

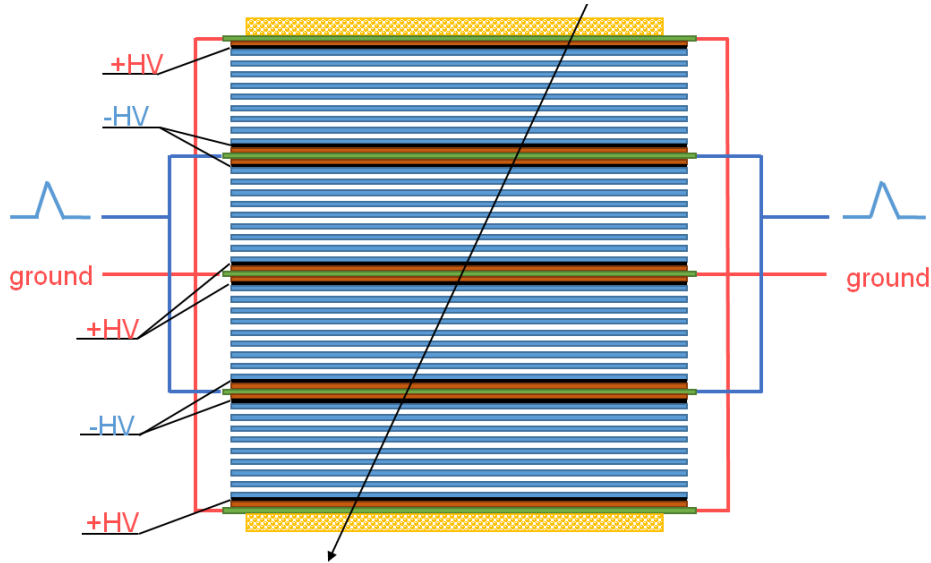
# Progress in Tsinghua

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# **Part I**

## **Next-Gen MRPC R&D**

# Introduction



Tsinghua Prototype, 2016

Gas gap width: **104um**

Number of gas gap: **32**

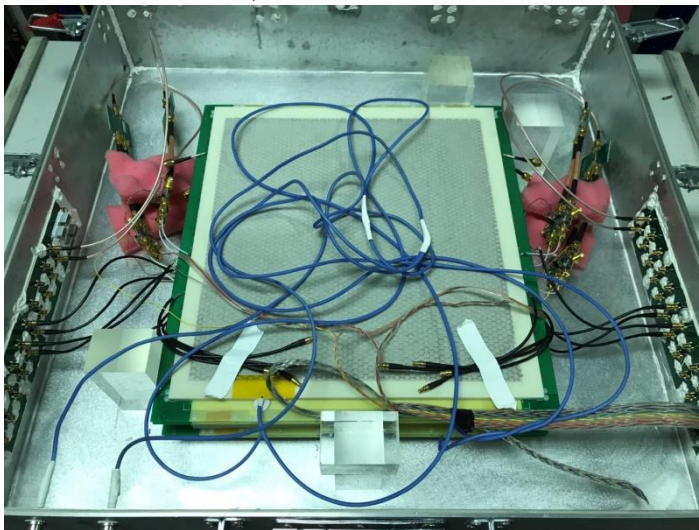
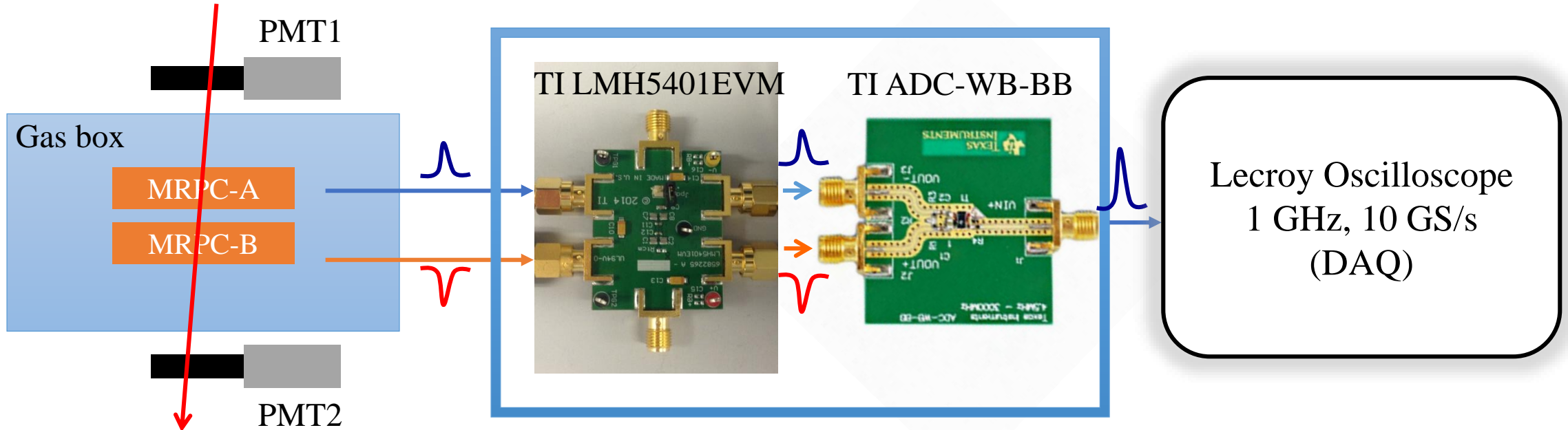
- **Narrower gap width** → fast charge dominant in the induced signal → Better timing resolution
- **More gas gaps** → Efficiency will be recovered

Tsinghua Prototype, 2018



	Single-stack MRPC
Gas Gap Width	<b>250 um</b> (fishing line)
Number of Gas Gaps	1 stack x 6 layers = <b>6</b>
Float Glass Thickness	<b>700 um</b>
Readout strip	<b>7 mm</b> x 270mm(3 mm internal)
Readout	differential, both ends

