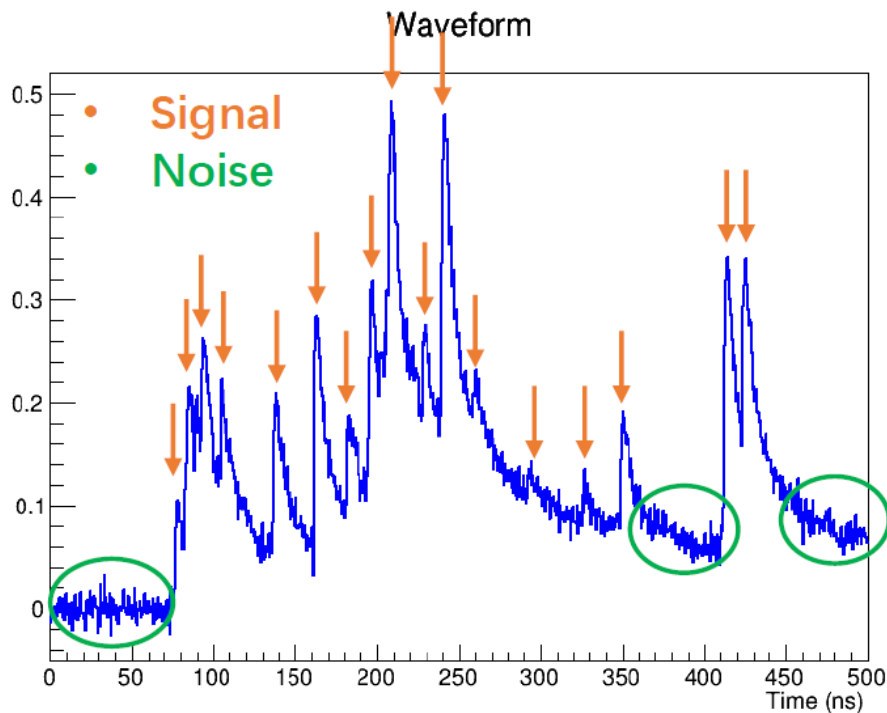
# Reconstruction algorithm: machine learning



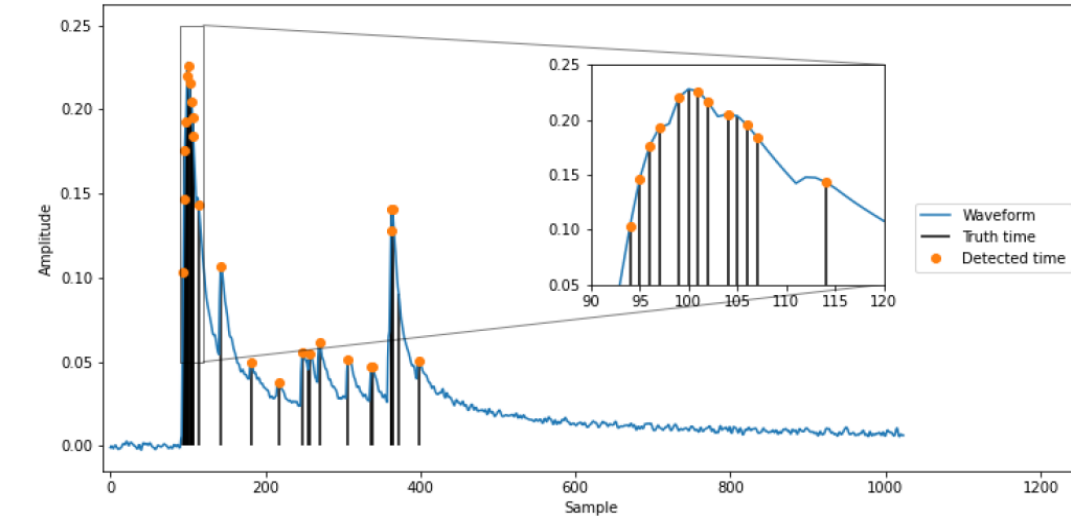
- Both primary electrons and secondary electrons contribute peaks on the waveform
- Find the number of peaks from primary electrons
- Machine learning can make full use of the waveform information,
- More effective in high pile-up region and less sensitive to noise

