

Progress on cluster counting for the 4th concept DC

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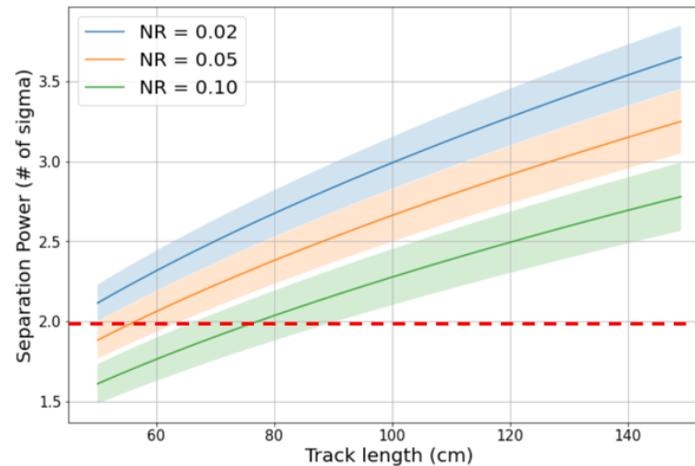
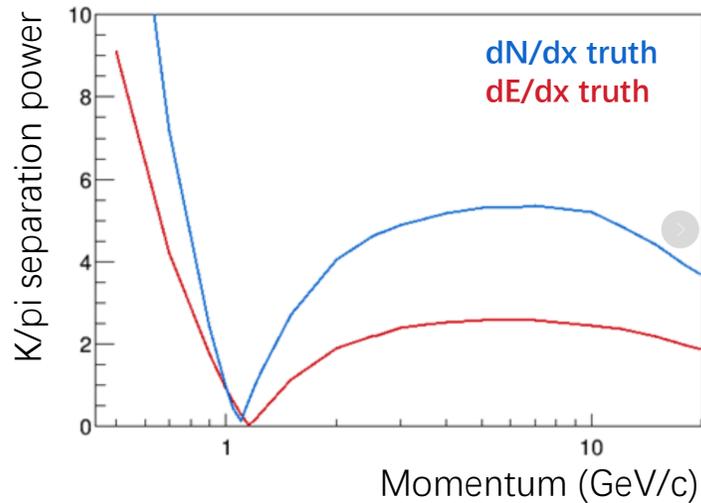


Motivation

- **Excellent PID is required for CEPC physics programs**
 - Preliminary requirement: $> 2\sigma$ for 20 GeV/c pion/kaon separation
- **The cluster counting (dN/dx) is the most promising breakthrough in PID**
 - Resolution has, potentially, a factor of 2 better than the dE/dx
- **Drift chamber with cluster counting can provide sufficient PID power while keeping a reasonable detector size**

Recaps of simulation

- **Ideal simulation (truth):**
 - K/pi separation with dN/dx is x2 better than dE/dx
- **Full simulation (more realistic, w/ electronics/noises/reconstruction):**
 - 2σ K/pi separation at 20 GeV/c
- **Determine DC parameters with full simulation**



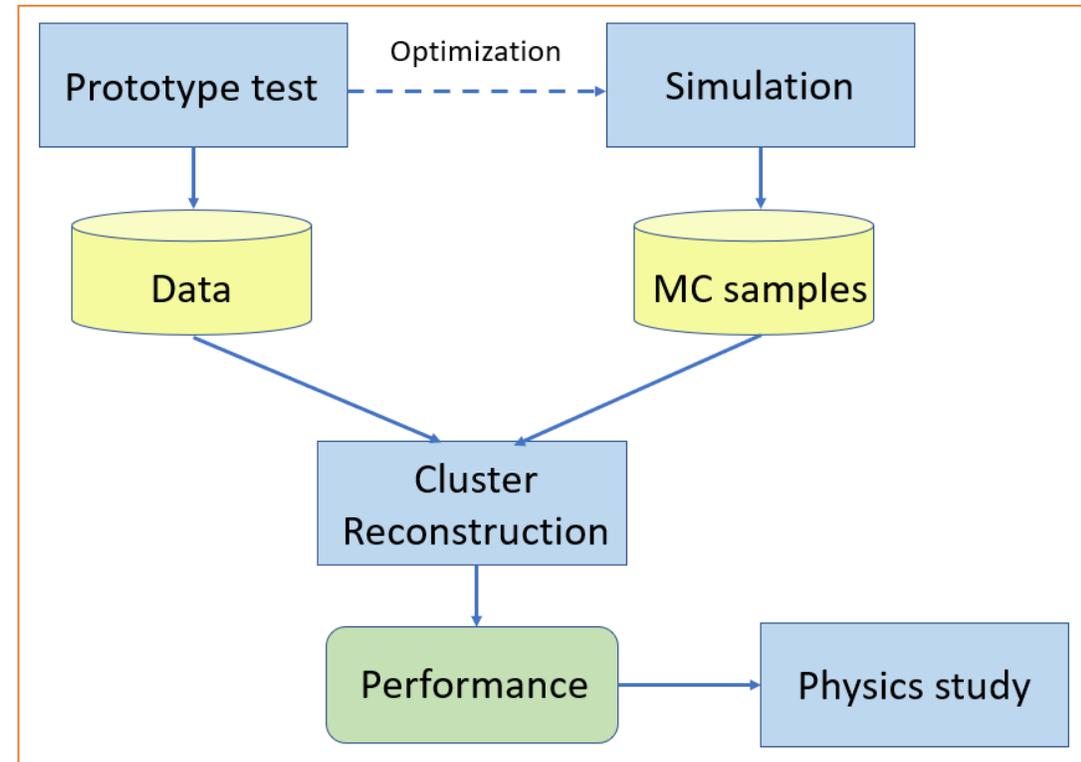
The resolution scales with $L^{-0.5}$
 (Noise ratio in beam-test is close to 10%)

DC Parameters	
Radius extension	800-1800 mm
Length of outermost wires ($\cos\theta=0.82$)	5143 mm
Thickness of inner CF cylinder	200 μm
Outer CF frame structure	Equivalent CF thickness: 1.63 mm
Thickness of end Al plate	35 mm
Cell size	18 mm \times 18 mm
# of cells	24766
Ratio of field wires to sense wires	3:1
Gas mixture	He/iC ₄ H ₁₀ =90:10

