

SKYNET vertex reconstruction

I L \wedge N C E



東京大学
THE UNIVERSITY OF TOKYO



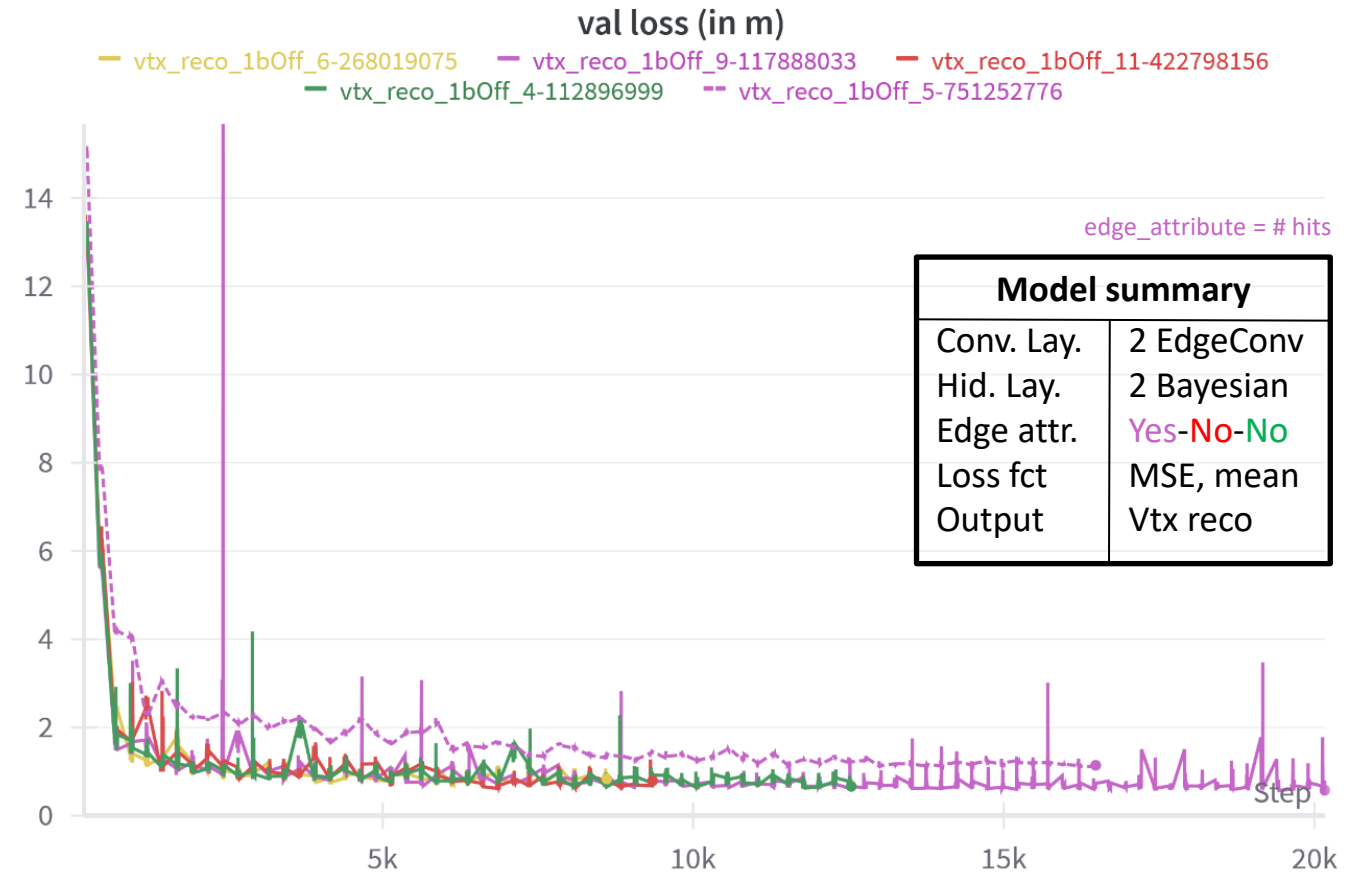
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SKYNET-VTX