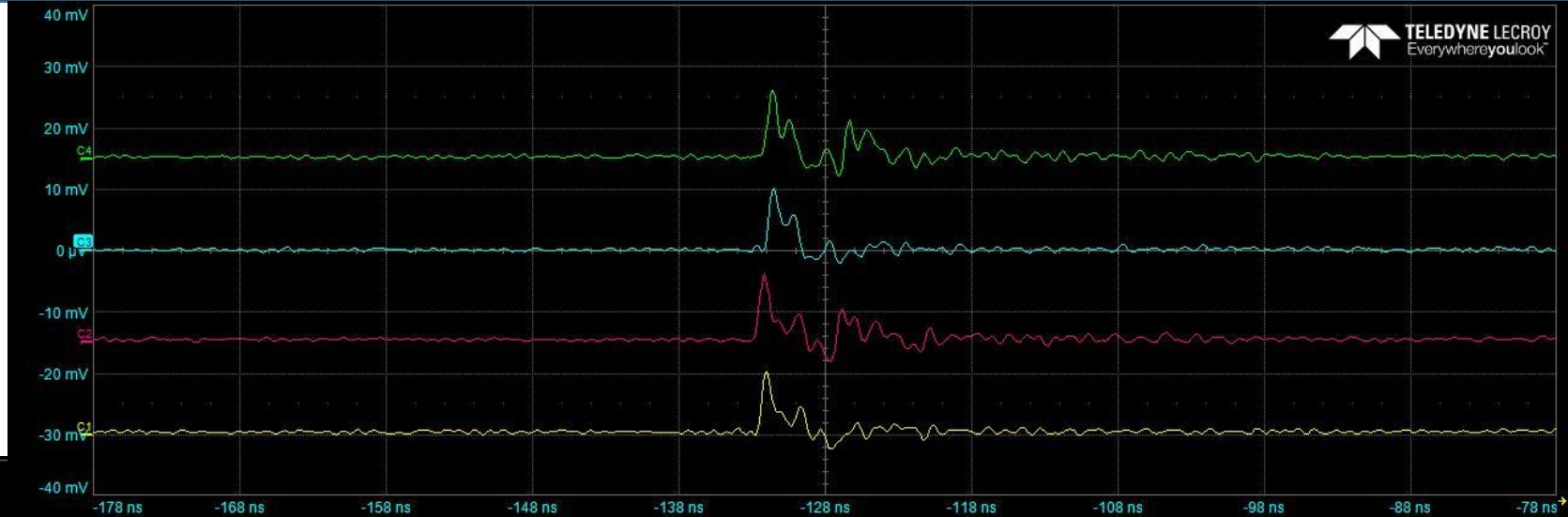
# Setup



# Setup

Channel Map

Ch2 — MRPC1 — Ch4  
Ch1 — MRPC2 — Ch3



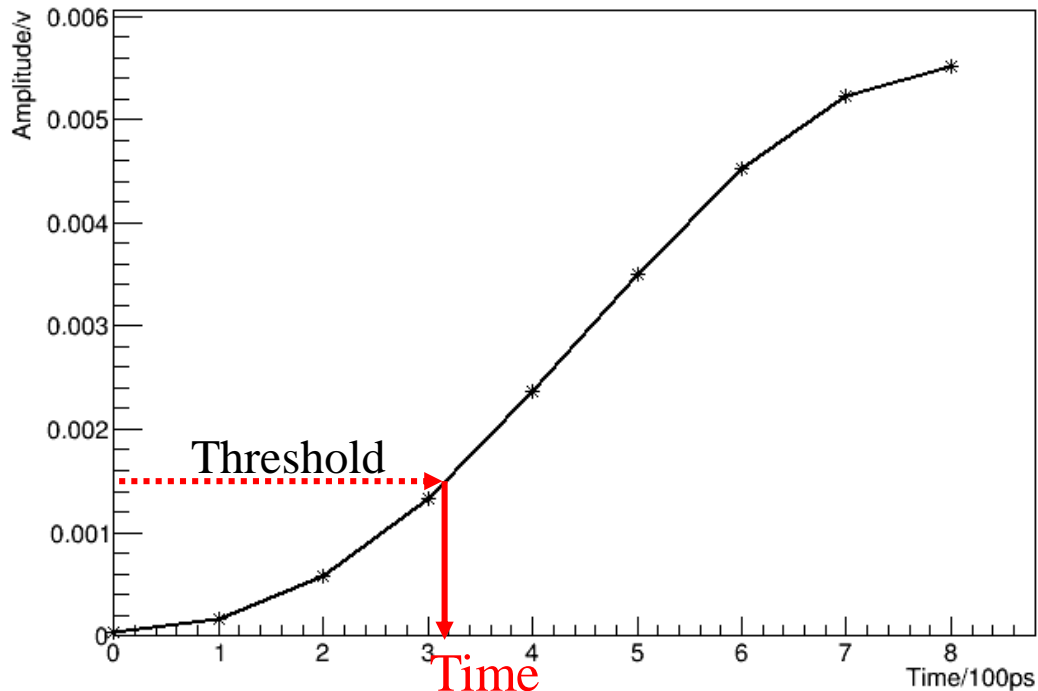
Sampling interval ~100ps  
Pedestal  $\sigma < 0.15\text{mV}$   
Rise time ~600ps

# Timing method

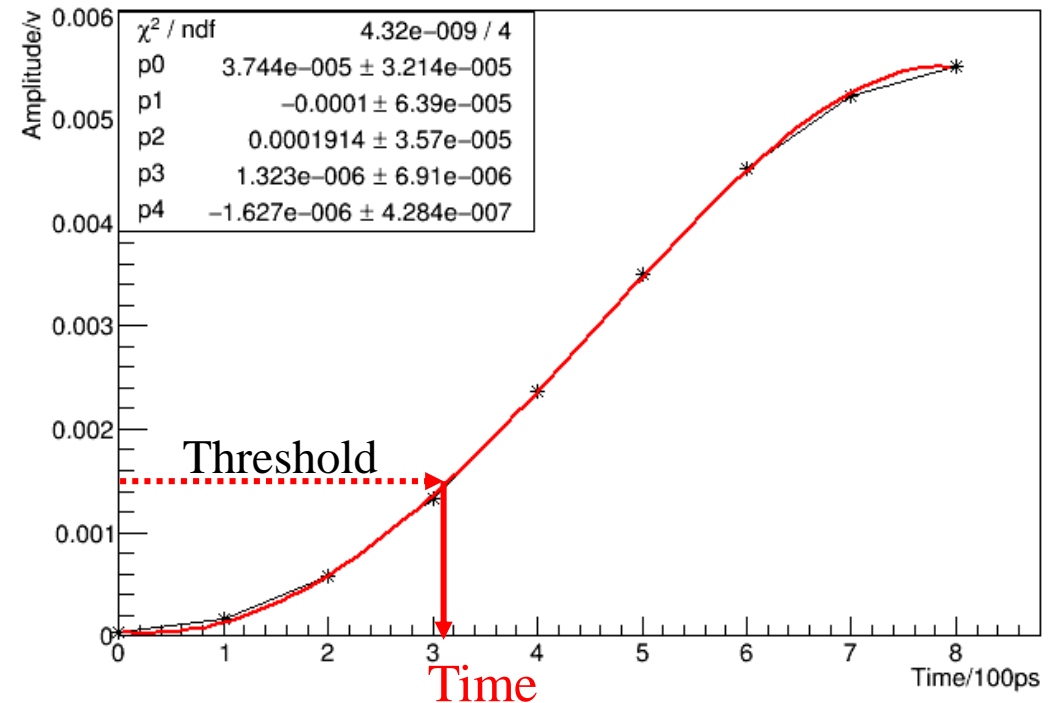
Algorithm to find a ref. time T (Fixed threshold)

