



## Electromagnetic Calorimeter Software for Super Tau-Charm Facility



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## Super Tau-Charm Facility



Electron-positron collider experiment

- □ High luminosity: beyond  $0.5 \times 10^{35} cm^{-2} \cdot s^{-1}$  @ 4GeV
- $\square$  Wide energy region: center-of-mass energy range of 2~7 GeV



#### **Requirement for ECAL**

## Requirements for ECAL

### □ Fast response

- Challenge of high Luminosity
  - High count rate
  - Extremely high background

### □ High precision

- Energy resolution
  - Better than 2.5% @1GeV
- Position resolution
  - Better than 5mm @1GeV
- $\succ$  Time resolution
  - Better than 300ps @1GeV



Energy distribution for photons



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## **ECAL** Design

### □ Total absorption calorimeter

- > Barrel:  $51 \times 132 = 6732$
- > Endcap:  $3 \times (85 + 102 + 136) \times 2 = 1938$
- Crystal Size:
  - $5 \times 5 \times 28(15X_0) \text{ cm}^3$

### □ Sensitive Unit

- Pure CsI (pCsI) crystal
  - Fast decay time (~30ns)
  - Good radiation hardness
  - Low light yield
- Avalanche photodiode (APD)
  - Short wavelength type
  - Large area  $(10 \times 10 \ mm^2 \times 4)$







### **ECAL Setup**

□ "Dead Material"

- ➤ 150-µm Teflon reflective film
- > 75-µm polyethylene insulating film
- > 75-µm Al electrostatic shielding film
- No supporting material
- □ Light Yield: 100 p.e./MeV
- □ Light Collection Non-uniformity
  - > collection efficiency:  $\epsilon(l) = 95\% + l/L \times 5\%$

Setup

- $\Box \sigma_{noise} = 1.0 \text{ MeV}$
- $\square E_{hit\_thres} = 5.0 \text{ MeV}$
- Secondary Particles Hit APD



320cm









- □ Charge sensitive amplifier and pole-zero cancellation with shaping time 100 ns
- □ High and Low Gain (300 MeV/3000MeV)
- □ 16-bit ADC (~16000 channels) with 80 MHz sampling rate



Based on OSCAR



### □ OSCAR: Offline Software of Super Tau-Charm Facility



## Simulation Algorithm



□ Based on Geant4 Simulation

Process the information for each Geant4 step

- Data Model (layer as a unit)
  - Thickness 2cm as a layer unit
  - Each unit with a energy-time distribution: points with (E,T)
  - ➢ Bin width 500ps for energy-time distribution
  - ✓ Save storage space
  - ✓ Minimal loss of information



## Digitization Algorithm

### Waveform shaping





#### Waveform fitting

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## **Digitization Algorithm**

### Waveform fitting

□ Template fit to extract Amplitude and Time

- > Find the points around amplitude peak to do the template fitting
- > Template shape function:  $f(t) = A \times f(t \tau) + p$
- $\lambda \chi^{2} = \sum_{i,j} (y_{i} A \cdot f(t_{i} \tau) p) \cdot S_{ij}^{-1} \cdot (y_{j} A \cdot f(t_{j} \tau) p)$  $\lambda pply \frac{\partial \chi^{2}}{\partial A} = 0, \frac{\partial \chi^{2}}{\partial \tau} = 0$

Where  $y_i$  and  $t_i$  are from readout waveform; the electronics foundation p in digitization is p = 0; A and  $\tau$  are the amplitude and time from fitting result;  $S_{ii}$  is the noise covariance matrix.

□ Pipeline fit for pile-up recovery





## **Reconstruction Algorithm**



□ A complete reconstruction algorithm of ECAL is developed





- ➢ Fitted by Crystal Ball function
- Energy resolution defined by  $\sigma_E = \frac{FWHM}{2.355}$

## Challenges of high background



Luminosity-related Background
 Radiative BhaBha Scattering (RBB)

Two Photon Process



Variation of the background counting rate with polar angle **Counting rate reaches the order of MHz** 

#### □ Single-beam related Background

- Thouschek Effect
- Coulomb Scattering
- Bremsstrahlung



Momentum distribution of background particles

Most background particles concentrate in the low momentum region

## Pile-up recovery



- Pile-up is superimposed on the signal waveform
  - Inaccurate fit results of amplitude and time
  - ➤ Larger resolution of energy and time



### **D** Pipeline fitting method

- Real-time online processing
- > Template fit once for each fitting
- Fit successful
  - $\rightarrow$  Remove template
  - $\rightarrow$  Ongoing processing

### □ Multi-template fit

Has been used to study the capability for pile-up recovery

## Pile-up Recovery

### Pipeline fit



- **D** Pipeline fitting method
  - Real-time online processing
  - Template fit once for each fitting
    - Each fitting begin with different ADC point
  - Fit successful
    - $A > E_{thr}$
    - $\triangle T < 12.5/2$
    - $\rightarrow$  Remove template
    - $\rightarrow$  Ongoing processing