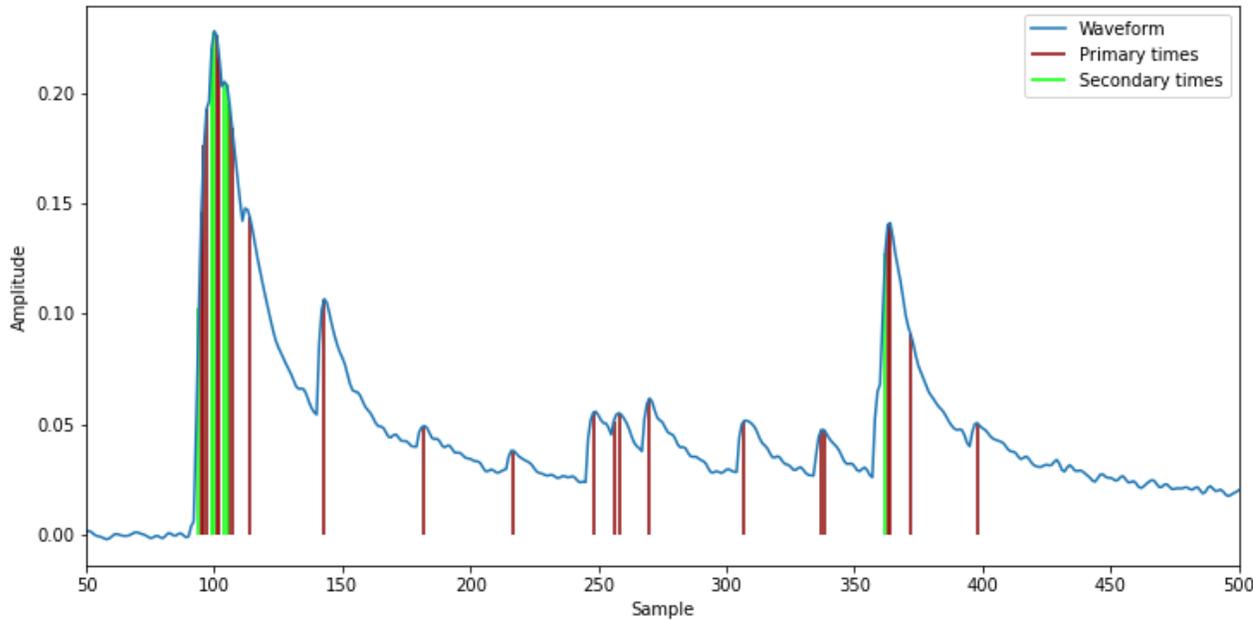
Major efforts in this talk

- Reconstruction algorithm
- Understanding beam test data

Cluster counting workflow



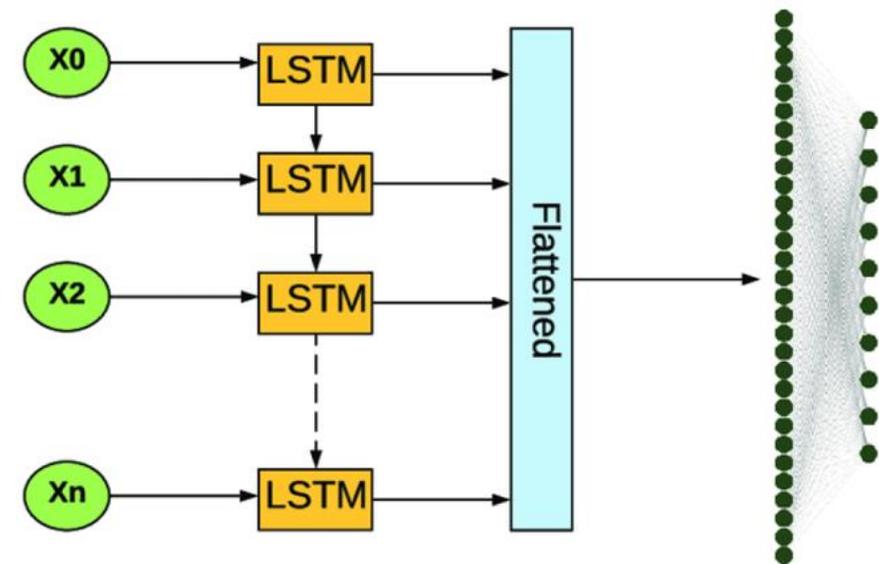
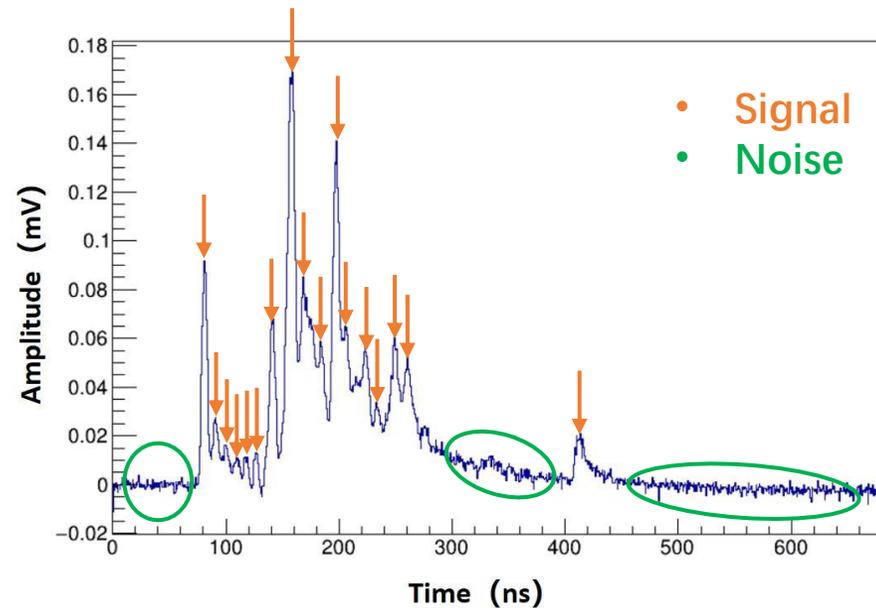
Reconstruction algorithm



- **Task:**
 - Both **primary electrons** and **secondary electrons** contribute peaks on the waveform
 - Find the number of peaks from **primary electrons**
- **Difficulties**
 - High pile-up
 - Could be noisy
- **Machine learning** can make full use of the waveform information, could be effective

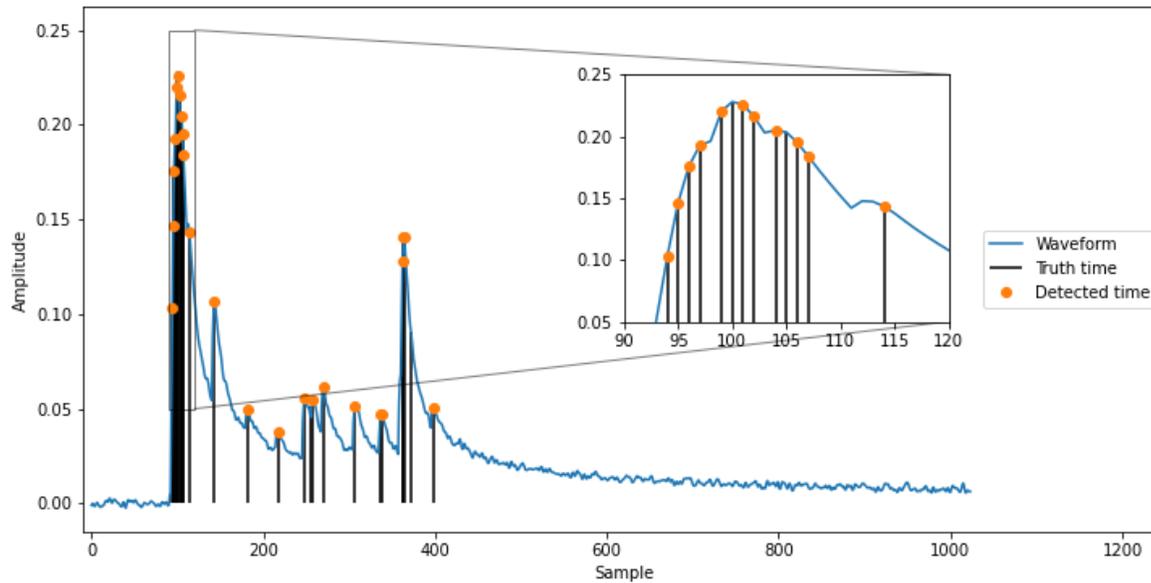
Step 1: Peak finding

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



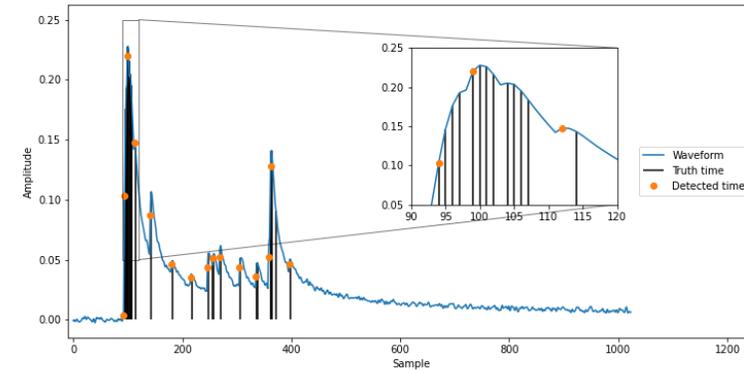
Compare to the derivative algorithm

Neural Network

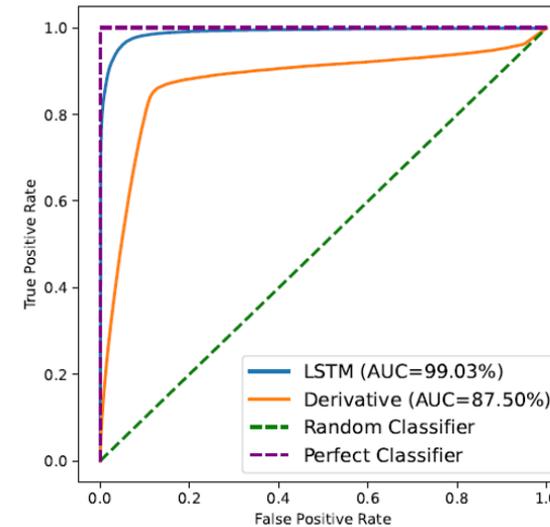


The NN can find the peaks more effective!