- Linear interpolation
- Polynomial fit

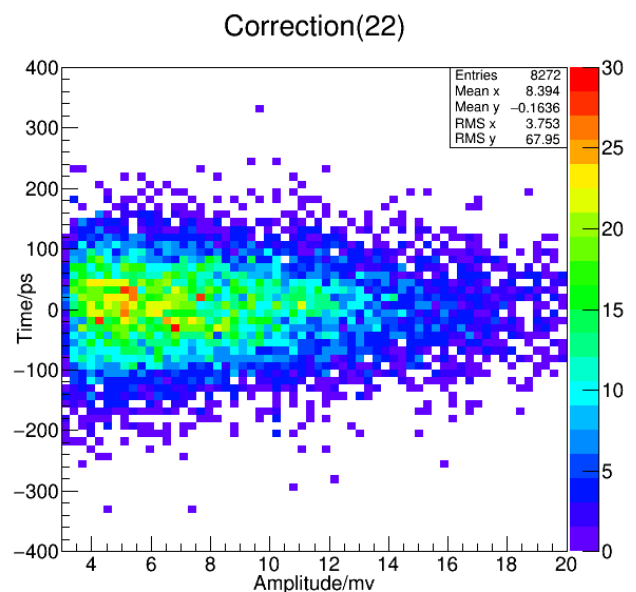
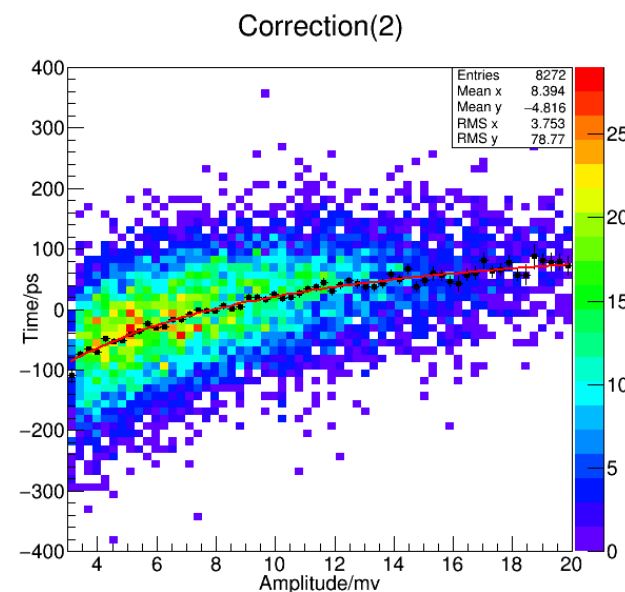
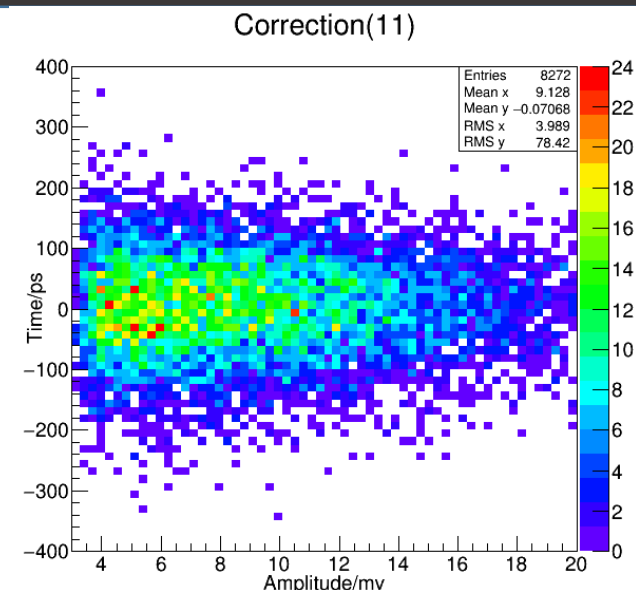
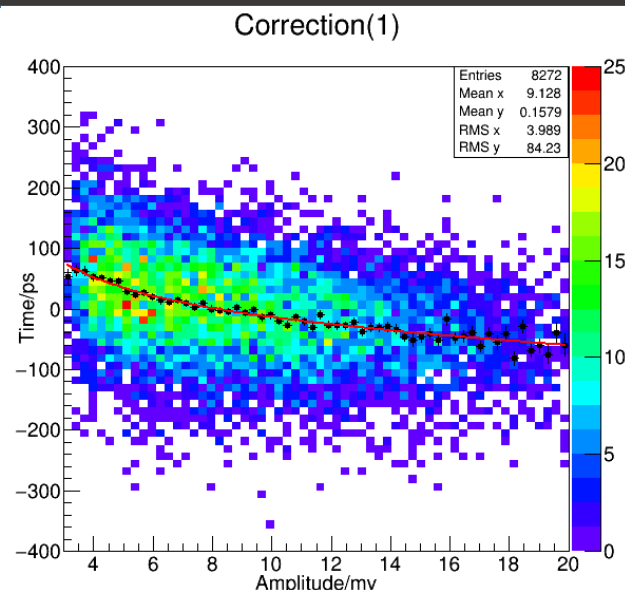
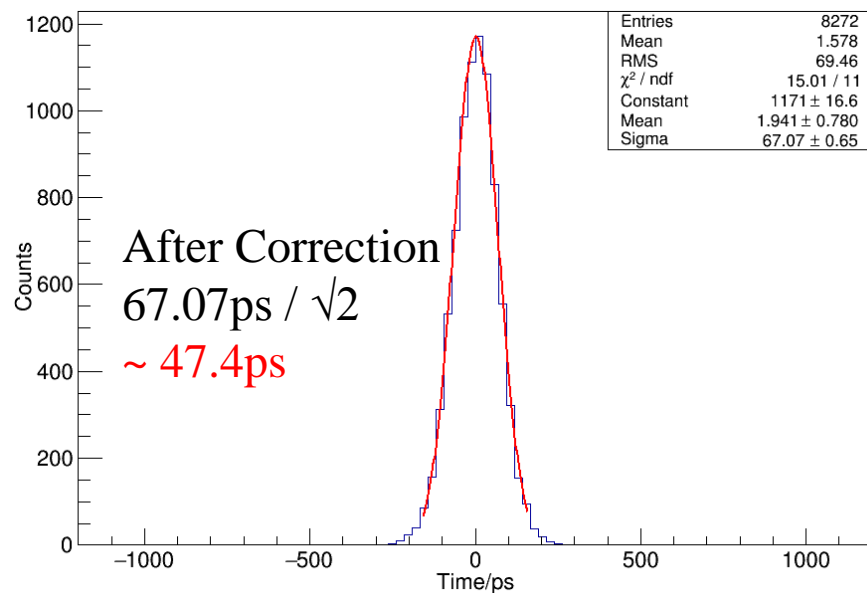
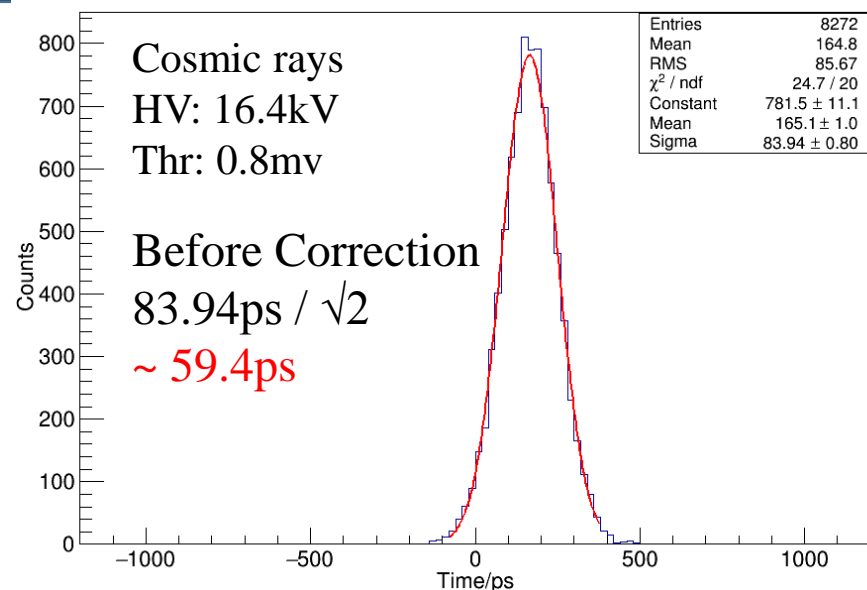
Graph