- General test on 10 MeV positrons in HK without dark noise
- LEAF performance: resolution (68% CL) \approx 45 cm and \sim 70 cm with dark noise
- First test: GNN regression for $\{x,y,z\}$
 - Resolution \approx 1 m

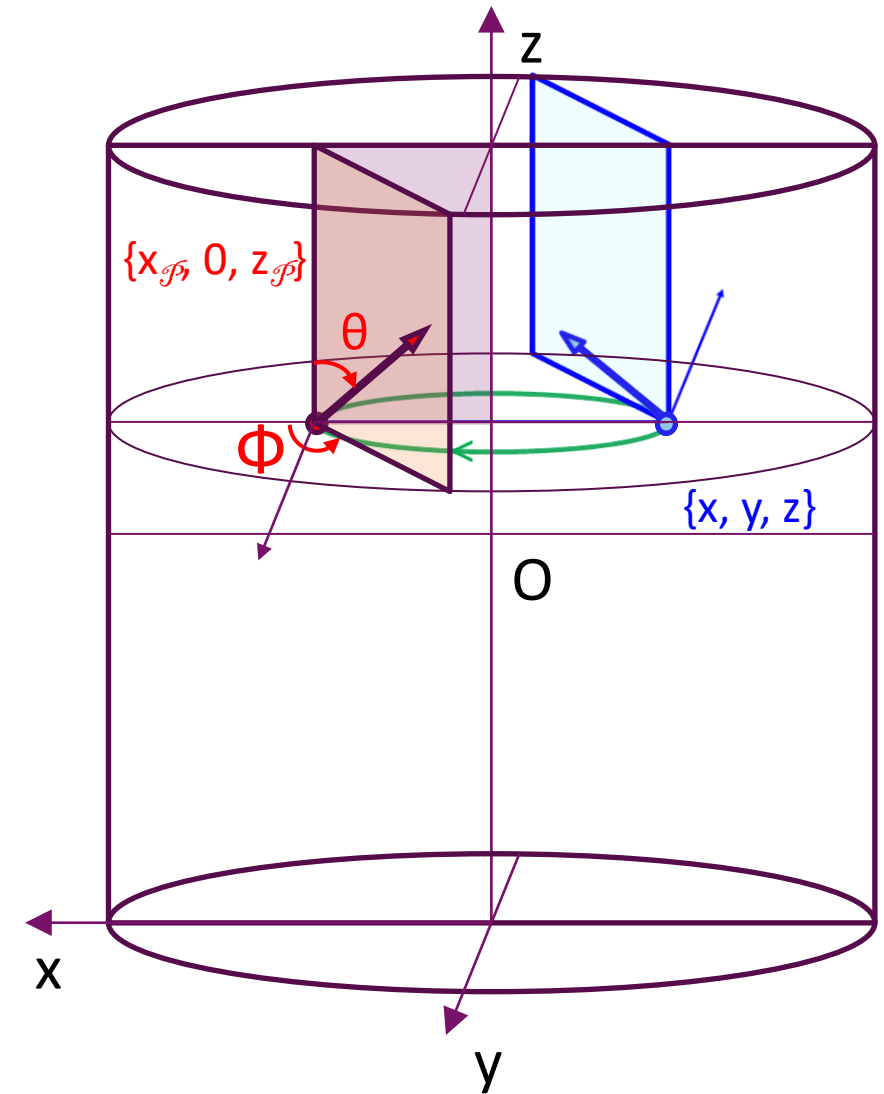


⇒ Not competitive even though the GNN is not optimized

SKYNET-VTX: SLICE EDITION

- Due to tank symmetry by revolution, all events can be mapped to the slice plane $\mathcal{P} := \{y=0, x \text{ \& } z>0\}$
 - Rotation around (Oz) the vertex & PMT hits by ϑ_{true}
 - Rotation around (Oz) by ϑ_{true} + flip the vertex & PMT hits
- Second test: **GNN regression for $\{x_{\mathcal{P}}, z_{\mathcal{P}}\}$ in \mathcal{P}**
 - Resolution ≈ 50 cm