Derivative



ROC

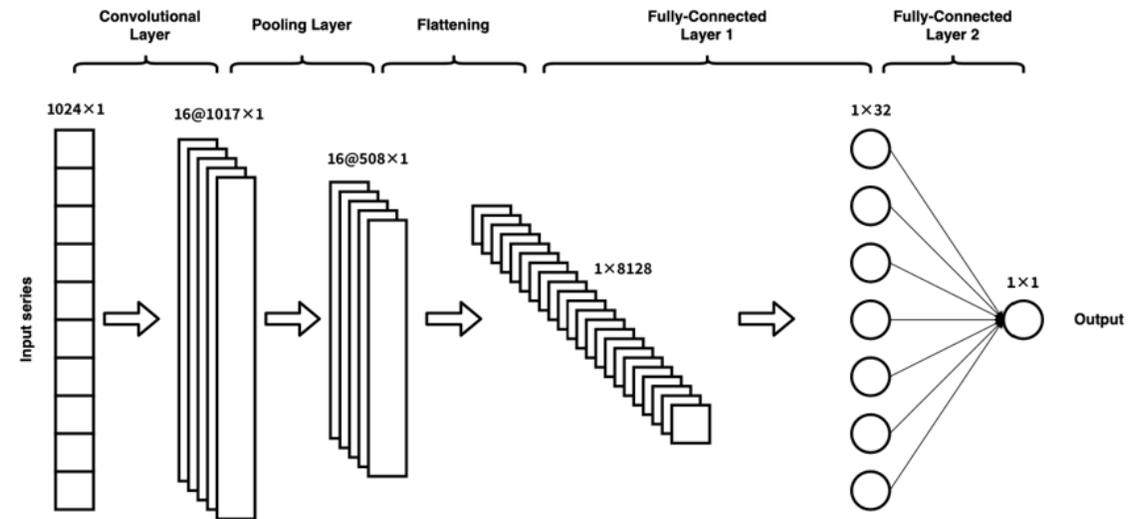
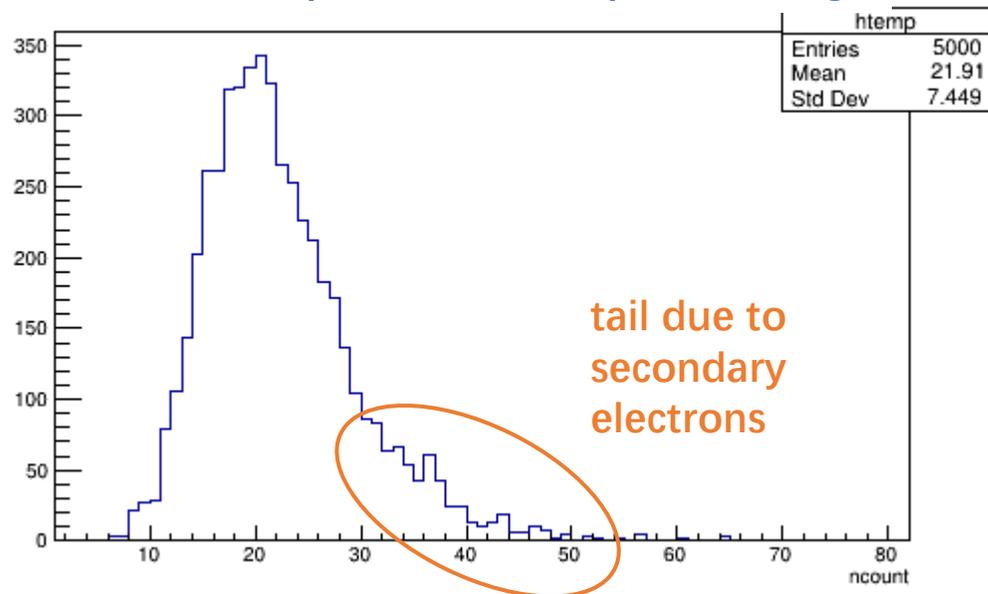