Graph



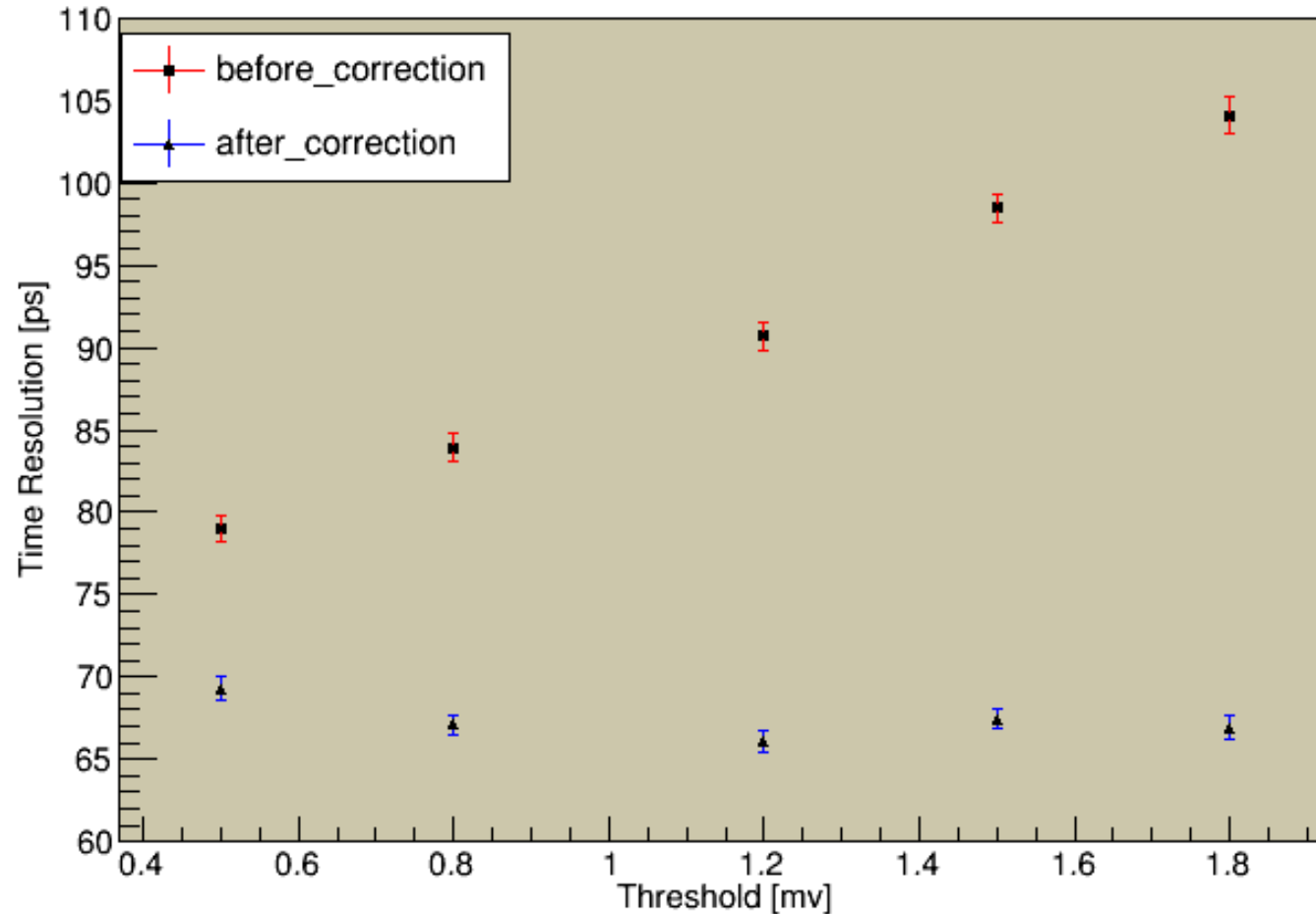
# Time Resolution---Linear Interpolation





# Time Resolution

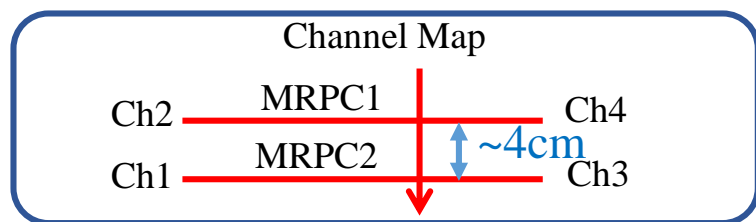
Time Resolution vs Threshold



- ✓ Time-slewing effect is more obvious when increasing the threshold.
- ✓ The time resolution after slewing correction is almost the same at different threshold.



# Time Resolution---Vertical incident



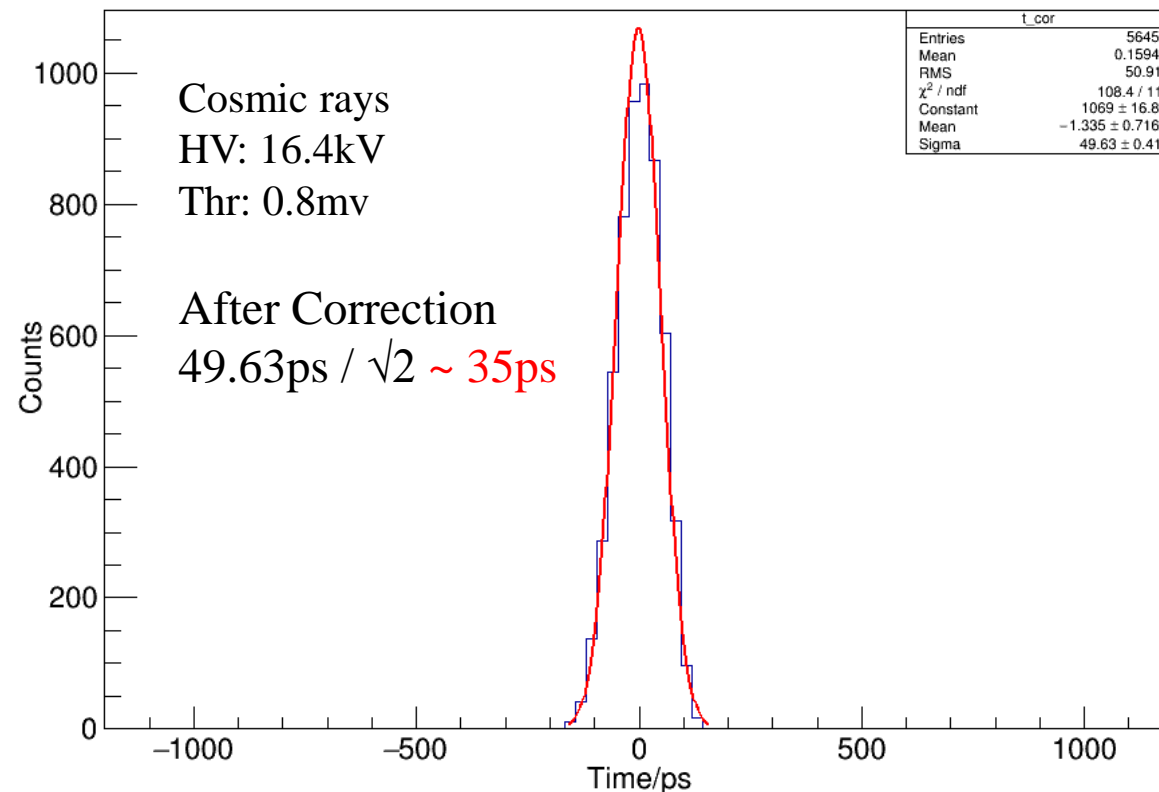
Cut:

Mean time difference:  $\sim 135\text{ps}$

Assuming  $\sigma(\text{MRPC}) \sim 50\text{ps}$

$\text{Mean} - 3\sigma < t_{\text{ch1}} - t_{\text{ch2}} < \text{Mean} + 3\sigma$

$\text{Mean} - 3\sigma < t_{\text{ch3}} - t_{\text{ch4}} < \text{Mean} + 3\sigma$



# Next

Next...

- ❑ Increase the applied high voltage (now 110kv/cm)
- ❑ Test of 4 strips at least (needs 16 channels)
- ❑ Test the 32-gaps MRPC (gas gap width~104 $\mu$ m)
- ❑ Beam test

...