# Step 1: Peak finding

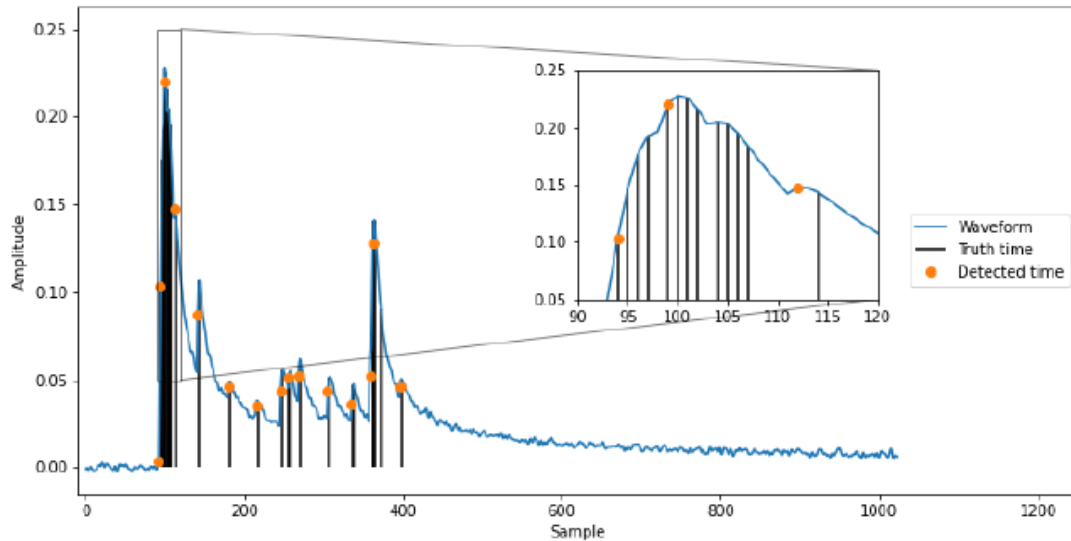
- To detect all peaks from the waveform
- Classification for “signals” and “noises”
- Architecture: Recurrent Neural Network (RNN)