- RNN
- Derivative

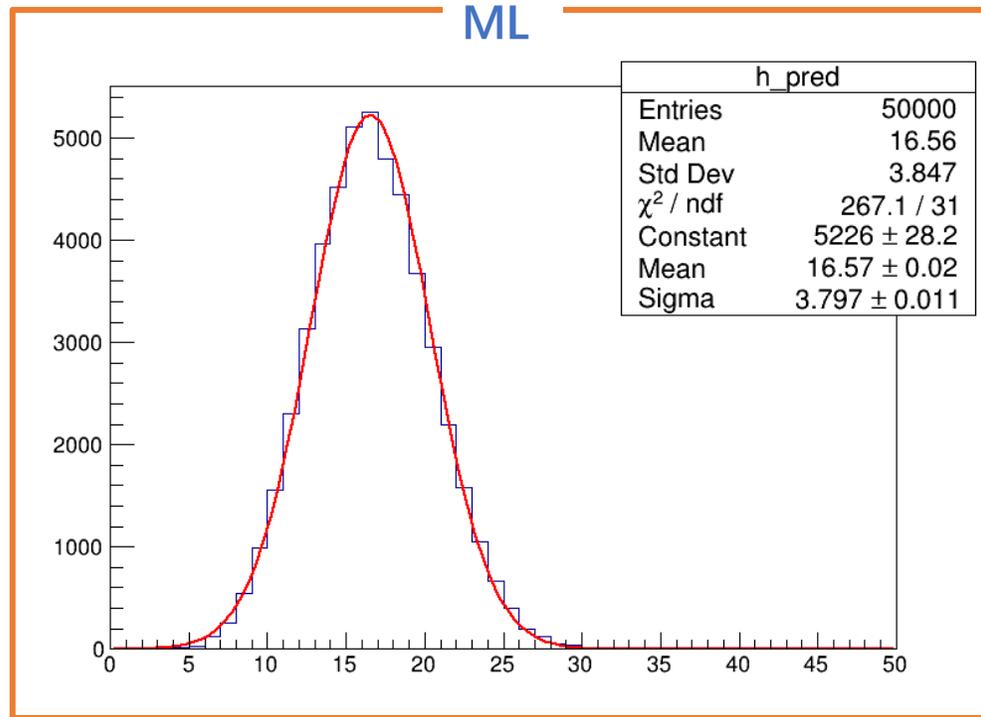
Step 2: Discrimination of the primaries

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

Detected peaks from the peak finding

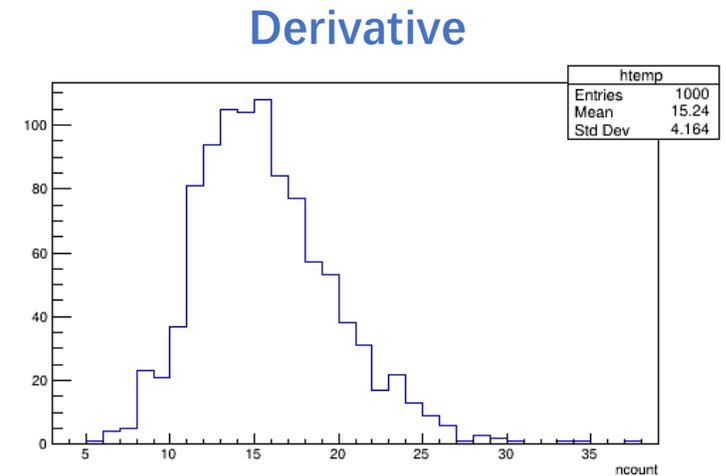


N_{cls} distributions



Resolution $\sim 23\%$

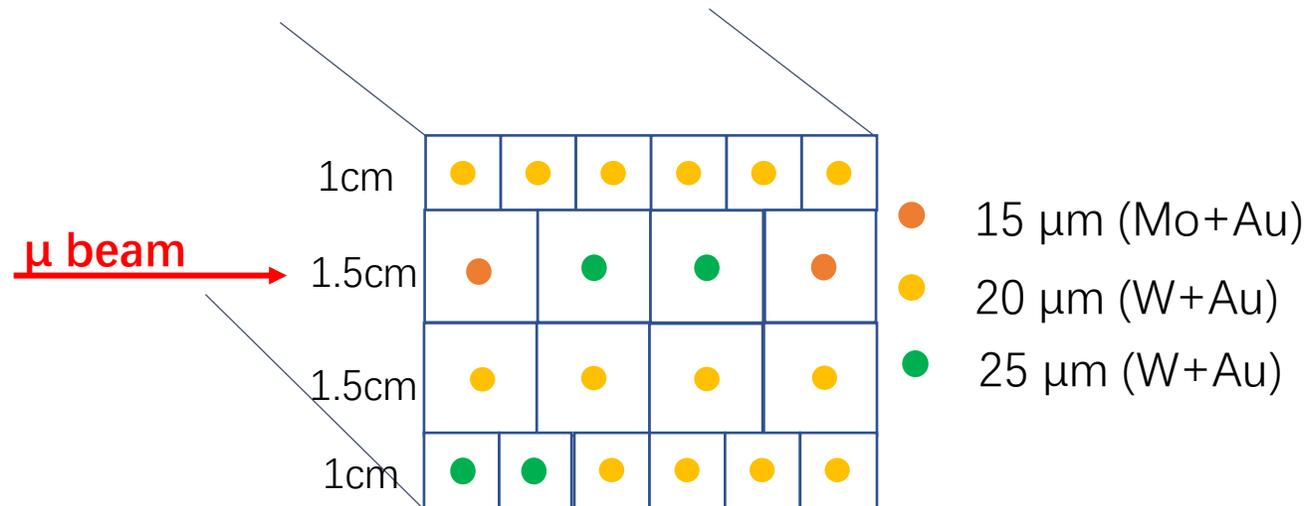
- ✓ Very good Gaussian-like distribution
- ✓ The resolution is very close to the truth value ($\sim 21\%$), which implies possible improvement on PID



✓ RMS/MEAN $\sim 27\%$

Beam test

- Two rounds of beam test organized by INFN group
 - First round: Nov. 2021 (Shuiting joined)
 - Second round: July 2022 (Shuaiyi joined)
- Cooperation between INFN and IHEP on data analysis is ongoing

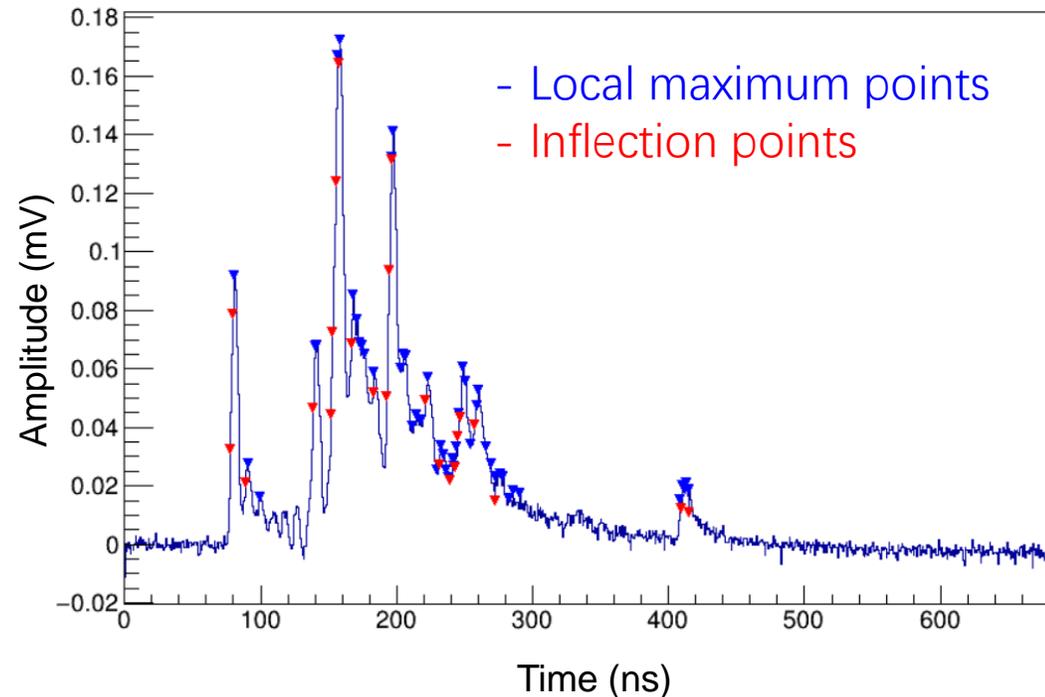


2nd round of test beam:

Federica's talk in Meeting on cluster counting in drift chambers, Jul 28, 2022

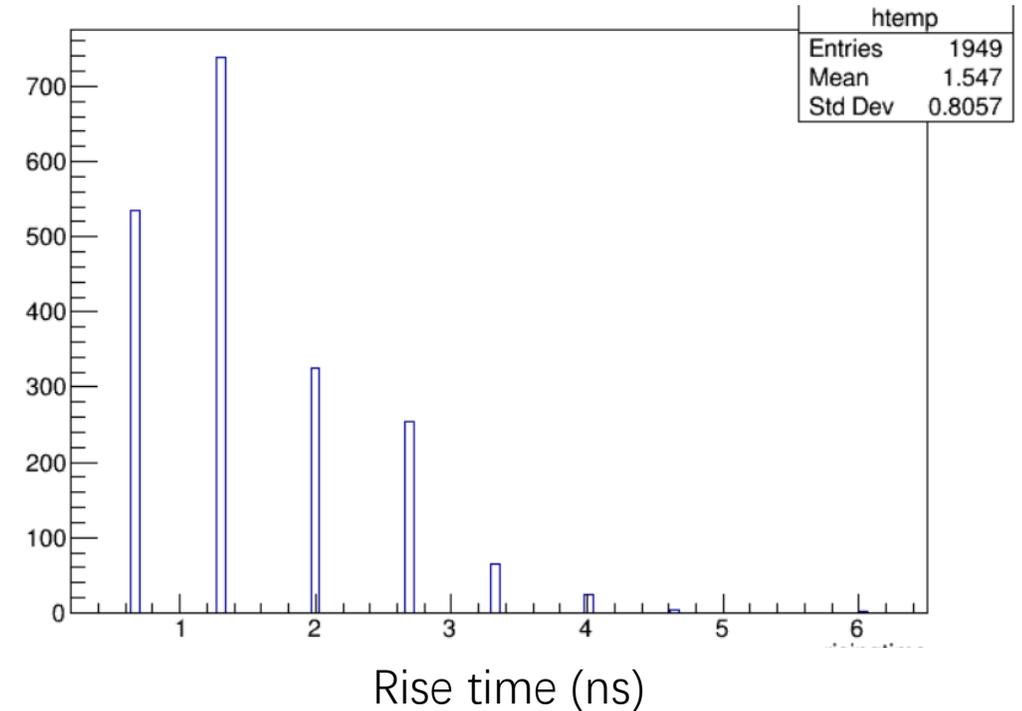
<https://indico.ihep.ac.cn/event/16837/>

Peak risetime estimation



- Require local maximum $>0.015\text{mV}$
- For each local maximum points, find the nearest early point, whose slope starts to decrease (inflection point)
- The time distance between the local maximum and the inflection points is recorded as the risetime.