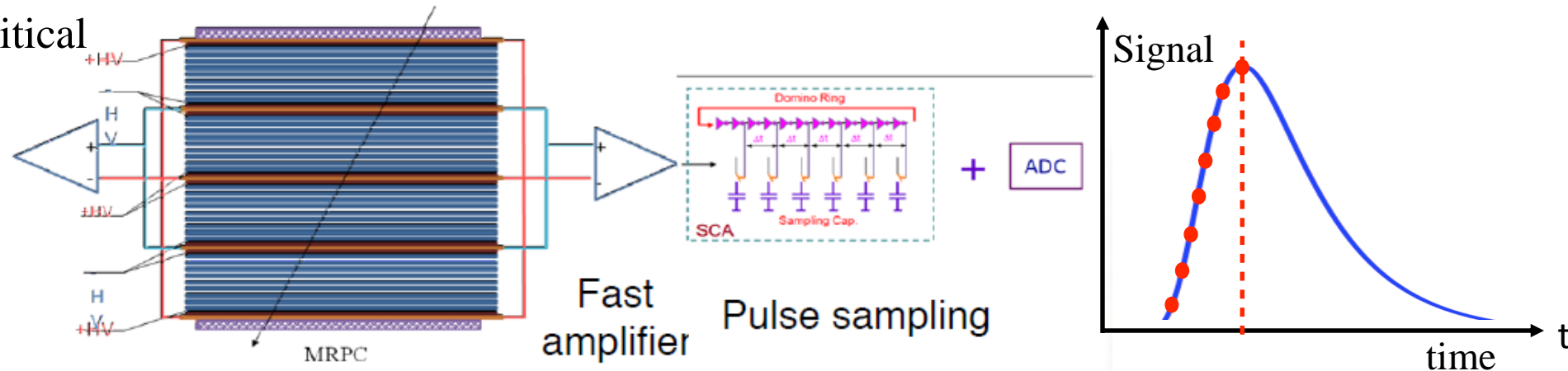
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## **Part II**

# **Study on the impedance of transmission lines in MRPC Detector**

# Introduction

- In SoLID experiment, the requirements for the Time-of-Flight(ToF) system are:
  - ✓  $\pi/k$  separation up to  $7\text{GeV}/c$
  - ✓ Time resolution  $< 20\text{ps}$
  - ✓ Rate capability  $> 10\text{kHz}/\text{cm}^2$
- The next generation MRPC is proposed by Tsinghua collaboration
- The electronics: fast amplifier and wave form digitizer system
- **Impedance matching** of the signal transmission line to the input impedance of the front-end electronics is very critical



# The impedance test

The impedance test platform based on Digital Sampling Oscilloscope (DSA8300) has been set up  
Based on a dual-channel Time Domain Reflectometry (TDR) sampling module  
It allows for differential or common mode **TDR** or **S-parameter** testing of two coupled lines

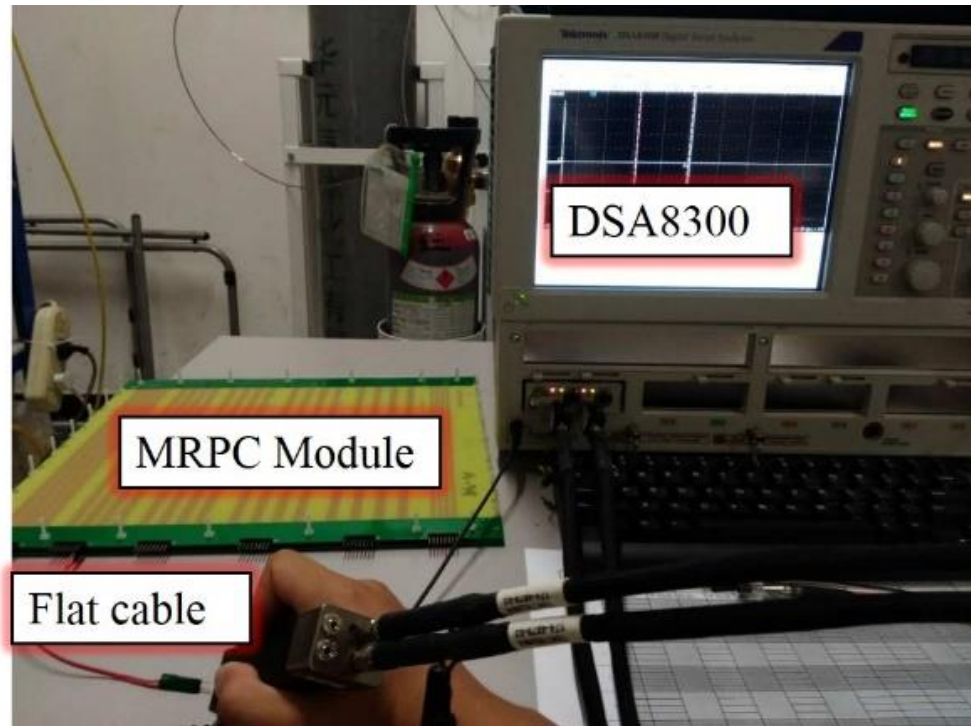
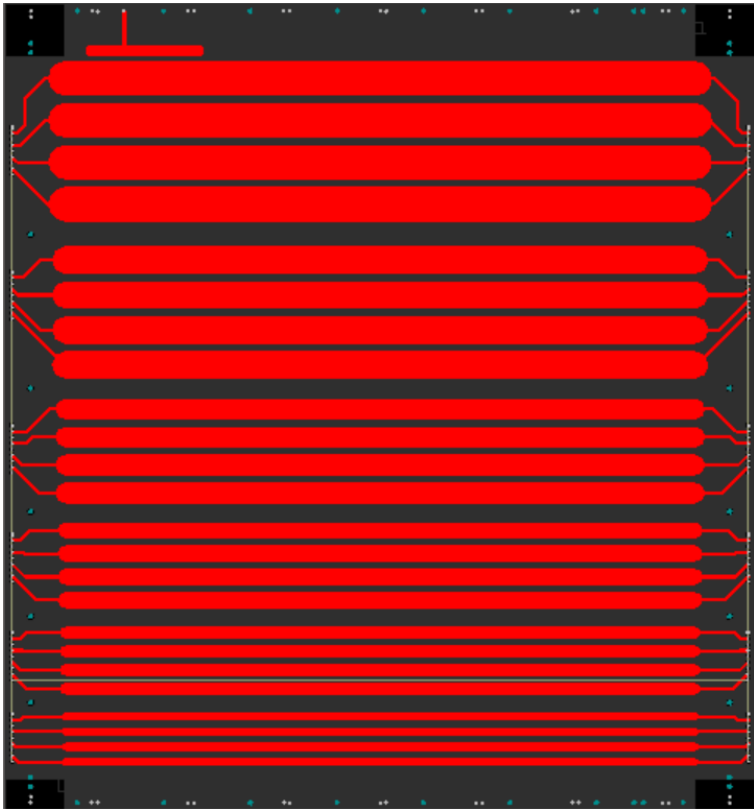


