



# Pile-up events Discrimination Based on Machine Learning in JUNO

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## Outline



#### Introduction

- JUNO
- Pile-up Events Characteristics
- Impact of <sup>14</sup>C pile-up on Energy Resolution
- Dataset
- Discrimination Based on Traditional Method
- Discrimination Based on Machine Learning
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  - 1D Transformer
  - KamNet
- Summary



## Introduction: JUNO



• JUNO starts data collection in 2025, its primary physics goal being to determine the neutrino mass ordering

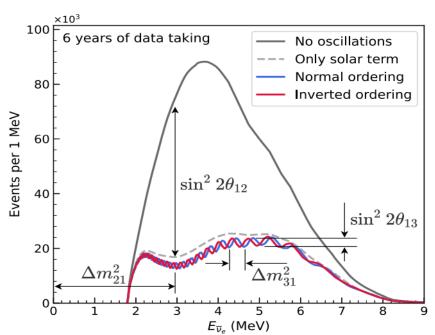
by measuring the reactor neutrino energy spectrum (IBD :  $\bar{\nu}_e + p \rightarrow e^+ + n$ )

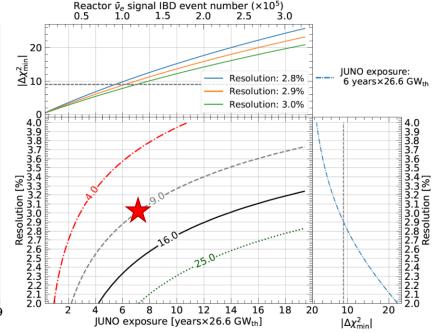
实验	KamLAND	Borexino	Daya Bay	JUNO
光阴极覆盖率	34%	34%	12%	78%
靶质量	1000 ton	300 ton	20 ton	20000 ton
能量分辨率	6.5%	5%	8%	3%

JUNO is 53 km away from both the Taishan and Yangjiang NPP

• Compared with other liquid scintillator (LS) detectors, JUNO has the **best energy resolution 3%@1MeV**, the

#### highest photocathode coverage, and the largest target volume









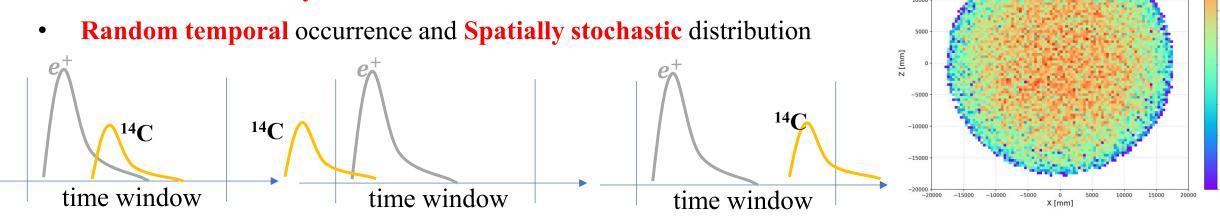


## Introduction: Pile-up Events Characteristics

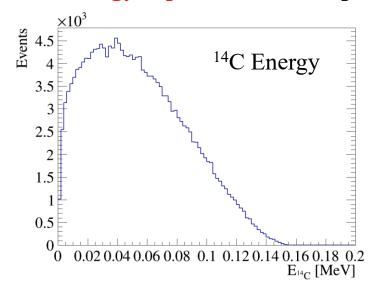


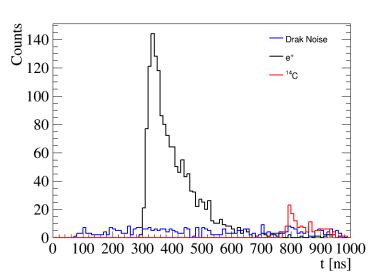
<sup>14</sup>C vertices

- $^{14}$ C  $\beta$ -decay pile up with e<sup>+</sup> signal in LS, forming pile-up events
  - 14C and e<sup>+</sup> are entirely uncorrelated



• Distinct energy separation:  $e^+$  deposit energy  $\geq 1.022$  MeV,  $^{14}$ C  $\beta$ -decay yields energy < 0.16 MeV