\Rightarrow 3D \rightarrow 2D regression: interesting results even though the GNN is not optimized

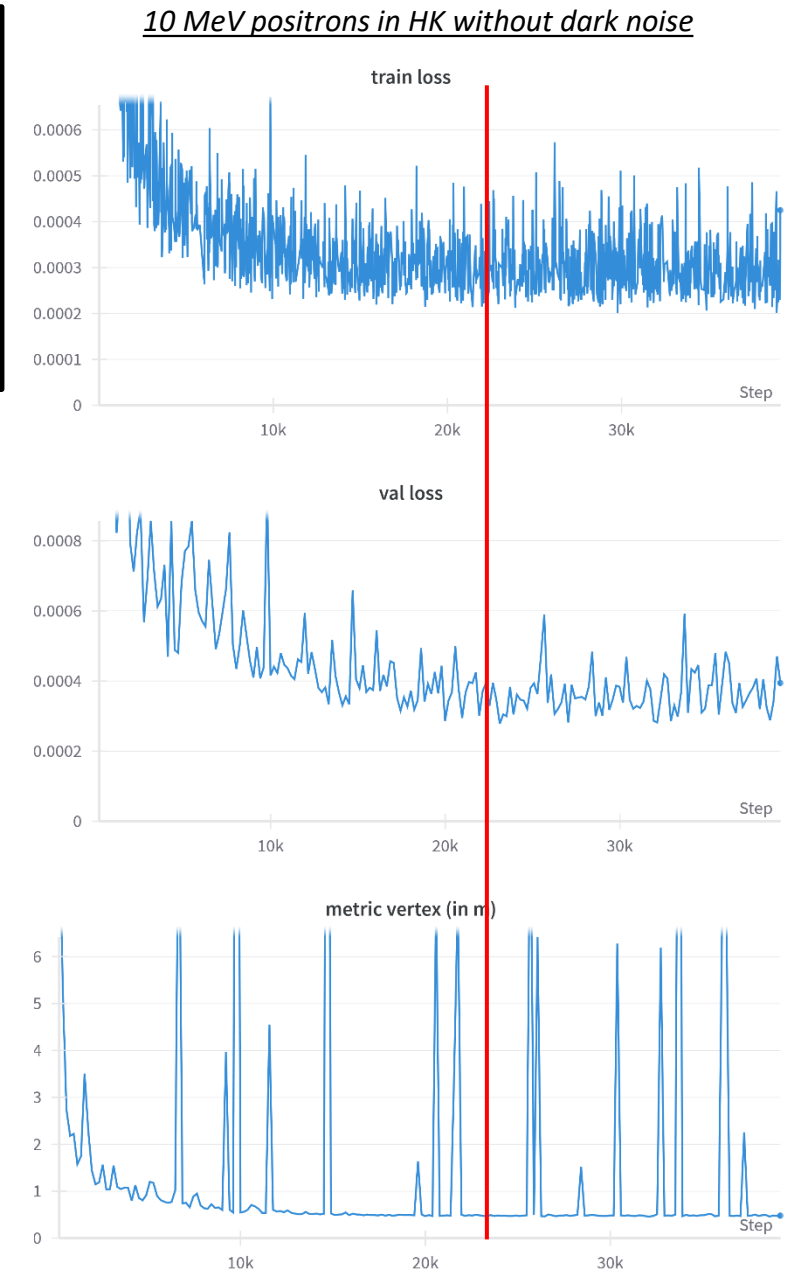


SKYNET-VTX: SLICE EDITION

- Map all events to \mathcal{P} then regression of mapped vertex
- Labels: map vertex in $\mathcal{P} \{x_{\mathcal{P}}, z_{\mathcal{P}}\}$
- Node features: PMT hits $\{X_{\mathcal{P}}, Y_{\mathcal{P}}, Z_{\mathcal{P}}, T\}$
- Convolution layers: DynamicEdgeConv