Figure 1 Impedance Test Platform



Figure 2 Differential TDR Waveforms

# The impedance test



PCB Design with different width of strips

## MRPC parameters:

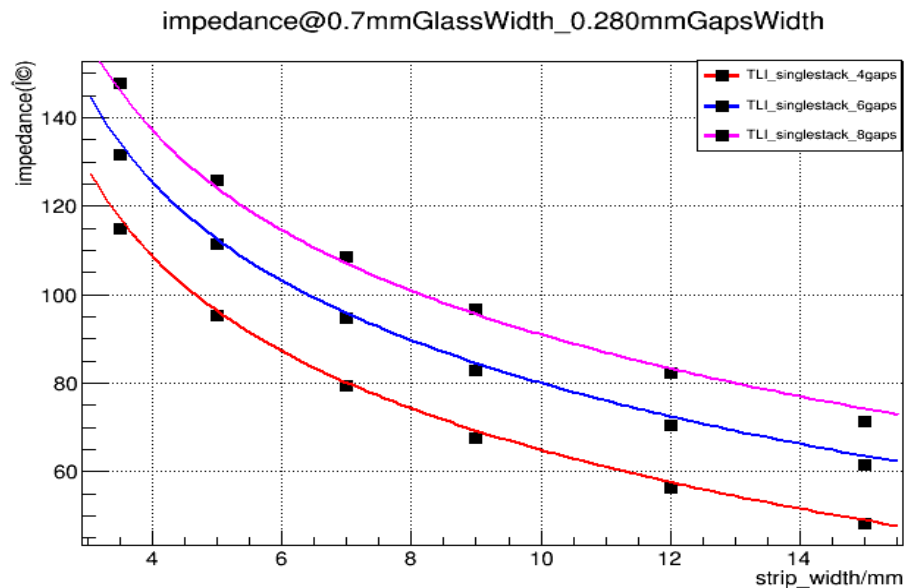
- Different width of strips: 3.5, 5, 7, 9, 12, 15(mm)
  - The number of gas gaps: 4, 6, 8
  - The number of stacks: 1, 2, 3, 4
  - The thickness of gas gaps: 0.12, 0.20, 0.28(mm)  
----Determined by fishing line
  - The thickness of resistive electrodes (**float glass**): 0.23, 0.7 (mm)
- **72** kinds of different structures of the detectors have been finished and tested
- **432** sets of impedance data

# Approximate formula for impedance estimation

Goal:

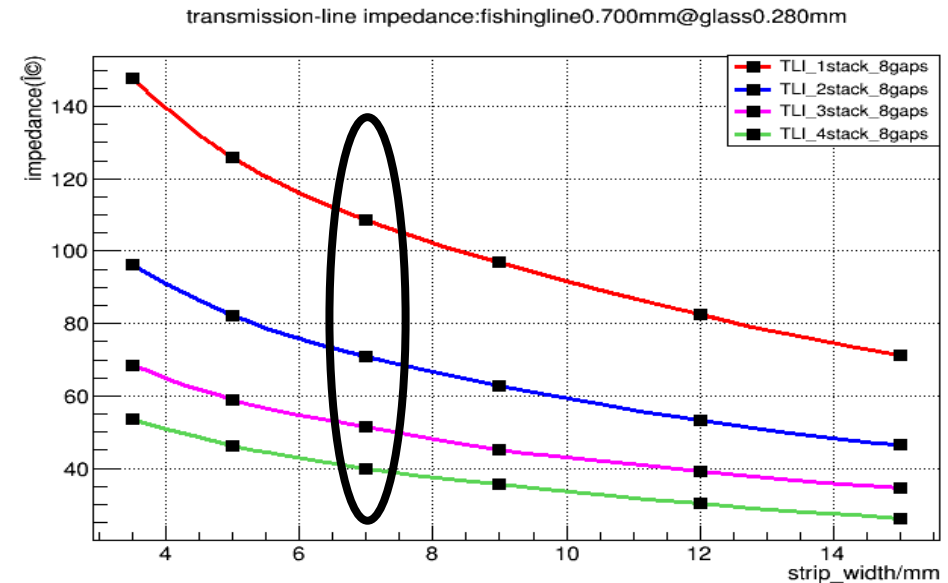
Study on the relationship between the impedance and the width of strip, the thickness of gaps.....

Develop an approximate formula for impedance estimation



Impedance Results of three single-stack MRPCs

$$Z \propto \log \frac{a}{\text{width of strip} + b}$$

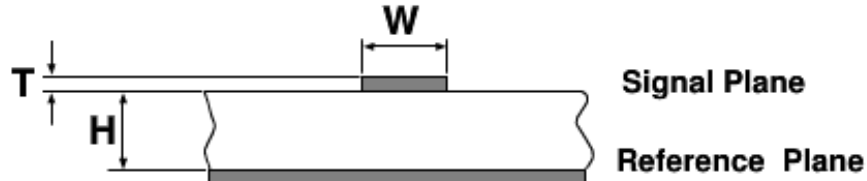


Impedance Results of MRPC modules with different stacks

$$Z \propto \frac{a}{\text{num of stacks} + b}$$

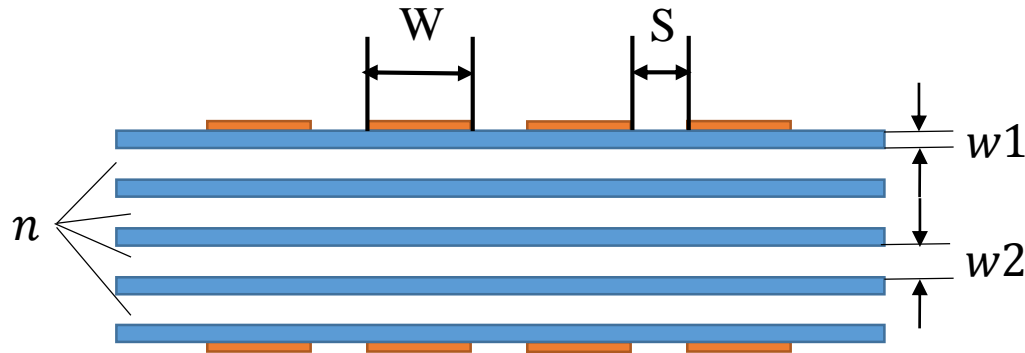


# Approximate formula for impedance estimation



(c) Surface Microstrip

$$Z_0 = \frac{87 \cdot \ln[5.98H / (0.8W + T)]}{\sqrt{\epsilon_r + 1.41}} \text{ in ohms}$$