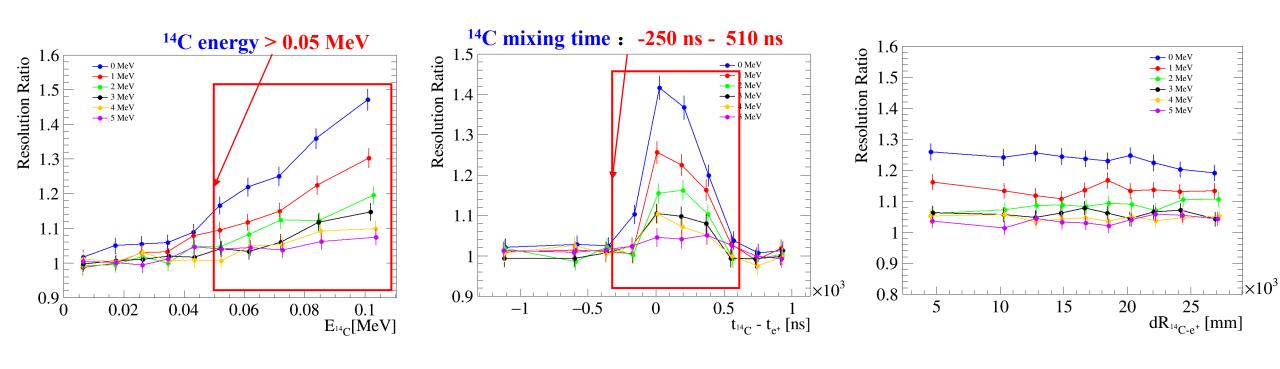




## Introduction: Impact of <sup>14</sup>C pile-up on Energy Resolution



• Considering <sup>14</sup>C Energy, Mixing time and Distance with e<sup>+</sup>



- About  $1\% \sim 2\%$  <sup>14</sup>C significantly affecting energy resolution
- Make discriminate e<sup>+</sup> and pile-up events be a significant challenges



## Dataset

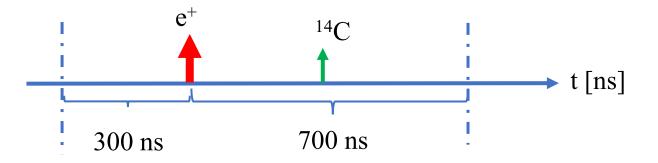


#### • Training Sample:

Training Sample	$t_{^{14}C}-t_{e}^{^+}[\mathrm{ns}]$	Position	Energy [MeV]	Number [ k ]
$e^+$		uniform	$0 \sim 5$	400
Pile-up in window	[0, 200)	uniform	$0 \sim 5$	400
Pile-up out window	$[-300, 0) \cup [200, 700)$	uniform	0 ~ 5	400

#### • Test Sample:

- (0, 1, 2, 3, 4, 5) MeV e<sup>+</sup> or pile-up events : 500 k events in total / MeV
- <sup>14</sup>C decay rate is set to 40 kHz





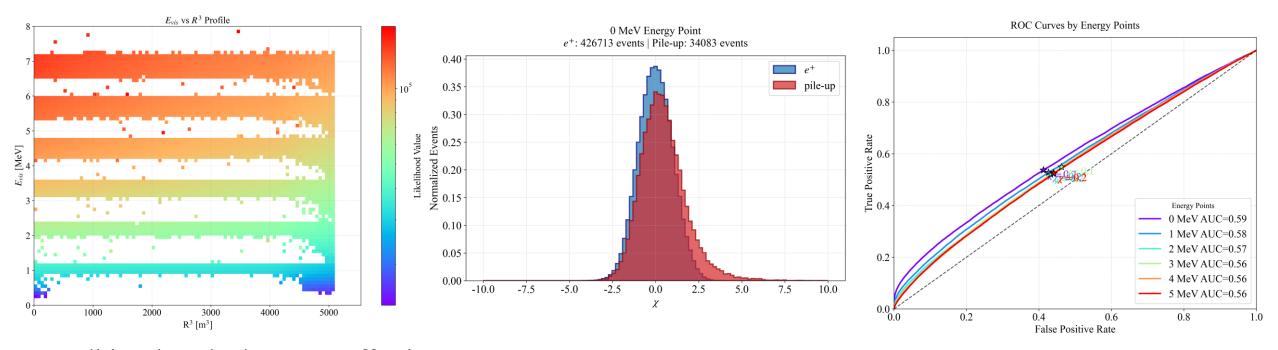
### Discrimination Based on Traditional Method