- Hidden and final layers: Bayesian
- Output: vertex $\{x_{\mathcal{P}}', z_{\mathcal{P}}'\}$
- Loss function: RMSE

⇒ **Best resolution @ 47.6 cm for RMSE**

Model summary	
Conv. Lay.	5 EdgeConv
Hid. Lay.	5 Bayesian
Edge attr.	# of hits
Loss fct	RMSE
Output	Vtx reco
Epoch	118
Reso. 68%	47.6 cm



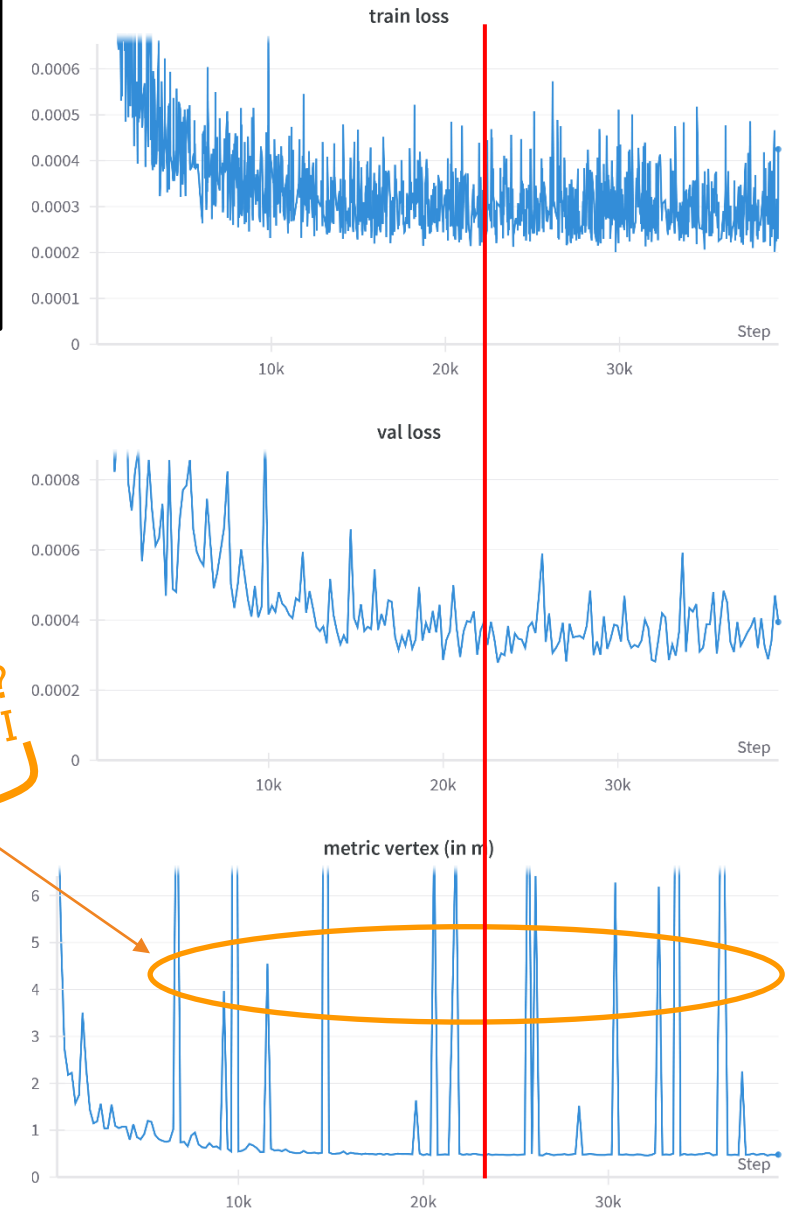
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 (... or do I have a bug ?)