**Approximate formula** for single-stack MRPC based on float glass:

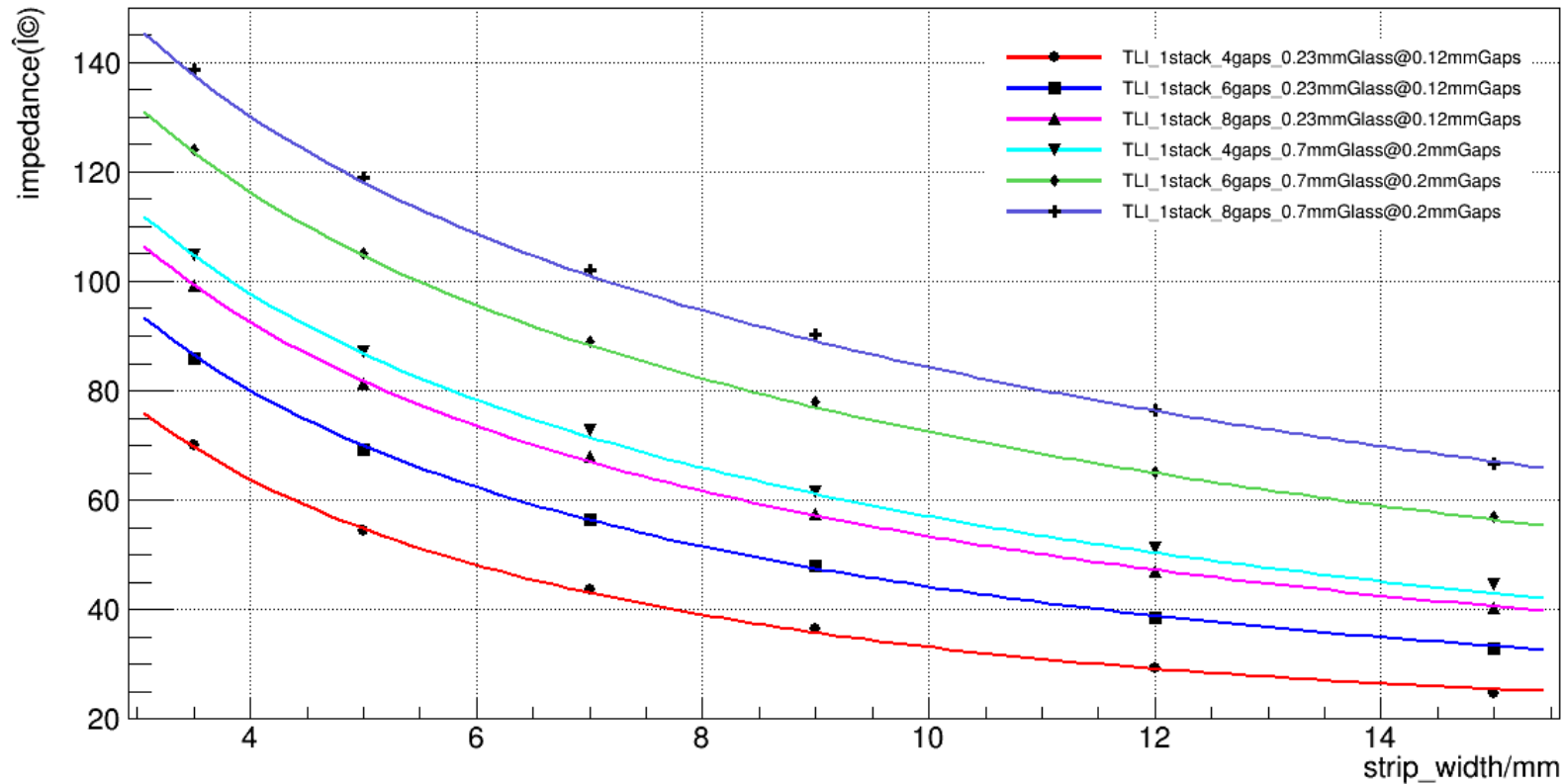
$$\begin{aligned} Imp = & \frac{a(1)}{\sqrt{\epsilon + a(2)}} \times \log \frac{a(3) \times n \times w2 + a(4) \times (n + 1) \times w1}{w - a(5)} \\ & + a(6) \times \sqrt{\frac{w}{n \times w2 + (n + 1) \times w1}} \\ & + a(7) \times \log \frac{w1}{w2} \end{aligned}$$

Coefficients:  $a(1) - a(7)$   
 Number of gas gaps:  $n$   
 Width of glass:  $w1$   
 Width of gaps:  $w2$   
 Width of strips:  $w$   
 Equivalent dielectric constant:  $\epsilon$

$$\epsilon = \left( \frac{n * w2 * \sqrt{\epsilon_{air}} + (n + 1) * w1 * \sqrt{\epsilon_{galss}}}{n * w2 + (n + 1) * w1} \right)^2$$

# Approximate formula for impedance estimation

The coefficients of the approximate formula can be determined by **nonlinear least squares algorithm** with **MATLAB**



Coefficients :

228.874961610702

8.82748996785656

5.85602450540971

0.630327134551313

0.327968220216055

33.2812166777354

21.6324521070679

RMSE= 0.8819

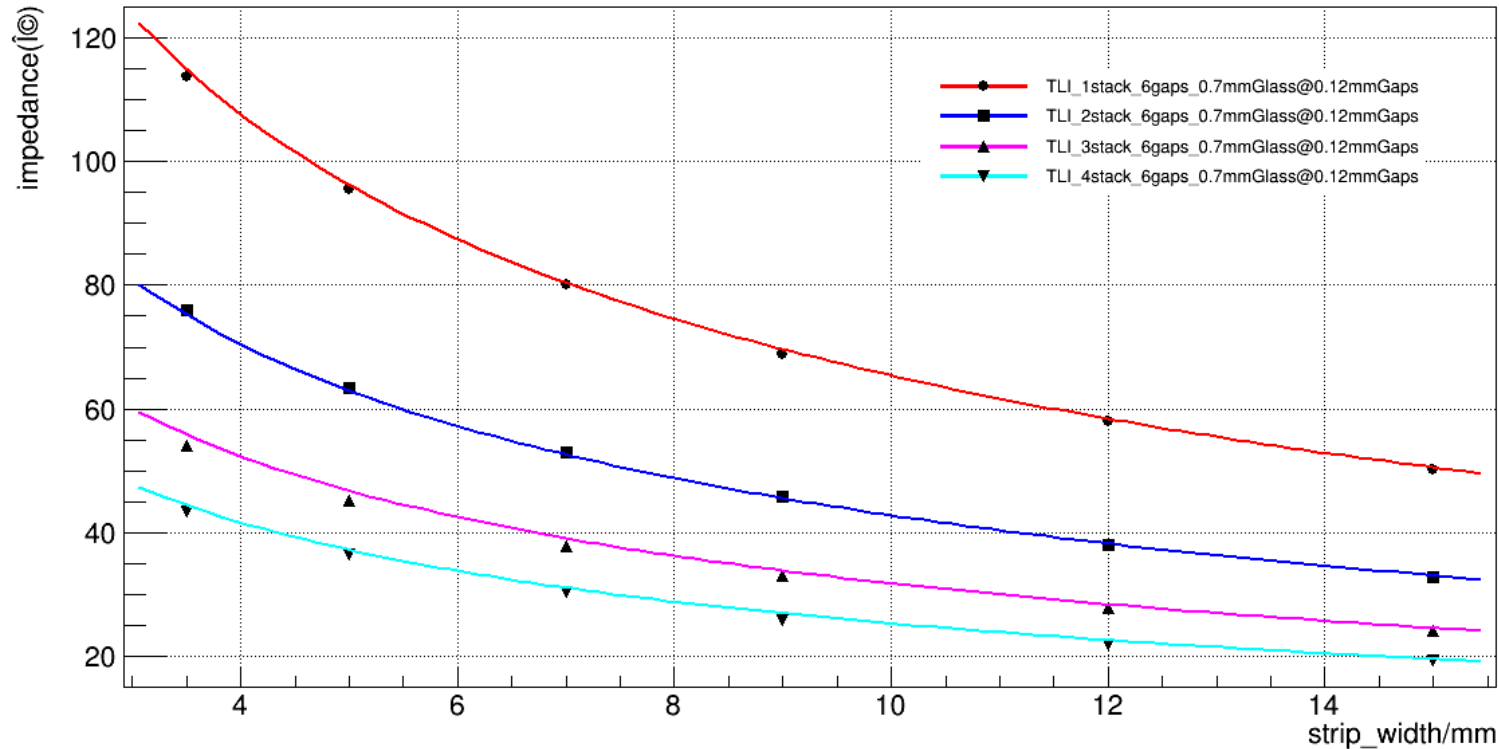
R-square= 0.9989

$$Imp = \frac{a(1)}{\sqrt{\varepsilon + a(2)}} \times \log \frac{a(3) \times n \times w2 + a(4) \times (n + 1) \times w1}{w - a(5)} + a(6) \times \sqrt{\frac{w}{n \times w2 + (n + 1) \times w1}} + a(7) \times \log \frac{w1}{w2}$$

# Approximate formula for impedance estimation

Approximate formula for different-stack MRPC based on float glass:

$$Imp = \frac{\frac{a(1)}{\sqrt{\varepsilon + a(2)}} \times \log \frac{a(3) \times n \times w2 + a(4) \times (n + 1) \times w1}{w - a(5)} + a(6) \times \sqrt{\frac{w}{n \times w2 + (n + 1) \times w1}} + a(7) \times \log \frac{w1}{w2}}{ns + a(8)}$$



Coefficients:  $a(1) - a(8)$

Number of gas gaps:  $n$

Number of stacks:  $ns$

Width of glass:  $w1$

Width of gaps:  $w2$

Width of strips:  $w$

Equivalent dielectric constant:  $\varepsilon$

RMSE= 1.2043

R-square= 0.9977

# Machine learning approaches--SVR

New idea:

**Predict** the impedance of transmission lines using machine learning approaches

✓ **Support vector machines (SVMs)** are a set of supervised learning methods used for [classification](#), [regression](#) and [outliers detection](#)

*The basic idea:*

Suppose we are given training data  $\{(x_1, y_1), \dots, (x_m, y_m)\} \in R^n \times R$ , where  $n$  means the dimension of input patterns.