- JUNO reconstruction algorithm based on one event hypothesis, likelihood will be calculated during reconstruction
- We define a

$$\chi = \frac{L_{event} - L_{mean}(E_{vis}, R^3)}{L_{std}(E_{vis}, R^3)}$$

Pile-up events χ is larger than pure e<sup>+</sup>



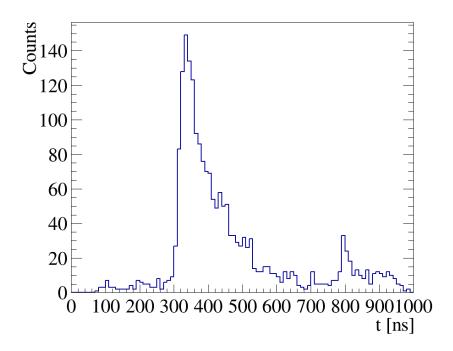
Traditional methods are not effective.

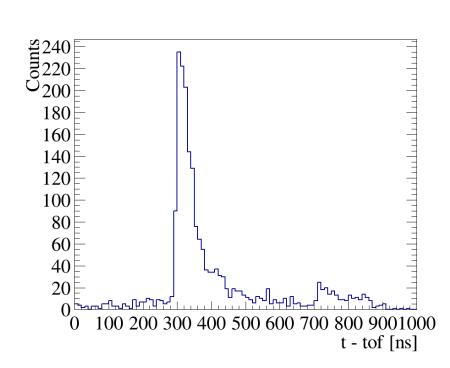


## Discrimination Based on Machine Learning: 1D CNN



- Convolutional neural network
- Input
  - hit time
  - hit time tof (Time of Flight)
- Output : probability of 2 categories





Input (2×1500)

Conv1d(2, 4, 3)

BatchNorm1d(4)

ReLU()+MaxPool1d(2,2)

Conv1d(4, 16, 3)

BatchNorm1d(16)

ReLU()+MaxPool1d(2,2)

Conv1d(16, 32, 3)

BatchNorm1d(32)

ReLU()+MaxPool1d(2,2)

Conv1d(32, 64, 3)

BatchNorm1d(32)+ReLU()

Conv1d(64, 64, 3)

BatchNorm1d(32)+ReLU()

MaxPool1d(2,2)