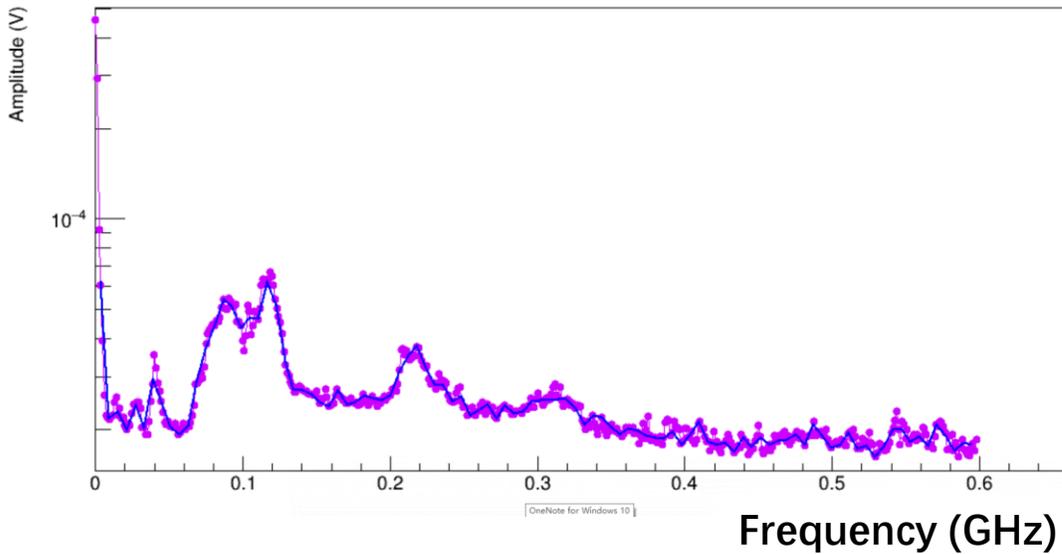
Risetime distribution



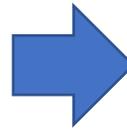
- Each bin is around 0.67 ns
- Averaged risetime is **1.55 ns**
- 95% of the risetimes less than 2.67 ns

Noise extraction

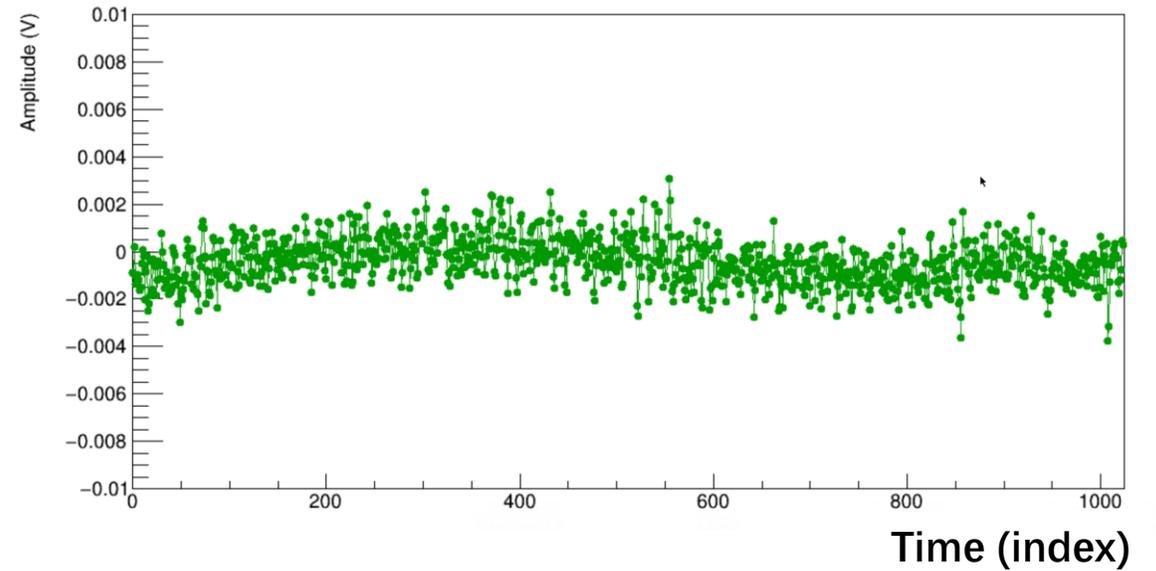
Data (frequency domain)



IFFT



Simulation (time domain)

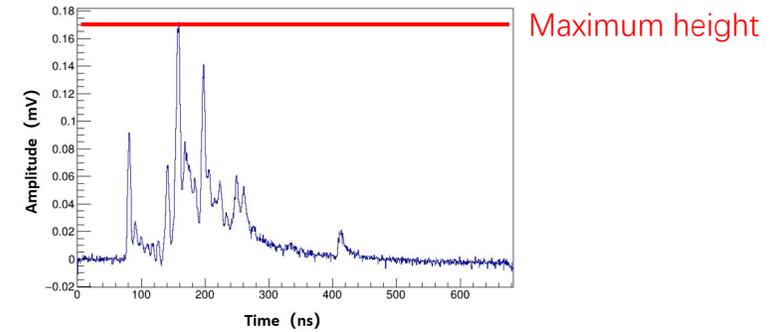


Noise distribution

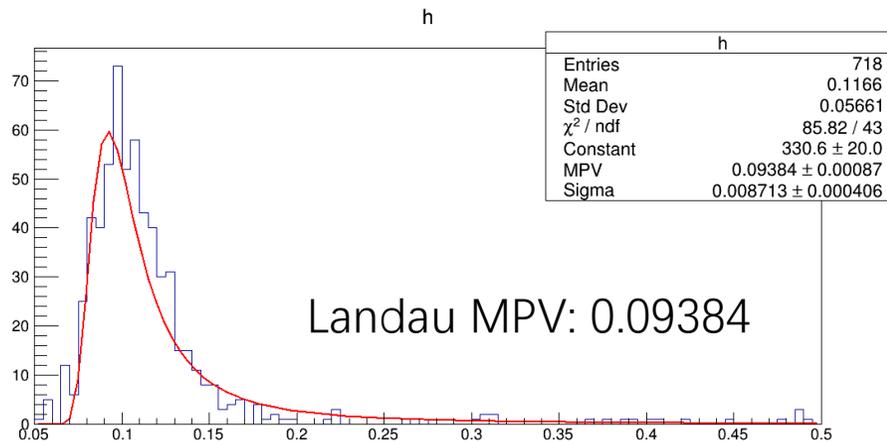
	Data	Simulation
Mean	0.00019	0.00058
RMS	0.00125	0.00125

Signal amplitude scale

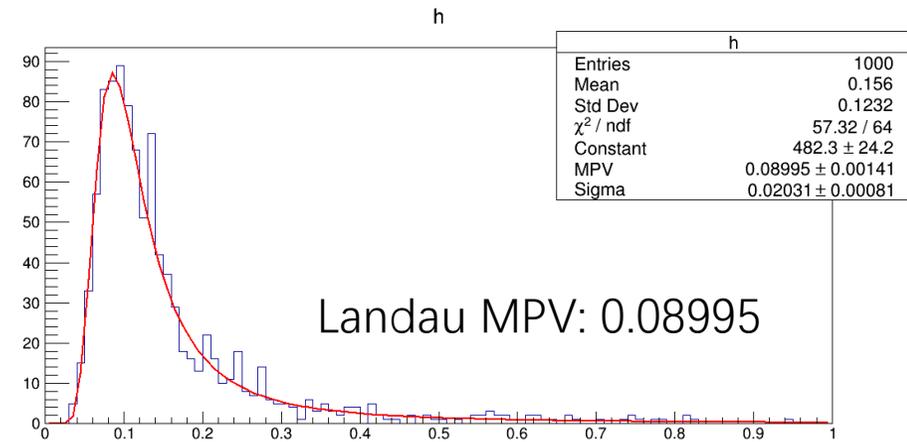
- Scale the amplitude of MC simulation to data