- Optimization

- **D** Pipeline fitting method
  - Real-time online processing
  - Template fit once for each fitting
    - Each fitting begin with different ADC point
    - Add one more fitting between two ADC points
  - Fit successful
    - $A > E_{thr}$
    - $\triangle T < 12.5/2$
    - $\chi^2 / ndf < 3$ :  $\chi^2 / ndf = [\sum_{i,j} (y_i - A \cdot f(t_i - \tau) - p) \cdot S_{ij}^{-1} \cdot (y_j - A \cdot f(t_j - \tau) - p)] / ((n - 2) \cdot (\sigma_{nos}^2 + (A \cdot 0.01)^2))$
    - $\rightarrow$  Cache and compared with next fit
    - $\rightarrow$  Remove template
    - → Ongoing processing

#### 2024/7/9

#### **Pipeline fit**

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# □ The performance between pipeline fit and single template fit for signal process without background

#### > Energy and time resolutions can achieve similar performance







## **Reconstruction Performance**

### **Energy Reconstruction**







## **Reconstruction Performance**

### **Time Reconstruction**





## **Reconstruction Performance**

**Position Reconstruction** 

 Splitting algorithms (used by BESIII and Panda)

$$\succ a_{ik} = \frac{E_k \times \exp(c \times \frac{r_{ik}}{R_M})}{\sum_{j=1}^m E_j \times \exp(c \times \frac{r_{ij}}{R_M})}$$

□ Barycenter method

$$\succ X_c = \sum_j^N W_j(E_j) \cdot X_j / \sum_j^N W_j(E_j)$$
  
Where  $W_j(E_j) = \max\{0, a - \sqrt{-\ln\left(E_j / \sum_j^N E_j\right)}\}$ 



✓ Meet the requirement





- □ The software of ECAL has been established based on OSCAR
- □ The simulated performance of ECAL meets requirements with the background concerned
  - ✓ Energy measurement with 2.27% @ 1 GeV
  - ✓ Time measurement with 153 ps @ 1 GeV
  - ✓ Position measurement with 4.0 mm @ 1 GeV

### Thanks for your listening!



# Back up



## ECAL Design —— Sensitive Unit

### □ Pure CsI crystal + APD photo-device

- Pure CsI (pCsI) crystal
  - ✓ Fast decay time
  - ✓ Good radiation hardness
  - ✓ Low light yield
- ➤ Crystal Size:
  - $\checkmark$  Total radiation length
    - $15 X_0$  (28 cm)
  - ✓ End face size front end:  $\sim 5 \times 5 \ cm^2$ back end:  $\sim 6.5 \times 6.5 \ cm^2$

#### Avalanche photodiode (APD)

- $\checkmark$  Short wavelength type
- $\checkmark \quad \text{Large area} \ (10 \times 10 \ mm^2 \times 4)$



APD

#### ECAL pCsI crystal unit

pCsl

Crystal	Pure Csl
Density (g/cm <sup>3</sup> )	4.51
Melting Point (°C)	621
Radiation Length (cm)	1.86
Moliere Radius (cm)	3.57
Refractive index	1.95
Hygroscopicity	Slight
Luminescence (nm)	310
Decay time (ns)	30 6
Light yield (%)	3.6 1.1
Dose rate dependent	No
D(LY)/dT (%/°C)	-1.4
Experiment	KTeV
	Mu2e

## Simulation Algorithm

### Data Model



□ Sizes of thickness and time bin width have been optimized

- > No large difference of time distribution for different thickness
  - Consider the non-uniformity may vary in the future : 2cm
- Different time bin width with different time resolution and central value
  - Difference in central value approximately equivalent to a shift of the template
  - Consider the resolution and the similarity to the template: 500ps



## Simulation Algorithm

Data Model



### **D** Energy distribution and comparison with template



## **Template Fitting**



• Template shape function:  $f(t) = A \times f(t - \tau) + p$ 

• 
$$\chi^2 = \sum_{i,j} (y_i - A \cdot f(t_i - \tau) - p) \cdot S_{ij}^{-1} \cdot (y_j - A \cdot f(t_j - \tau) - p)$$
  