FC-1024

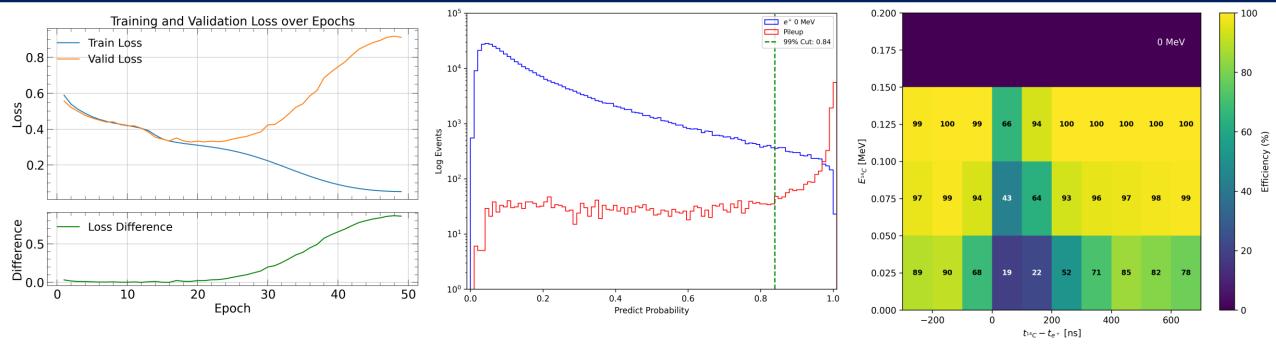
FC-128

FC-2



## Discrimination Based on Machine Learning: 1D CNN





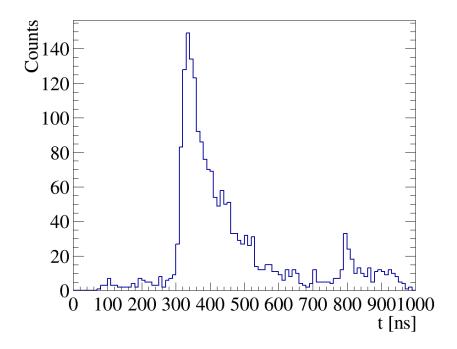
- The most difficult discrimination region is when the mixing time is between 0-200 ns.
- $^{14}$ C deposit energy > 0.05 MeV and mixing time between 0 200 ns, the accuracy of CNN can reach 43% 94%.

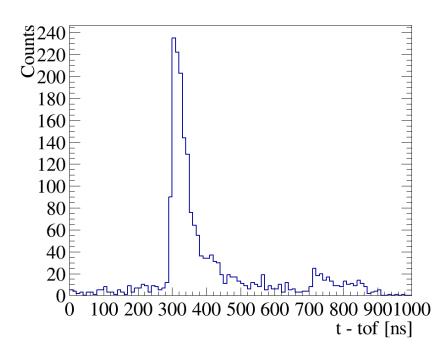


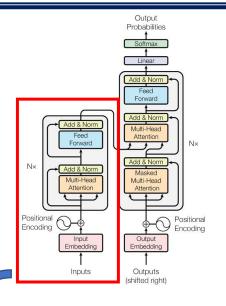
## Discrimination Based on Machine Learning: 1D Transformer



- Transformer
- Input
  - hit time
  - hit time tof (Time of Flight)
- Output : probability of 2 categories







Input (2×1500)
Conv1d(2, 32, 1)
BatchNorm1d(32)
PositionalEncoding(32)

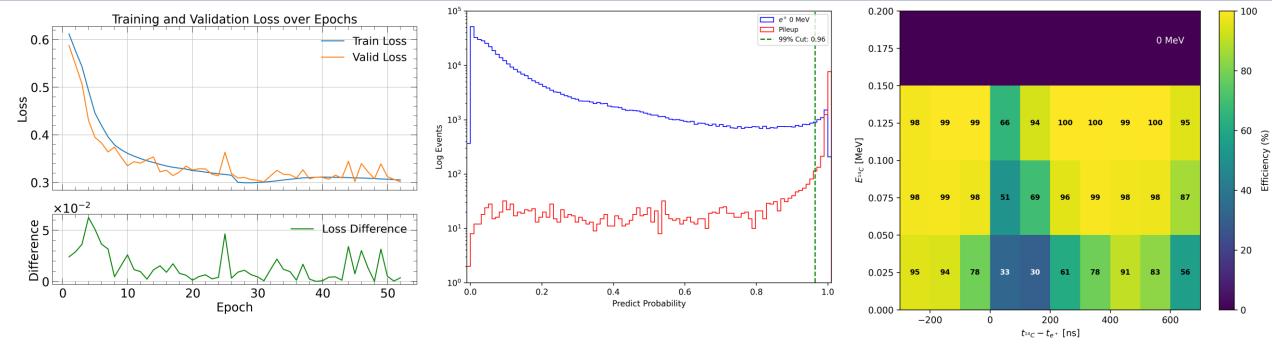
2\*TransformerEncoderLayer
(32,8,2048,0.1)
torch.mean(dim=0)

FC-1024
ReLU()
FC-128
ReLU()
FC-2



# Discrimination Based on Machine Learning: 1D Transformer





- $^{14}$ C deposit energy > 0.05 MeV and mixing time between 0 200 ns, the accuracy of Transformer can reach 51% 94%.
- Transformer's performance is **better than CNN**

