## **Unbinned likelihood approach**

to telescope-calorimeter alignment

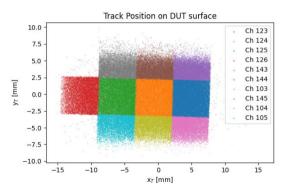
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LUXE ECAL-P meeting November 17, 2025

### Introduction



I started to think more about the problem after Michal's talk in Warsaw:



We have clear correlation between extrapolated track position and calorimeter hit position.

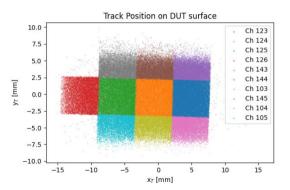
⇒ It should be possible to extract telescope-calorimeter alignment parameters directly from this dependence.

We should be able to use most of collected data for this procedure (without a need to select event samples with special topology) resulting in high alignment precision...

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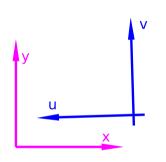
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And having a week free from teaching duties...



### **Alignment parameters**



We need to transform track position (x,y) in the telescope reference frame to position (u,v) in the calorimeter frame.

(u,v) correspond to cell column and row number (integer values for cell center)

General form of transformation:

$$u = -a_x (x \cos \phi + y \sin \phi) + u_0$$
  
$$v = a_y (-x \sin \phi + y \cos \phi) + v_0$$

note horizontal coordinate sign flip

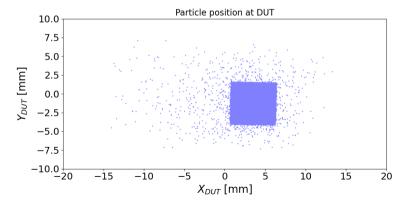
Main parameters:  $u_0$ ,  $v_0$  and  $\phi$ 

Parameters  $a_x$  and  $a_y$  should reflect sensor cell size/pitch, but we can also include them in the fit...



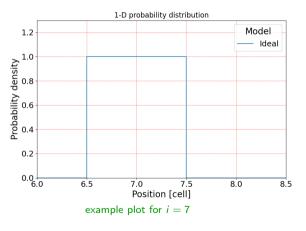
#### Likelihood definition

Example position distribution for events with cell #94 hit in plane #0, run 1492. For the perfect detector performance, we should just see a square





#### Likelihood definition



### 1-D probability density model

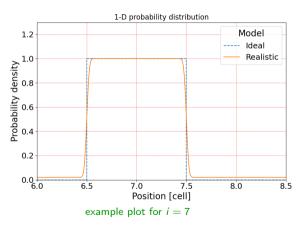
For the ideal detector:

$$p(v) = \begin{cases} 1 & for |u-i| < 0.5 \\ 0 & |u-i| \ge 0.5 \end{cases}$$

where i is the column number of the cell



#### Likelihood definition



## 1-D probability density model

Realistic detector, assuming finite telescope position resolution  $(\sigma)$  and noise  $(p_0)$ :

$$p(v) = p_1 \cdot SF\left(\frac{|u-i|-\frac{1}{2}}{\sigma}\right) + p_0$$

where *SF*() is the "survival function" of the normal distribution (from SciPy library)

 $\sigma$  can be an additional model parameter  $p_1$  and  $p_0$  should rather be fixed



#### Unbinned likelihood

Assuming no correlation between x and y, 2-D likelihood can be written as:

$$p(u,v) = p_1 \cdot SF\left(\frac{|u-i|-\frac{1}{2}}{\sigma_X}\right) \cdot SF\left(\frac{|v-j|-\frac{1}{2}}{\sigma_y}\right) + p_0$$

for calorimeter hit in cell (i,j).



#### Unbinned likelihood

Assuming no correlation between x and y, 2-D likelihood can be written as:

$$p(u,v) = p_1 \cdot SF\left(\frac{|u-i| - \frac{1}{2}}{\sigma_X}\right) \cdot SF\left(\frac{|v-j| - \frac{1}{2}}{\sigma_y}\right) + p_0$$
within coll (i.i.)

for calorimeter hit in cell (i, j).

For a set of events (k = 1 ... N) unbinned likelihood can be defined as:

$$\mathcal{L} = \prod_{k} p(u_k, v_k)$$

where  $(u_k, v_k)$  track positions in calorimeter frame, depend on the alignment parameters.



#### Unbinned likelihood

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These parameters can be then obtained by maximizing  $\mathcal L$  or minimizing the function:

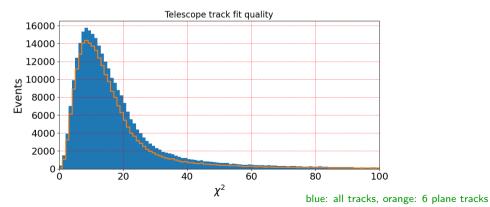
$$-2 \cdot \ell = -2 \cdot \ln \mathcal{L} \approx \chi^2$$

I just tried 'minimize' method from SciPy 'optimize' library and it worked...