Average max. height in Data

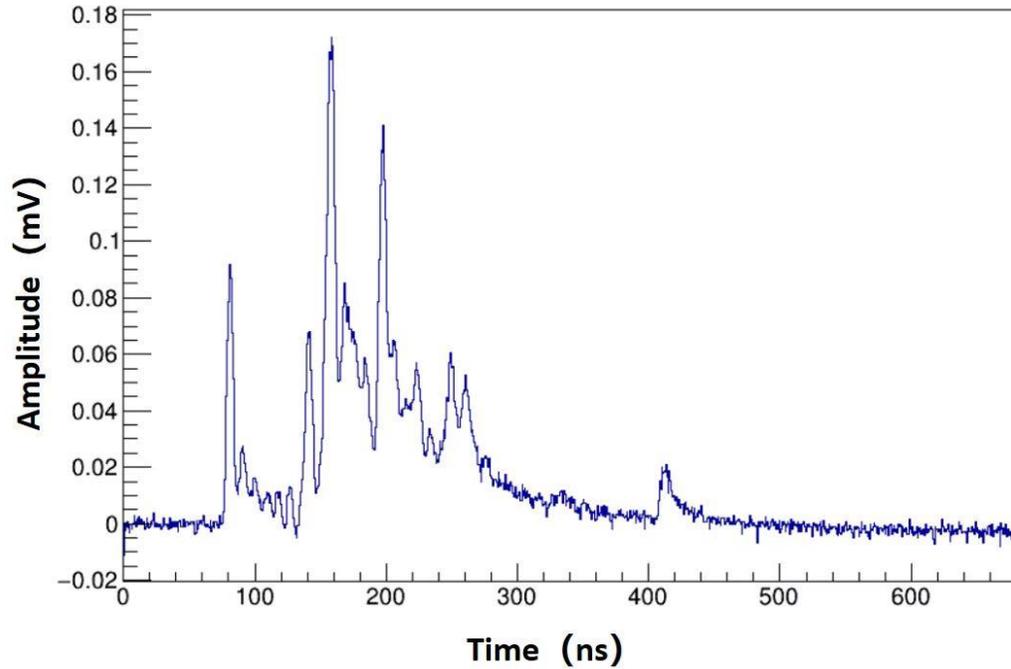


Average max. height in scaled MC

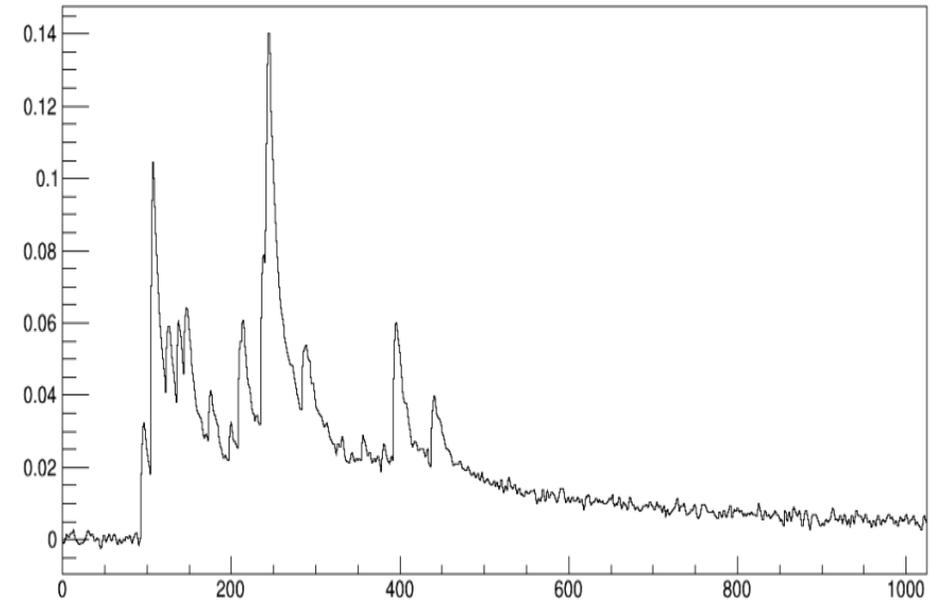


MC tuning with data

Data

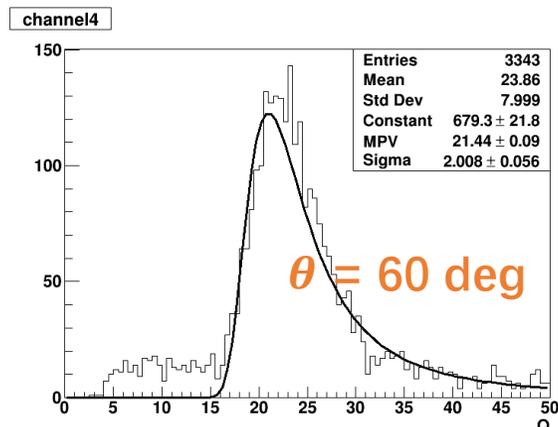
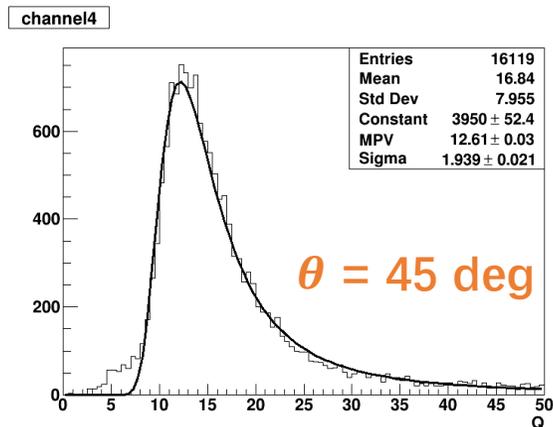
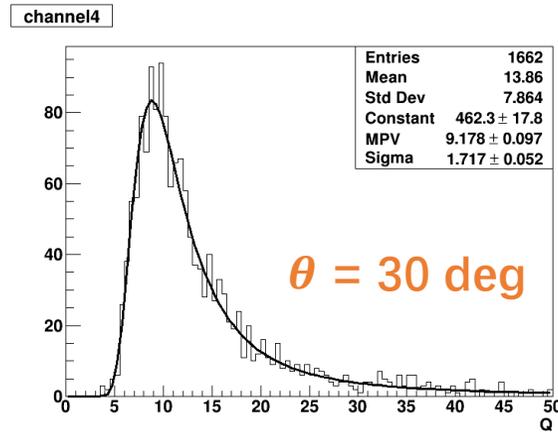
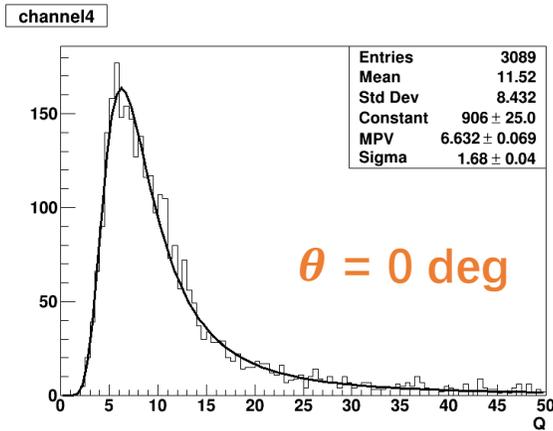
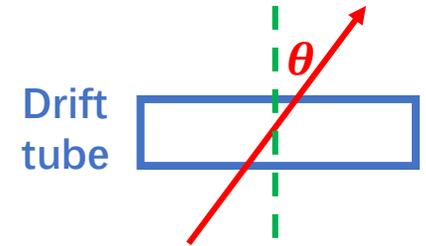


Simulation

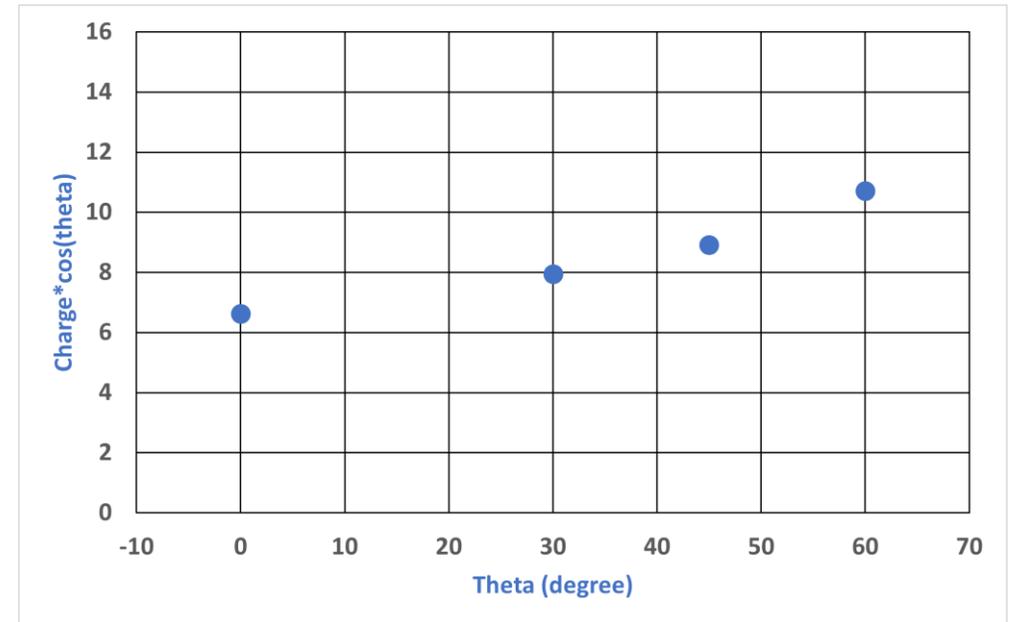


Tuning MC with data: (1) Rise-time, (2) Noise, (3) Amplitude
➔ Better data/MC consistency