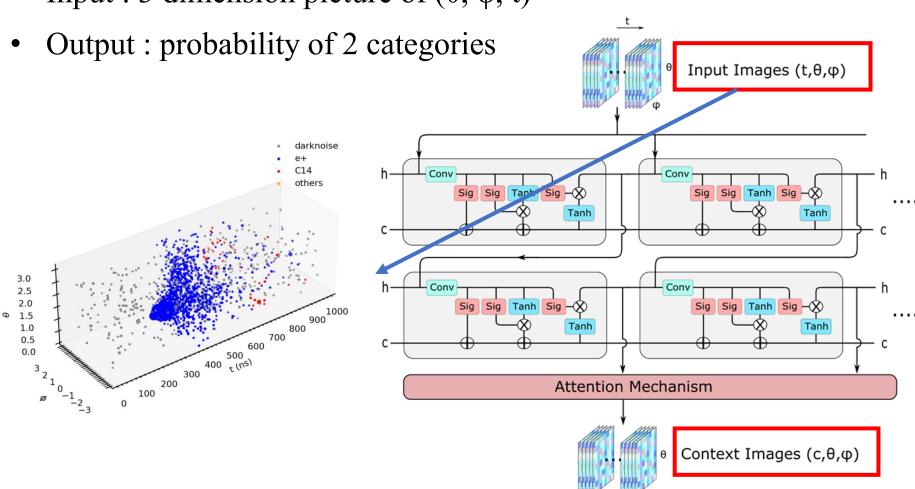


## Discrimination Based on Machine Learning: KamNet

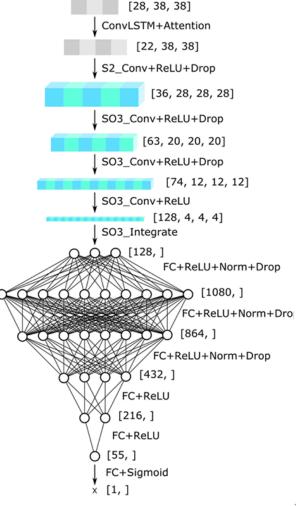


• KamNet : Sphere CNN + ConvLSTM + Attention

• Input : 3 dimension picture of  $(\theta, \varphi, t)$ 



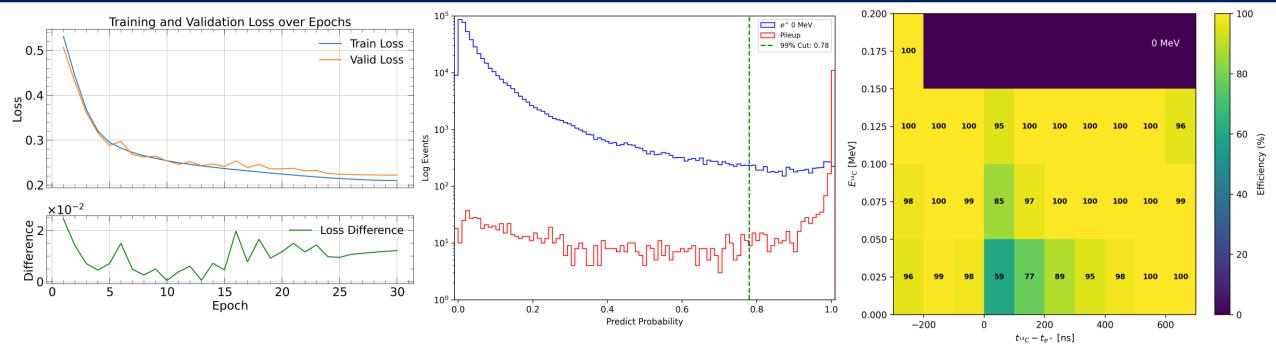
A.Li et al, Phys.Rev.C, 2023





## Discrimination Based on Machine Learning: KamNet





- $^{14}$ C deposit energy > 0.05 MeV and mixing time between 0 200 ns, the accuracy of KamNet can reach 85% 100%.
- KamNet's performance is best in these three models



## Summary



- $^{14}$ C deposit energy > 0.05 MeV and mixing time between 0 ~ 200 ns is critical for events reconstruction and discrimination of pile-up events
- CNN reach 43% 94% accuracy in the critical region
- Transformer reach 51% 94% accuracy in the critical region
- KamNet reach 85% 100% accuracy in the critical region, with the best performance
- Draft article has been completed.