#### Photon Response Model of MCP-PMT

- based on the onsite PMT testing data

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#### **Outline**

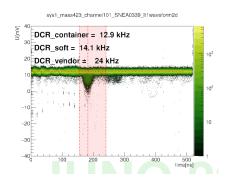
Brief Introduction

2 traing and test of CNN

Summary

# the "big signals" of MCP PMT

The typical waveforms <sup>1</sup> of MCP PMT, compared with dynode PMT.



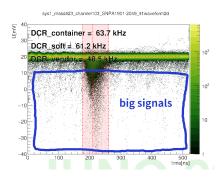


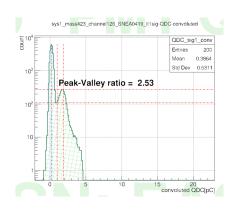
图: waveforms of HAMAMATSU PMT

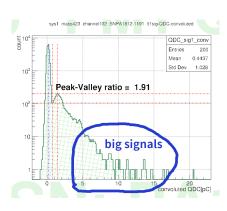
■: waveforms of MCP PMT

<sup>&</sup>lt;sup>1</sup>gain= 1E7, $\mu \simeq 0.1$ 

## the "big signals" of MCP PMT

The "long tail" in charge spectrum<sup>2</sup> of MCP PMT, compared with dynode PMT.





SPE of HAMAMATSU PMT

SPE of MCP PMT

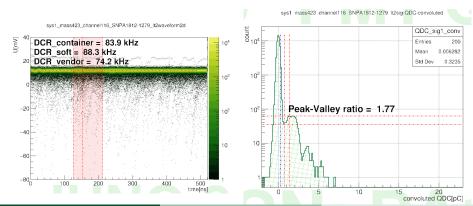
 $<sup>^{2}</sup>$ gain= 1E7, $\mu \simeq 0.1$ 

Based on the container testing data, we can acquire wavefroms of the MCP PMT in 5 different illumination levels:

- 1 dark noise mode [no light incident]
- 2 non-trigger window @1 p.e
- $3 \mu \simeq 0.1 \text{ p.e}$
- $\bullet$   $\mu \simeq 1$ p.e
- **5**  $\mu = \text{multi-p.e} [\text{by laser}]$

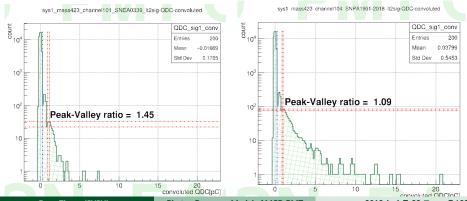
case 1:[dark noise]

The typical waveform and charge spectrum of MCP PMT@gain = $10^7$ . If we suppose all the dark counts is caused by single thermal electron, then those fake multi-p.e events are caused by the magnification of MCP.

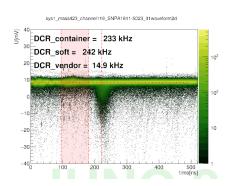


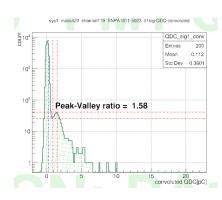
case 1:[dark noise]

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select the time interval before "trigger window",we can see similar QDC spectrum with dark noise case.

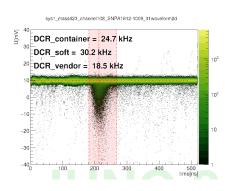


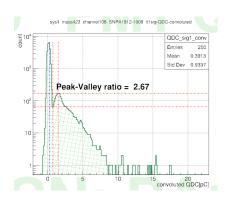


#### photon response characters of MCP PMT @ $\mu \simeq 0.1$

case 3:  $\mu \simeq 0.1$ 

In the trigger window, we can still see those "big signals" with charge >3p.e.