• Apply  $\frac{\partial \chi^2}{\partial A} = 0, \frac{\partial \chi^2}{\partial \tau} = 0, \frac{\partial \chi^2}{\partial p} = 0$ :

$$\begin{cases} \sum_{i,j} f_{ki} \cdot S_{ij}^{-1} \cdot (y_j - Af_{kj} - Bf'_{kj} - p) = 0 \\ \sum_{i,j} f'_{ki} \cdot S_{ij}^{-1} \cdot (y_j - Af_{kj} - Bf'_{kj} - p) = 0 \\ \sum_{i,j} 1 \cdot S_{ij}^{-1} \cdot (y_j - Af_{kj} - Bf'_{kj} - p) = 0 \end{cases}$$

$$\begin{pmatrix} F_k \cdot S^{-1} \cdot F_k^T & F_k \cdot S^{-1} \cdot F_k'^T & F_k \cdot S^{-1} \cdot I \\ F'_k \cdot S^{-1} \cdot F_k^T & F'_k \cdot S^{-1} \cdot F_k'^T & F'_k \cdot S^{-1} \cdot I \\ I \cdot S^{-1} \cdot F_k^T & I \cdot S^{-1} \cdot F_k'^T & I \cdot S^{-1} \cdot I \end{pmatrix} \cdot \begin{pmatrix} A \\ B \\ p \end{pmatrix} = \begin{pmatrix} F_k \cdot S^{-1} \cdot Y \\ I \cdot S^{-1} \cdot Y \end{pmatrix}$$

$$\begin{pmatrix} A \\ B \\ p \end{pmatrix} = \begin{pmatrix} F_k \cdot S^{-1} \cdot F_k^T & F_k \cdot S^{-1} \cdot F_k'^T & F_k \cdot S^{-1} \cdot I \\ F'_k \cdot S^{-1} \cdot F_k^T & F'_k \cdot S^{-1} \cdot F'_k' & F'_k \cdot S^{-1} \cdot I \\ I \cdot S^{-1} \cdot F_k^T & I \cdot S^{-1} \cdot F'_k' & I \cdot S^{-1} \cdot I \end{pmatrix}^{-1} \cdot \begin{pmatrix} F_k \cdot S^{-1} \cdot Y \\ F'_k \cdot S^{-1} \cdot Y \\ I \cdot S^{-1} \cdot F_k' & I \cdot S^{-1} \cdot F'_k' & I \cdot S^{-1} \cdot I \end{pmatrix}$$

## Nonnegative Least Square (NNLS)



#### Algorithm *fnnls* :

Input:  $\mathbf{A} \in \mathbf{R}^{m \times n}$ ,  $\mathbf{b} \in \mathbf{R}^m$ Output:  $\mathbf{x}^* \ge 0$  such that  $\mathbf{x}^* = \arg \min \|\mathbf{A}\mathbf{x} - \mathbf{b}\|^2$ . Initialization:  $P = \emptyset, R = \{1, 2, \cdots, n\}, \mathbf{x} = \mathbf{0}, \mathbf{w} = \mathbf{A}^T \mathbf{b} - (\mathbf{A}^T \mathbf{A}) \mathbf{x}$ repeat

- 1. Proceed if  $R \neq \emptyset \land [\max_{i \in R}(w_i) > tolerance]$
- 2.  $j = \arg \max_{i \in R} (w_i)$
- 3. Include the index j in P and remove it from R
- 4.  $\mathbf{s}^P = [(\mathbf{A}^T \mathbf{A})^P]^{-1} (\mathbf{A}^T \mathbf{b})^P$ 4.1. Proceed if  $\min(\mathbf{s}^P) \leq 0$ 4.2.  $\alpha = -\min_{i \in P} [x_i/(x_i - s_i)]$ 4.3.  $\mathbf{x} := \mathbf{x} + \alpha(\mathbf{s} - \mathbf{x})$ 4.4. Update R and P 4.5.  $\mathbf{s}^P = [(\mathbf{A}^T \mathbf{A})^P]^{-1} (\mathbf{A}^T \mathbf{b})^P$ 4.6.  $\mathbf{s}^R = \mathbf{0}$ 5.  $\mathbf{x} = \mathbf{s}$ 6.  $\mathbf{w} = \mathbf{A}^T (\mathbf{b} - \mathbf{A}\mathbf{x})$

#### Convention:

- b: A real pulse with m points
- x: fitted amplitudes for n pulses
- A: the ith column of A represents the template for the ith pulse and of course each template has m points.
- P: passive set currently not fixed amps
- R: active set currently fixed amplitudes



**Multi-template** 

## Pile-up Recovery

### Multi-template fit

Fit all the potential waveforms with template
 Isolate signals by time
 The fit minimizes the χ<sup>2</sup> defined as:

$$\chi^{2} = \left(\sum_{j=1}^{N} A_{j} \overrightarrow{p_{j}} - \vec{S}\right)^{T} C^{-1} \left(\sum_{j=1}^{N} A_{j} \overrightarrow{p_{j}} - \vec{S}\right)^{T}$$

Where:

N is the number of templates; vector  $\vec{S}$  comprise the readout samples; vector  $\overrightarrow{p_j}$  is the waveform template;  $A_j$  are the amplitudes, which are obtained by the fit; C is the noise covariance matrix.