Charge distribution checks



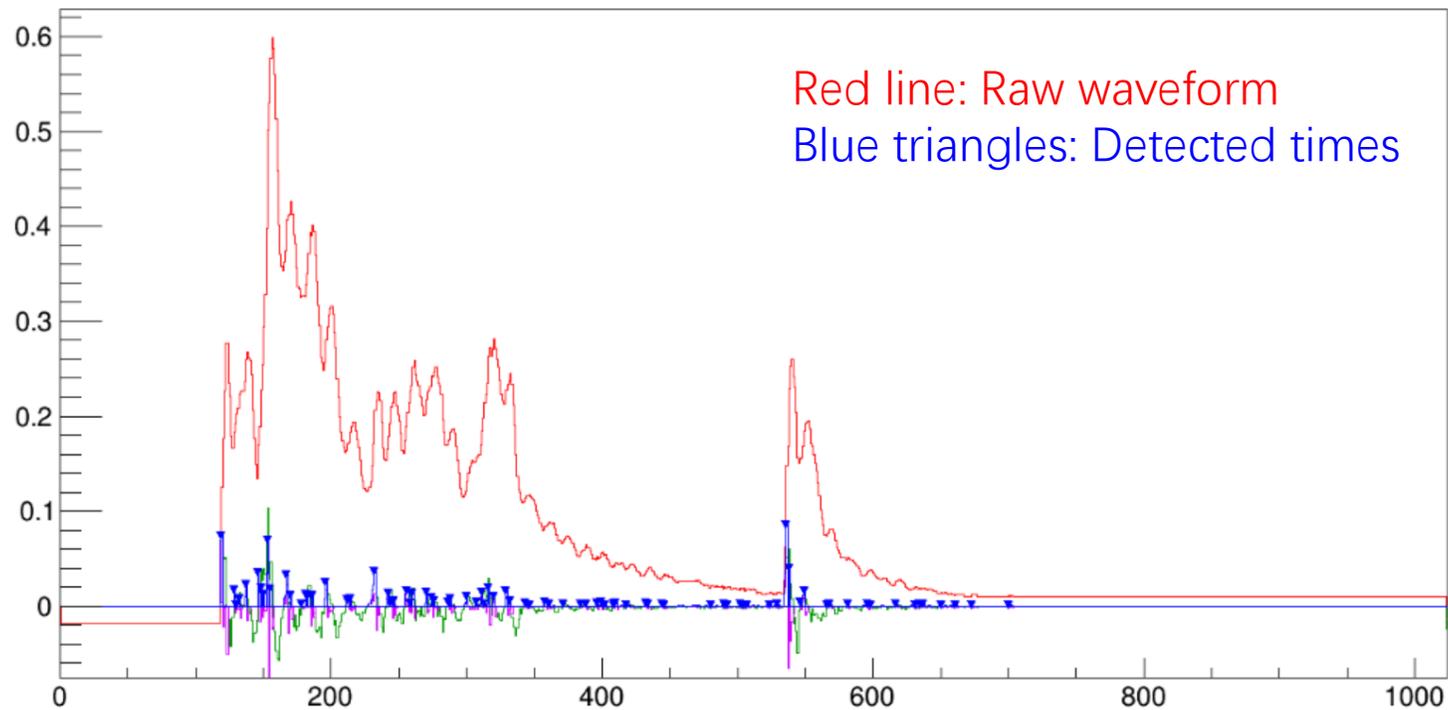
Charge MPV scaled by track length



→ Variation due to space charge

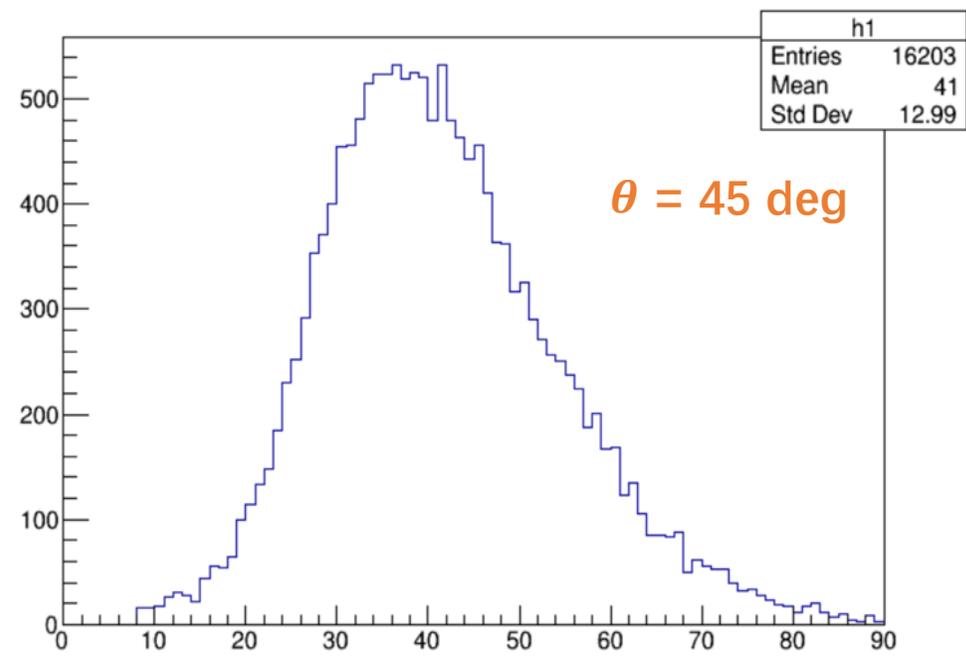
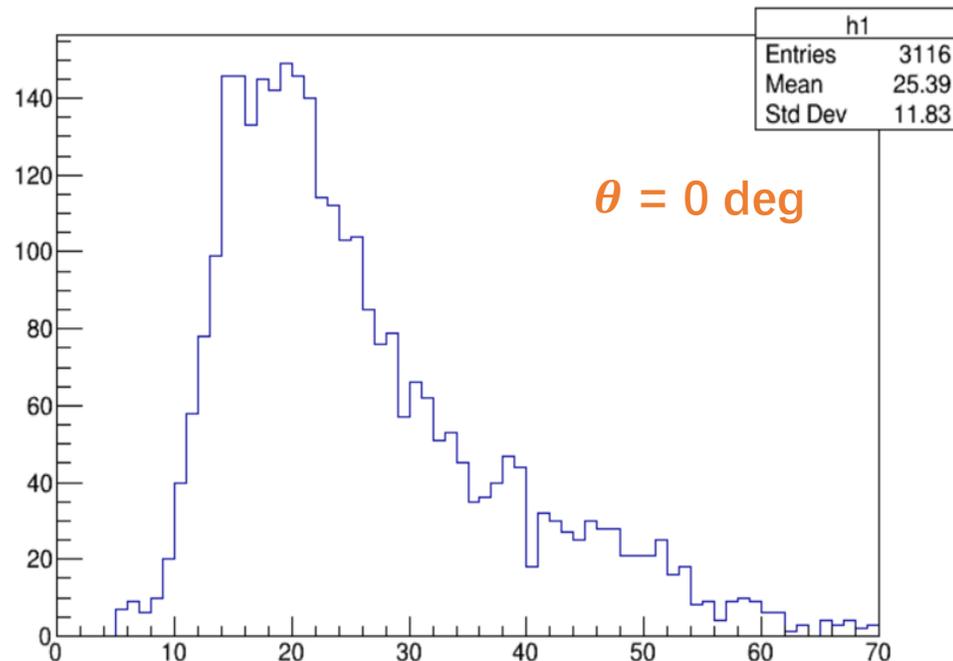
Preliminary peak finding with derivatives

- Apply the algorithm based on 1st and 2nd derivatives on data
- Demonstrated by an example from the 2nd round test beam data