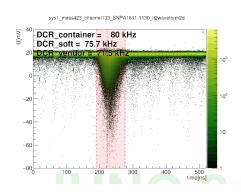
**\blacksquare**: waeforms @  $\mu \simeq 0.1$ 

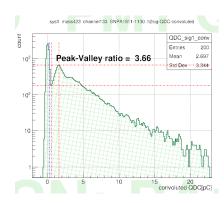
**3:** QDC @  $\mu \simeq 0.1$ 

## photon response characters of MCP PMT @ $\mu \simeq 1$

case 4:  $\mu \simeq 1$ 

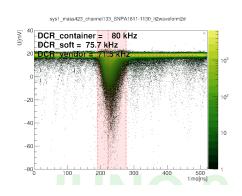
When the light intensity increase to  $\mu \simeq 1$  we can see a continuous "long tail" with charge >5p.e; this is a clear clue that MCP will magnify little part of electrons with abnornal large gain.

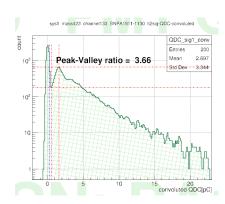




case 5:  $\mu > 2p.e$ .

When the light intensity increase to  $\mu \simeq 1$  we can see a continuous "long tail" with charge >5p.e; this is a clear clue that MCP will magnify little part of electrons with abnornal large gain.





#### number

To conclude the above information, we find:

• The Gain of MCP-PMT is not stable enough, It has small propobilty to magnify single p.e to unreasonable large charge.

#### the expected photon number

If we do a "cut" is the charge spectrum@0.25 spe, the averager photon number  $\mu$  can be acquired by<sup>3</sup>

$$\mu = -\ln(\frac{N_0}{N})\tag{1}$$

where  $N_0$  is the number of pedestal (0 p.e) events, N is the total event number.

However, if we know explicitly the photon number of specific event, the  $\mu$  value is :

$$\mu = 1 \times \mathbf{n}_1 + 2 \times \mathbf{n}_2 + \dots + \mathbf{N} \times \mathbf{n}_{\mathbf{N}} \tag{2}$$

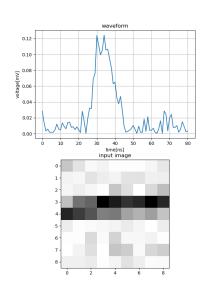
where  $n_N$  is the number of N p.e events.

<sup>&</sup>lt;sup>3</sup>E. H. Bellamy et al /Nucl. Instr. and Meth. m Phys . Res. A 339 (1994) 468-476

#### input of CNN

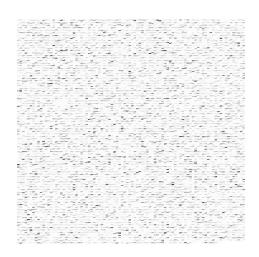
training data slecetion and pre-process:

- random selection from different PMTs
- 1.5<QDC<1.7 for 1p.e</li>
- 3.1<QDC<3.3 for 2p.e
- 4.7<QDC<4.9 for 3p.e</li>
- 81ns ROI  $\rightarrow$  9×9 2D image
- normalization



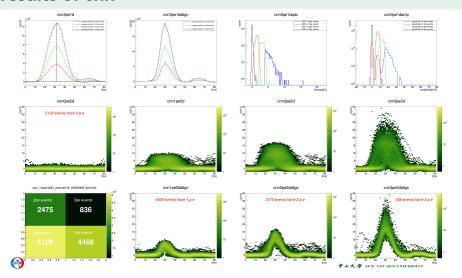
# **CNN** parameters

- 30k training waveform samples
- 2 convolution layers
- 4 output tags
- accuracy $\simeq 0.95$



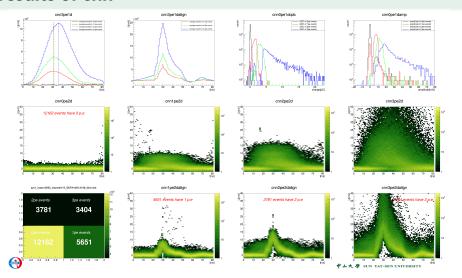
: input data

#### results of cnn



#### **图: HAMAMATSU PMT**

#### results of cnn



#### **II:** NNVT PMT

#### summary

- PSD by CNN provide a new option for PDE evaluation.
- can achieve sim0.95 acuuracy with the traditional method using simple NN.
- much faster than tradition methods in PDE evaluation.
- CNN can ectract more infromation from waveforms.

#### to list:

- refine the training samples and network structure.
- compare the accuracy in more details, for example using the reference tubes in container system.
- improve the input data quality.

# **THANKS**

# **BACK-UP**