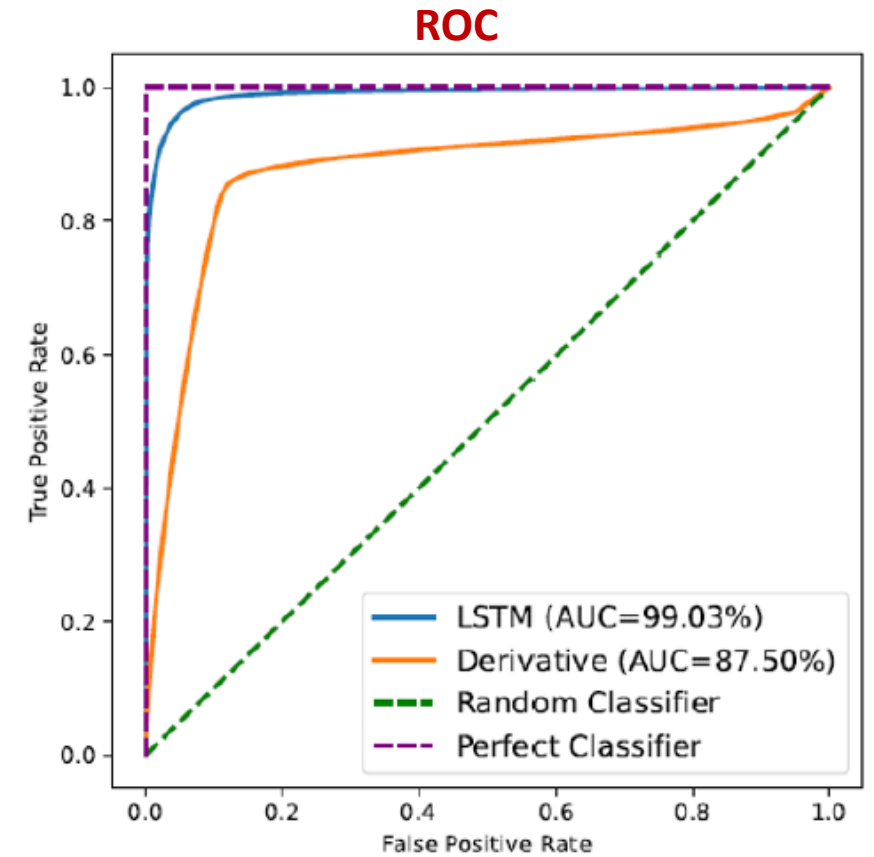
# Compare to the derivative algorithm



**Neural  
Network**



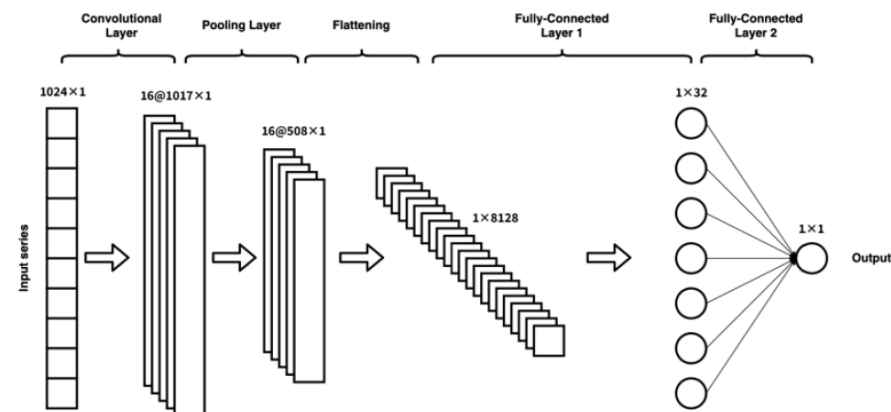
**Derivative**



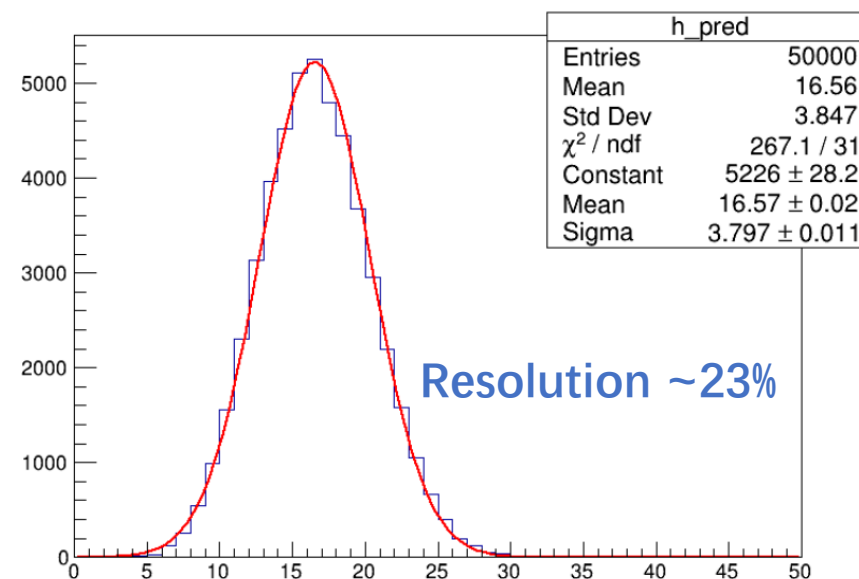
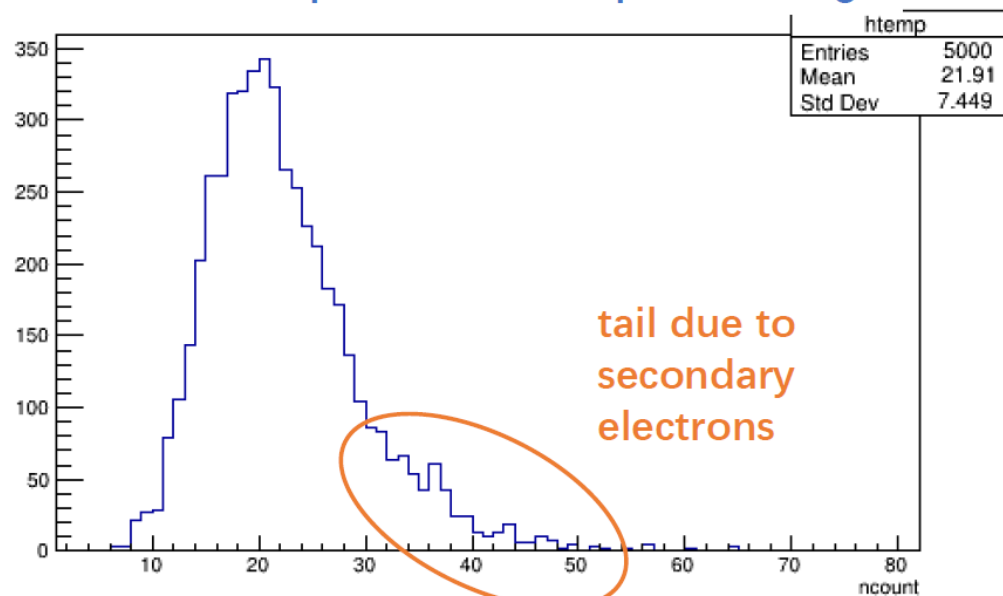
**The NN can find the peaks more effective!**