In  $\varepsilon$ -SV regression, we begin by describing the case of linear functions  $f$ , taking the form:

$$f: R^n \rightarrow R \text{ with } f(x) = \mathbf{W} \cdot \mathbf{X} + b \text{ and } \mathbf{W} \in R^n, b \in R$$

We can write this problem as a convex optimization problem:

Minimize:  $\frac{1}{2} \|\mathbf{W}\|^2$

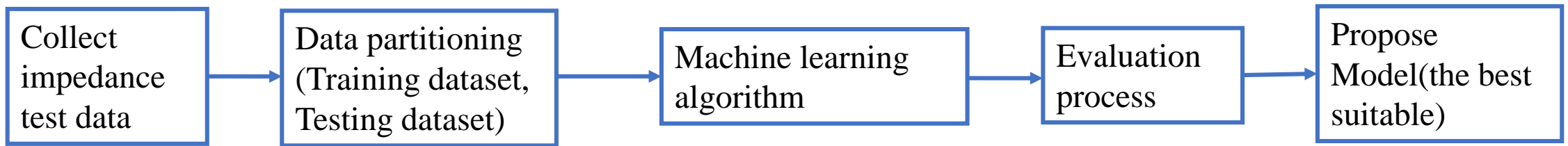
Subject to  $y_i - \mathbf{W} \cdot \mathbf{X}_i - b \leq \varepsilon$   
 $\mathbf{W} \cdot \mathbf{X}_i - b - y_i \leq \varepsilon$

# Machine learning approaches--SVR

**class sklearn.svm.SVR** (kernel='rbf', degree=3, gamma='auto', coef0=0.0, tol=0.001, C=1.0, epsilon=0.1, shrinking=True, cache\_size=200, verbose=False, max\_iter=-1)

The free parameters : the values of the cost function C, the width  $\epsilon$  and kernel function

## Workflow diagram of experiment design



numofstack	numofgaps	widthofglass	widthofgaps	widthofstrips	dielectric	Impedance
1	4	0.23	0.28	3.5	3.11	90.47
1	4	0.23	0.28	5.0	3.11	75.33
1	4	0.23	0.28	7.0	3.11	60.67
1	4	0.23	0.28	9.0	3.11	52.07
1	4	0.23	0.28	12.0	3.11	42.22
1	4	0.23	0.28	15.0	3.11	35.56
2	4	0.23	0.28	3.5	3.11	63.80
2	4	0.23	0.28	5.0	3.11	50.42
2	4	0.23	0.28	7.0	3.11	42.25
2	4	0.23	0.28	9.0	3.11	35.31
2	4	0.23	0.28	12.0	3.11	28.94
2	4	0.23	0.28	15.0	3.11	24.40
3	4	0.23	0.28	3.5	3.11	45.48
3	4	0.23	0.28	5.0	3.11	37.21
3	4	0.23	0.28	7.0	3.11	30.59

- SVR  
(parameter setting,  
kernel type selection)

- Mean absolute error(MAE)
- Mean squared error(MSE)
- $R^2$  score, the coefficient of determination

# Machine learning approaches--SVR

Model Attributes	Kernel type	Degree	C	Gamma	MAE	MSE	R <sup>2</sup> score
Number of stacks, Number of gaps, Width of glass, Width of gaps, Width of strip, Equivalent dielectric constant	'rbf'	3	500	0.06	1.254	2.315	0.996
	<b>'rbf'</b>	<b>3</b>	<b>600</b>	<b>0.07</b>	<b>1.124</b>	<b>2.002</b>	<b>0.996</b>
	'rbf'	3	500	0.08	1.188	2.222	0.996
	'rbf'	3	500	0.1	1.344	2.664	0.995
	'rbf'	3	1000	0.1	1.576	3.026	0.994
	'rbf'	3	3000	0.1	1.682	3.401	0.994
	'rbf'	3	500	0.2	2.401	7.04	0.987
	'rbf'	3	1000	0.2	2.396	7.058	0.987
	'poly'	3	1000	0.1	3.16		0.957

“rbf”(radial basis function):  $\exp(-\gamma \cdot |u-v|^2)$

# Machine learning approaches--SVR

kernel='rbf' C=600 Gamma=0.07

numofstack	numofgaps	thicknessofglass	thicknessofgaps	widthofstrips	dielectric	Impedance_Test	Impedance_Predict
1	5	0.23	0.28	3.5	3.1300	<b>101.45</b>	<b>99.40</b>
1	5	0.23	0.28	5.0	3.1300	<b>83.02</b>	<b>82.33</b>
1	5	0.23	0.28	7.0	3.1300	<b>69.76</b>	<b>67.26</b>
1	5	0.23	0.28	9.0	3.1300	<b>59.41</b>	<b>57.88</b>
1	5	0.23	0.28	12.0	3.1300	<b>48.27</b>	<b>47.11</b>
1	5	0.23	0.28	15.0	3.1300	<b>41.35</b>	<b>39.00</b>
1	7	0.23	0.28	3.5	3.0635	<b>119.53</b>	<b>117.96</b>
1	7	0.23	0.28	5.0	3.0635	<b>99.46</b>	<b>100.98</b>
1	7	0.23	0.28	7.0	3.0635	<b>84.14</b>	<b>84.08</b>
1	7	0.23	0.28	9.0	3.0635	<b>73.14</b>	<b>73.39</b>
1	7	0.23	0.28	12.0	3.0635	<b>59.80</b>	<b>60.13</b>
1	7	0.23	0.28	15.0	3.0635	<b>50.69</b>	<b>50.28</b>

- Mean absolute error: **1.124**
- Mean squared error: **2.002**
- R<sup>2</sup> score, the coefficient of determination: **0.996**



# Data Analysis Methods

	Machine learning approaches--SVR	Approximate formula
Resistive electrodes	Float glass( $\epsilon=6.5$ )	Float glass( $\epsilon=6.5$ )
The number of stacks	1,2,3,4	1,2,3,4
The number of gas gaps	4 ~8	4 ~8
Width of strips	3.5~15	3.5~15
Thickness of gaps	0.12~0.28	0.1~0.3
Thickness of glass	0.23~0.7	0.23~0.7

- The two methods for impedance estimation in MRPC detector show a good performance in a condition shown in the table above

# Conclusion

- 72 kinds of different structures of float glass MRPC have been finished and the impedance has been tested
- An approximate formula for different-stack MRPC based on float glass has been proposed. It shows great performance of impedance estimation in MRPC detector.

$$Imp = \frac{\frac{a(1)}{\sqrt{\varepsilon + a(2)}} \times \log \frac{a(3) \times n \times w2 + a(4) \times (n + 1) \times w1}{w - a(5)} + a(6) \times \sqrt{\frac{w}{n \times w2 + (n + 1) \times w1}} + a(7) \times \log \frac{w1}{w2}}{ns + a(8)}$$

- This study is also done to predict the impedance of transmission lines using machine learning approaches----SVR(Support Vector Regression)



**Thank you for your attention!**