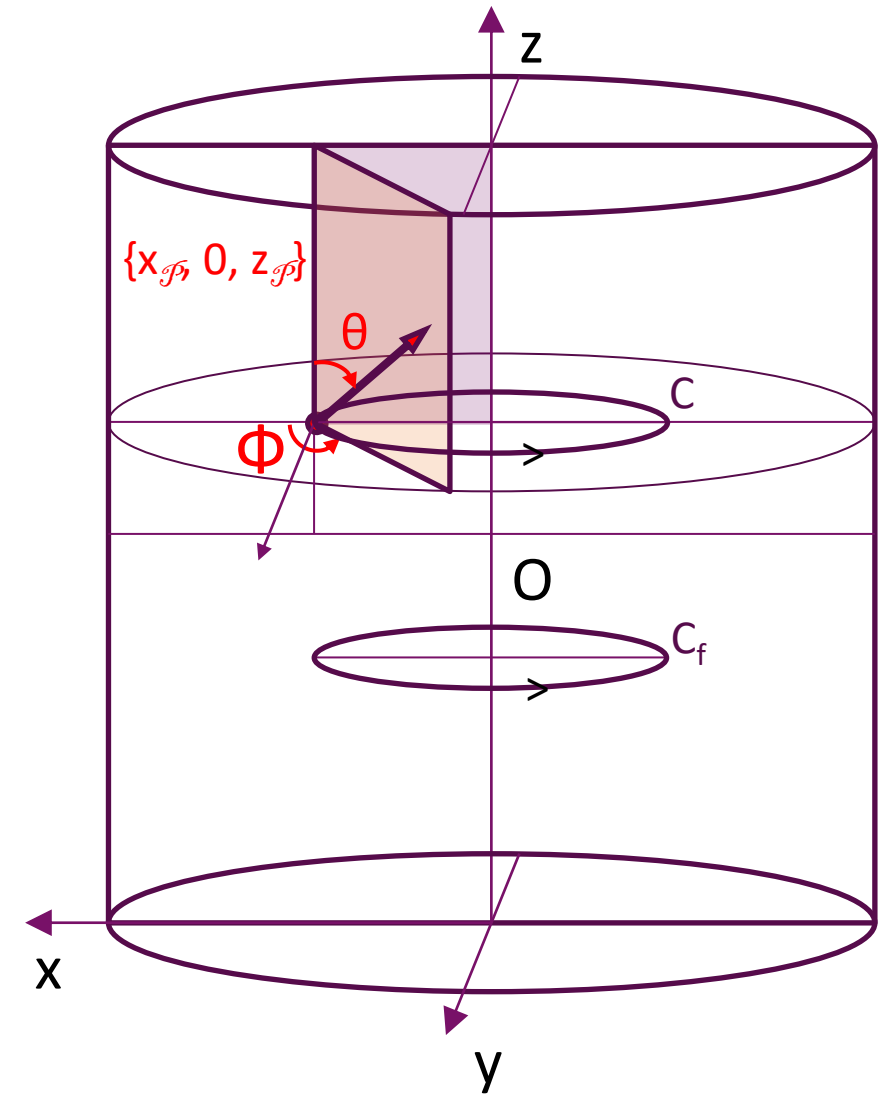
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10 MeV positrons in HK without dark noise