# Step 2: Discrimination of the primaries

- To discriminate primary peaks from the secondary ones
- Regression problem
- Architecture: 1D Convolutional Neural Network (CNN)



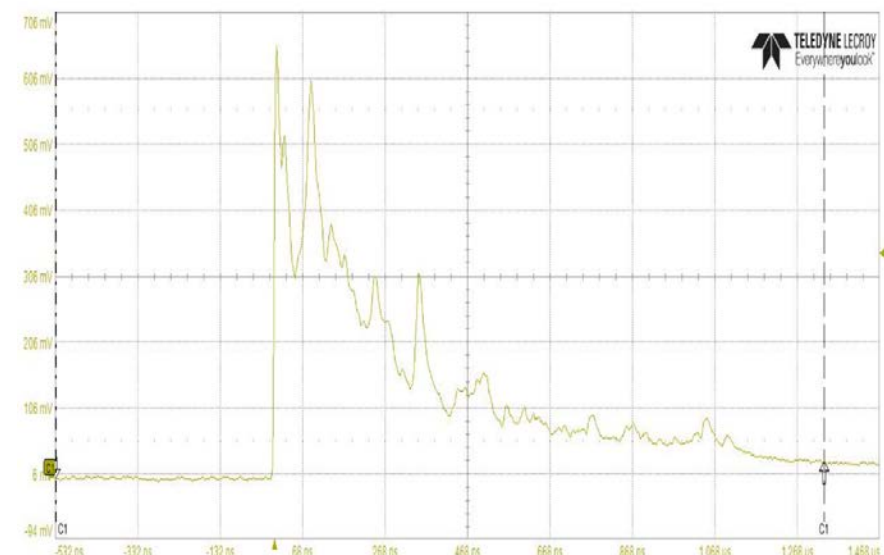
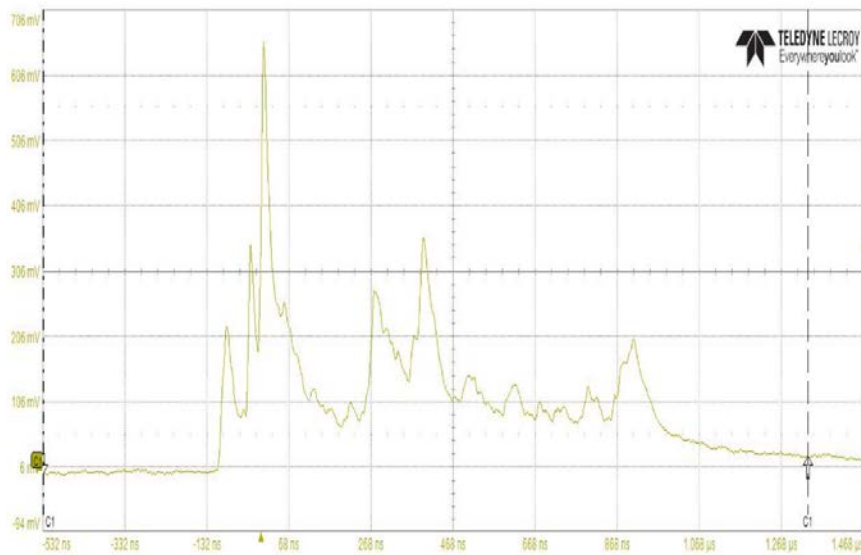
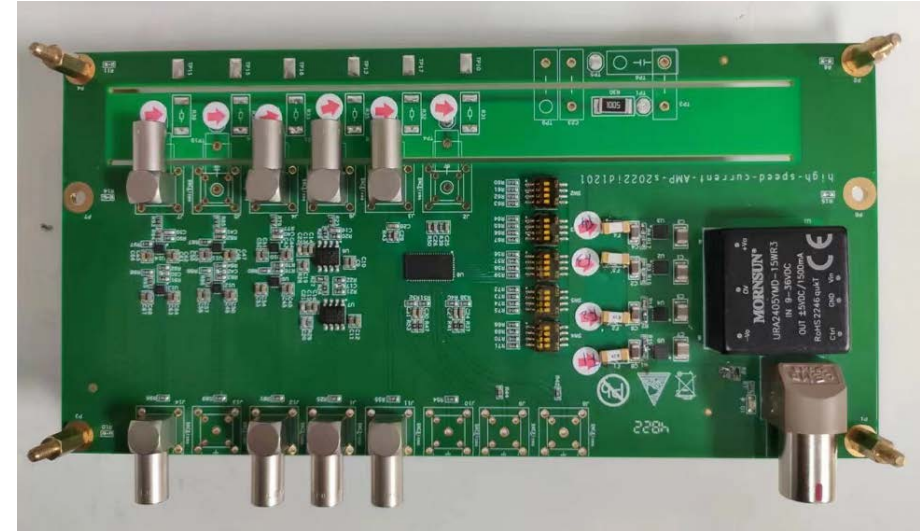
Detected peaks from the peak finding