N_{cls} distributions

- A very preliminary comparison between $\theta = 0^\circ$ & 45°
- Next to do
 - Optimize the event selection
 - Optimize the parameters of the peak finding algorithm
 - Apply a discrimination algorithm to remove secondary peaks
 - Try ML algorithm



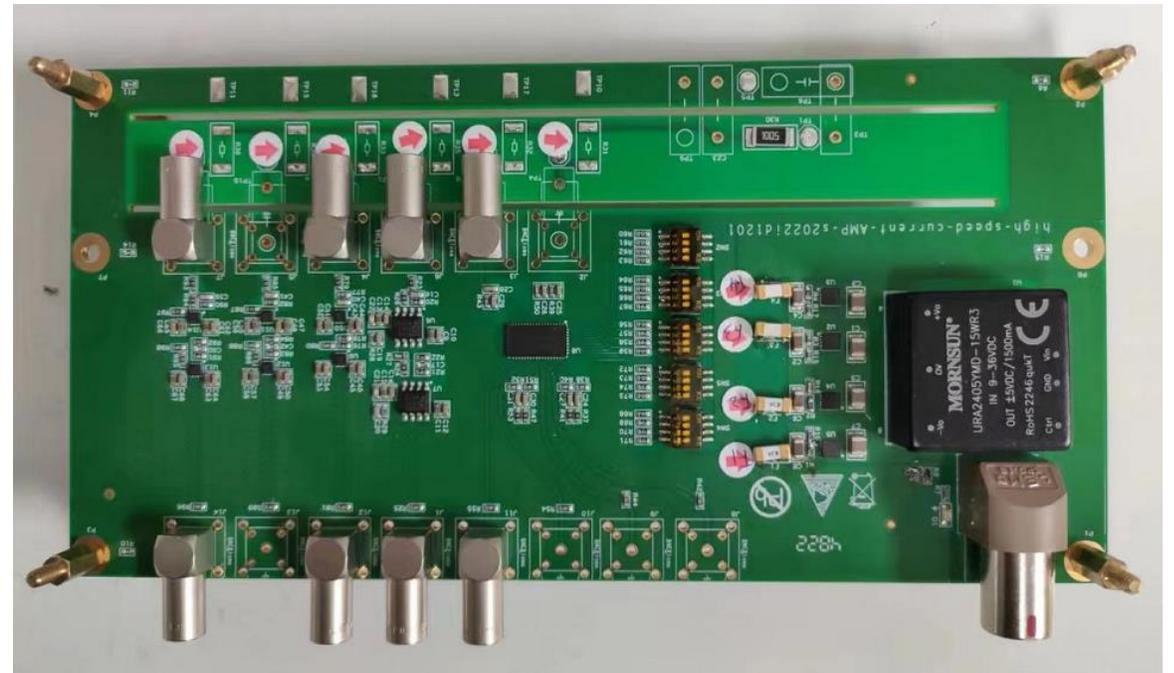
Collaborations on MOST-MAECI application

(China-Italy Science and Technology Cooperation)

- **Project: INTRePID (INnovative TRacking and Particle IDentification)**
 - Cluster counting and cluster timing for particle identification and improved spatial resolution in drift chambers at the next generation of lepton colliders
- **Cooperation between INFN and IHEP 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

Progress on hardware: new preamplifiers

- High bandwidth current sensitive preamplifiers based on LMH6522 and AD8099 were designed
- Simulation results show the bandwidth and the gain meet the test requirements
- Test is on going. Preliminary tests verify that the gain is big enough, but noise should be further reduced



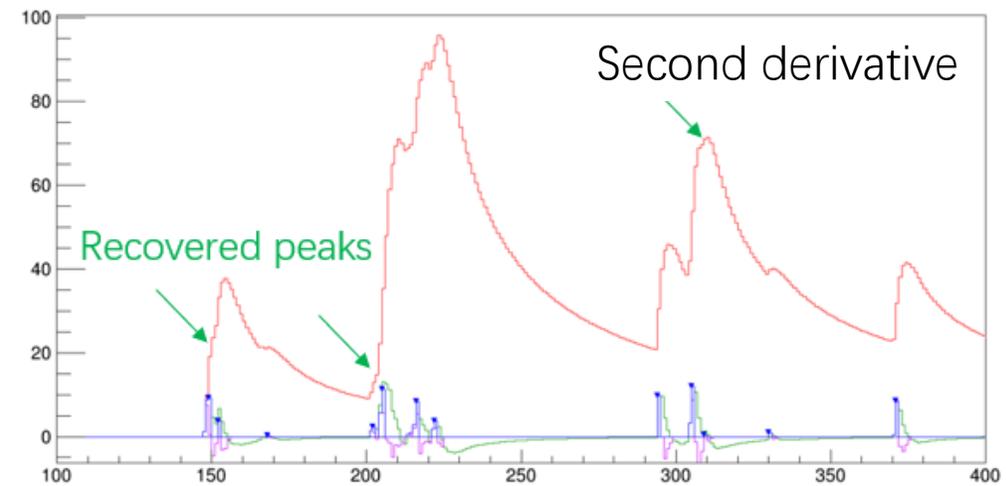
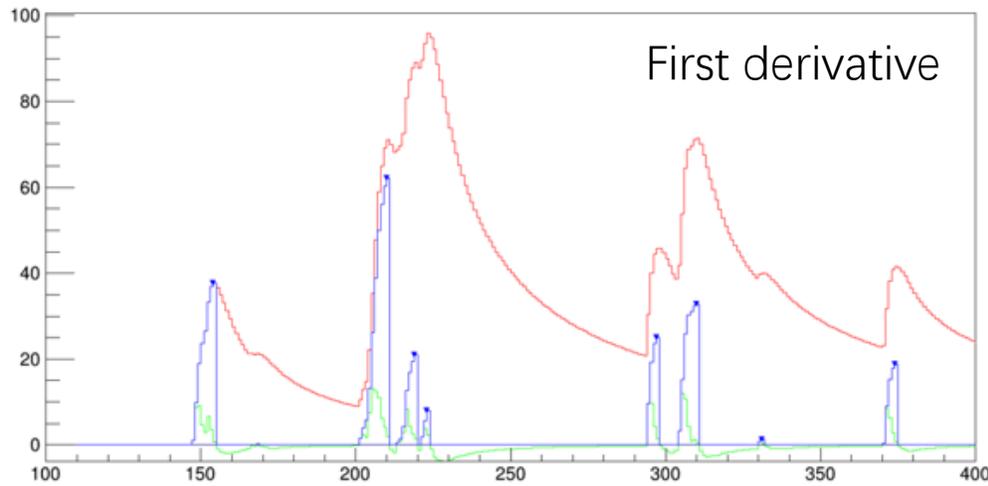
Summary and outlook

- **A machine learning based reconstruction algorithm is developed**
- **Simulation is tuned with the experimental data**
- **Plans**
 - Better understand the beam test data and fine tuning the MC simulation
 - Optimize the reconstruction algorithm for beam test
 - Extract dN/dx parameters from full simulation and perform physics studies

Backup

Reconstruction algorithm with derivatives

- Peak finding algorithm based on 1st and 2nd order derivatives
 - Fast and efficient
 - Good pile-up recovery ability on the rising edge



Pile-up on the falling edge is easier to recover.
However, it is not the case for pile-up on the rising edge.

Beam test objectives

Beam test plans:

1. First of all, need to demonstrate the **ability to count clusters**:
at a fixed $\beta\gamma$ (e.g. muons at a fixed momentum) count the clusters by
 - doubling and tripling the track length and changing the track angle;
 - changing the gas mixture.
2. Establish the **limiting parameters** for an efficient cluster counting:
 - **cluster density** (by changing the gas mixture)
 - **space charge** (by changing gas gain, sense wire diameter, track angle)
 - **gas gain saturation**
3. In optimal configuration, **measure the relativistic rise as a function of $\beta\gamma$** , both in **dE/dx** and in **dN_{cl}/dx** , by scanning the muon momentum from the lowest to the highest value (from a few GeV/c to about 250 GeV/c at CERN/H8).
4. Use the experimental results to fine tune the predictions on performance of **cluster counting** for **flavor physics** and for **jet flavor tagging** both in **DELPHES** and in **full simulation**

Space charge from BESIII dE/dx