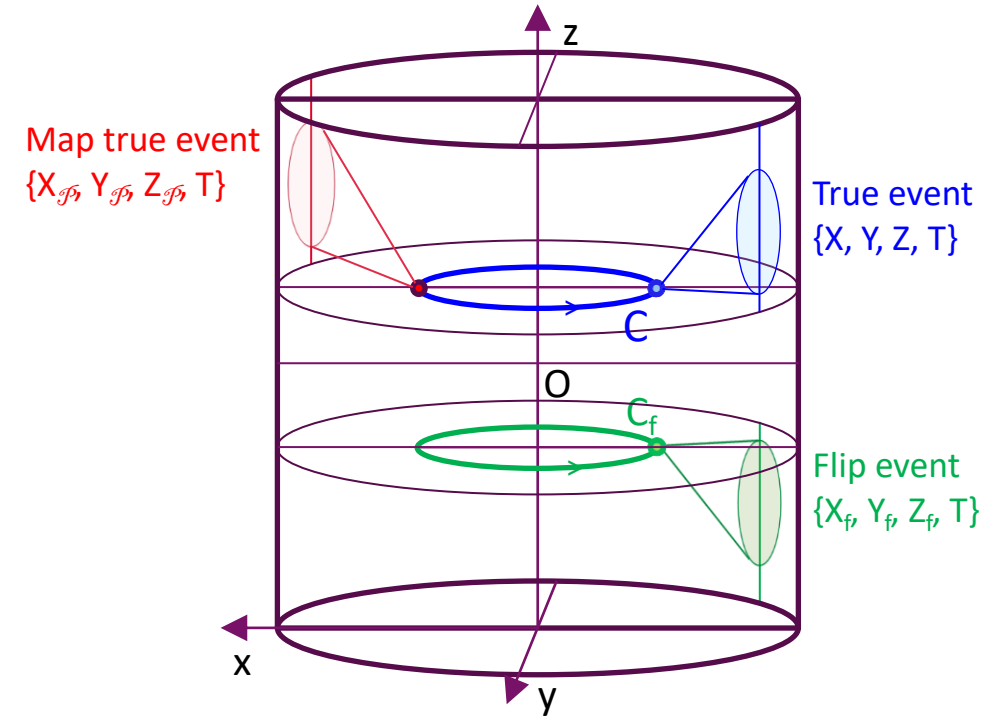
SLICE EDITION, MAIN ISSUES

- Reconstruct $\{x_{\mathcal{P}}, z_{\mathcal{P}}, \mathbf{t}\}$ and not $\{x, y, z, \mathbf{t}\}$
 - Additional step to find the best vertex over two circular paths C and C_f