Very good Gaussian-like distribution  
✓ The resolution is very close to the truth value (~21%), which implies possible improvement on PID

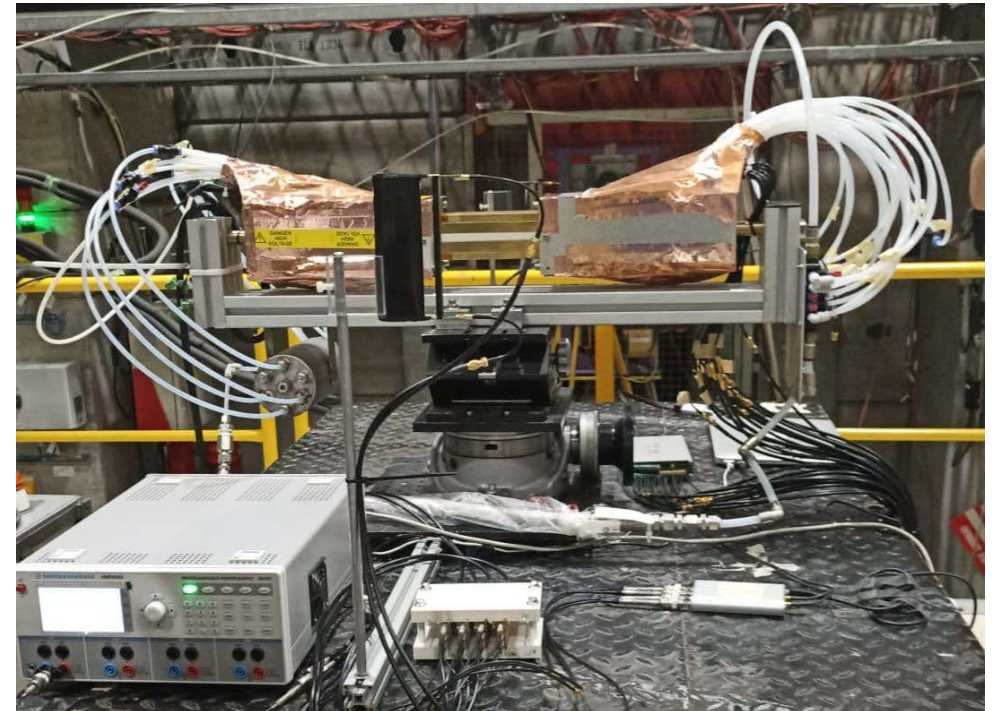
# Electronics development

- High bandwidth current sensitive preamplifiers based on LMH6522 and AD8099 were designed
- Test with a detector prototype



# Collaboration with INFN group

- Joined prototype beam test organized by INFN group
  - First round: Nov. 2021 (Shuiting joined)
  - Second round: July 2022 (Shuaiyi joined)
  - Data analysis
- Regular meeting between INFN and IHEP
- Collaboration on MOST-MAECI project application, focusing on
  - Beam tests for the application of the cluster counting/timing techniques
  - Simulation and reconstruction
  - Design and deployment of real-time ML algorithms on FPGAs



# Summary and outlook

- Drift Chamber is proposed in CEPC 4<sup>th</sup> conceptual detector for particle identification
- Simulation studies show that  $3\sigma$  K/ $\pi$  separation at 20GeV/c can be achieved with 1m track length and 2% noise level
- Cluster counting algorithm based on machine learning is developed and shows promising performance for MC samples
- Plans:
  - Optimize the reconstruction algorithm and apply to beam test data
  - Extract dN/dx parameters from full simulation and perform physics studies
  - Detector and readout electronics prototyping and test