## Data sample

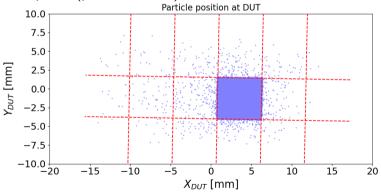
Presented results were obtained for data from Run 1492, selecting events with only one telescope track, fitted to all planes, with  $\chi^2 < 20$ :





## Single cell fit

Example fit (plane 0, cell 94)



Alignment from 3 parameter fit:

$$u_0 = 18.64495 \pm 0.00081$$
  
 $v_0 = 7.22515 \pm 0.00147$ 

$$\phi = -0.01953 \pm 0.00192$$

position in "cell units"



#### Multi-cell fit

For higher accuracy of the fit, and to fit more model parameters, we need to include more cells

Example fit (plane 0) Particle position at DUT 10.0 7.5 5.0 [mm] roa 2.5 0.0 -2.5-5.0-7.5-10.0 -20 -15-1010 15 20

 $X_{DUT}$  [mm]

Alignment from 3 parameter fit:

$$u_0 = 18.64438 \pm 0.00027$$
  
 $v_0 = 7.22621 \pm 0.00027$ 

$$\phi = -0.01758 \pm 0.00021$$

$$\sigma_{\phi}$$
 reduced by order of magnitude!



#### Multi-cell fit

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Alignment from 5 parameter fit:

$$u_0 = 18.64439 \pm 0.00027$$
  
 $v_0 = 7.22587 \pm 0.00030$   
 $\phi = -0.01760 \pm 0.00021$   
 $a_x^{-1} = 5.51901 \pm 0.00131 \text{ mm}$   
 $a_y^{-1} = 5.52846 \pm 0.00286 \text{ mm}$ 

matches the expected pitch



#### Multi-cell fit

For higher accuracy of the fit, and to fit more model parameters, we need to include more cells

Example fit (plane 0) Particle position at DUT 10.0 7.5 5.0 *[mm] put* 2.5 0.0 -2.5-5.0-7.5-10.0 -20 -15-1010 15 20  $X_{DUT}$  [mm]

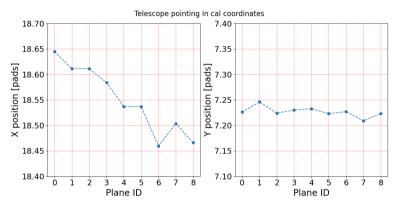
Alignment from 7 parameter fit:

$$u_0 = 18.64460 \pm 0.00022$$
  
 $v_0 = 7.22609 \pm 0.00025$   
 $\phi = -0.01748 \pm 0.00017$   
 $a_x^{-1} = 5.51887 \pm 0.00108$  mm  
 $a_y^{-1} = 5.52968 \pm 0.00246$  mm  
 $\sigma_x = a_x \cdot 76.86 \pm 1.37$  µm  
 $\sigma_y = a_y \cdot 76.54 \pm 1.66$  µm  
independent telescope precision estimate



## Multi-cell fit with 7 parameters

Summary of results for all available telescope planes

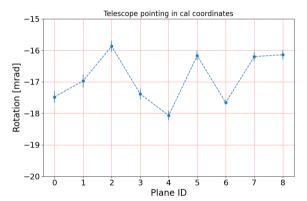


Horizontal dependence ( $\sim 1$  mm shift) corresponds to about  $0.7^{\circ}$  rotation



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Summary of results for all available telescope planes



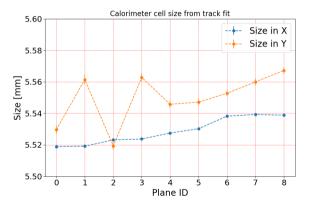
Consistent results between planes ( $\phi \approx 1^{\circ}$ ).

Mechanical precision:  $\frac{10\mu m}{50mm} = 0.2 \text{ mrad}$ 



### Multi-cell fit with 7 parameters

Summary of results for all available telescope planes



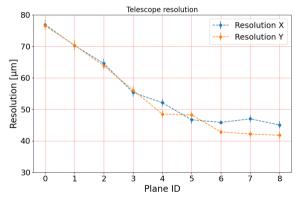
Clear dependency on the plane ID: beam divergence?

Nominal pixel pitch is 5.53 mm.



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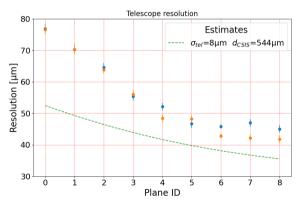


Beam enters from the right ?!



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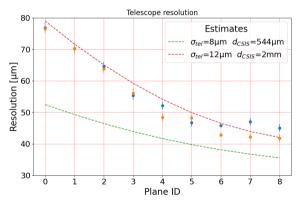


Beam enters from the right ?! Does not match expected dependence ?!



## Multi-cell fit with 7 parameters

Summary of results for all available telescope planes



Beam enters from the right ?! Telescope performance worse and more multiple scattering ?!

### Conclusions



With large statistics of collected data, unbinned likelihood method can be used to extract alignment parameters with high precision.

First tests of the method very promising, but the model is still simplified...

Many features of the model still need to be studied/improved:

- gap between cells
- gap between sensors
  - ⇒ we could also use two sets of alignment parameters for two sensors!

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- "signal" normalization (should it depend on  $a_x$  and  $a_y$ ?)
- "background" normalization (can it be included as fit parameter?)

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- gap between cells
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  - ⇒ we could also use two sets of alignment parameters for two sensors!
- "signal" normalization (should it depend on  $a_x$  and  $a_y$ ?)
- "background" normalization (can it be included as fit parameter?)
- data selection (removal of noisy cells?)
- fit convergence (fails for data of poor quality)