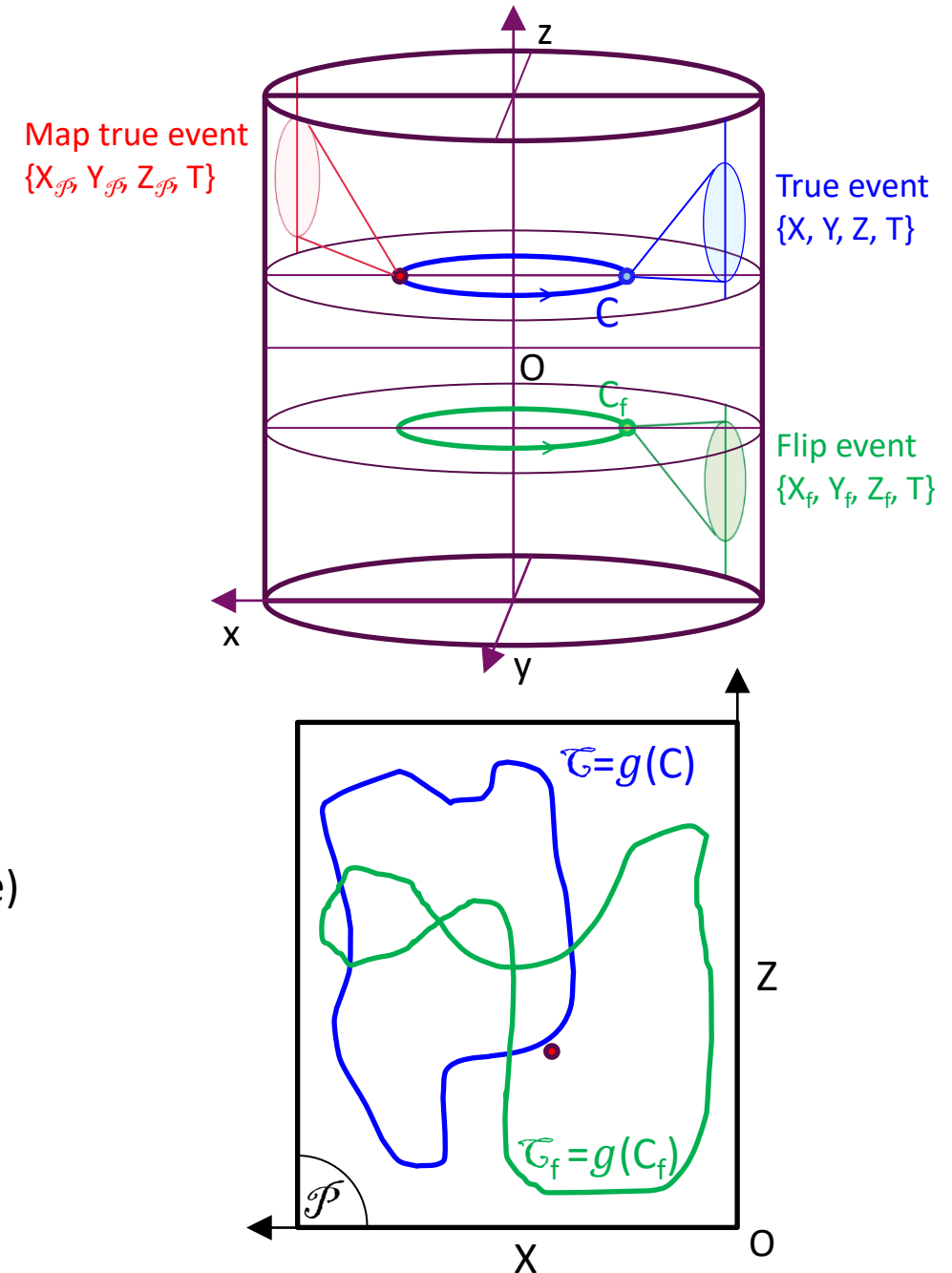
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- Real event PMT info $\{X, Y, Z, T\}$ and not $\{X_{\mathcal{P}}, Y_{\mathcal{P}}, Z_{\mathcal{P}}, T\}$
 - Not possible to directly map PMTs to \mathcal{P}
 - C : path generated by $\{X, Y, Z, T\}$ rotated from 0 to 2π
 - C_f : path generated by $\{X, Y, Z, T\}$ flipped around (Ox) then rotated from 0 to $2\pi \rightarrow \{X_f, Y_f, Z_f, T\}$



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- Define a function g over C and C_f : $g(\{X, Y, Z, T\}_\theta) \mapsto \{x_{\mathcal{P}}, z_{\mathcal{P}}, t\}_\theta$
 - $\mathcal{C}=g(C)$ and $\mathcal{C}_f=g(C_f)$ in \mathcal{P} are continuous (trivial proof if GNN function is continuous, proof by computation on a grid otherwise)
 - Best candidate vertex = minimize a continuous a continuous function f defined in \mathcal{P} i.e. over \mathcal{C} and \mathcal{C}_f
$$f(\theta) := \text{NLL} = \sum_{\text{PMT}(\theta)} -\log(P(t_{\text{PMT}(\theta)} - t_{\text{VTX}} - \text{TOF}))$$
 - In the slice \mathcal{P} , **best candidate vertex** \sim **map true event**



CONTINUOUS FUNCTIONS g, f IN \mathcal{P}

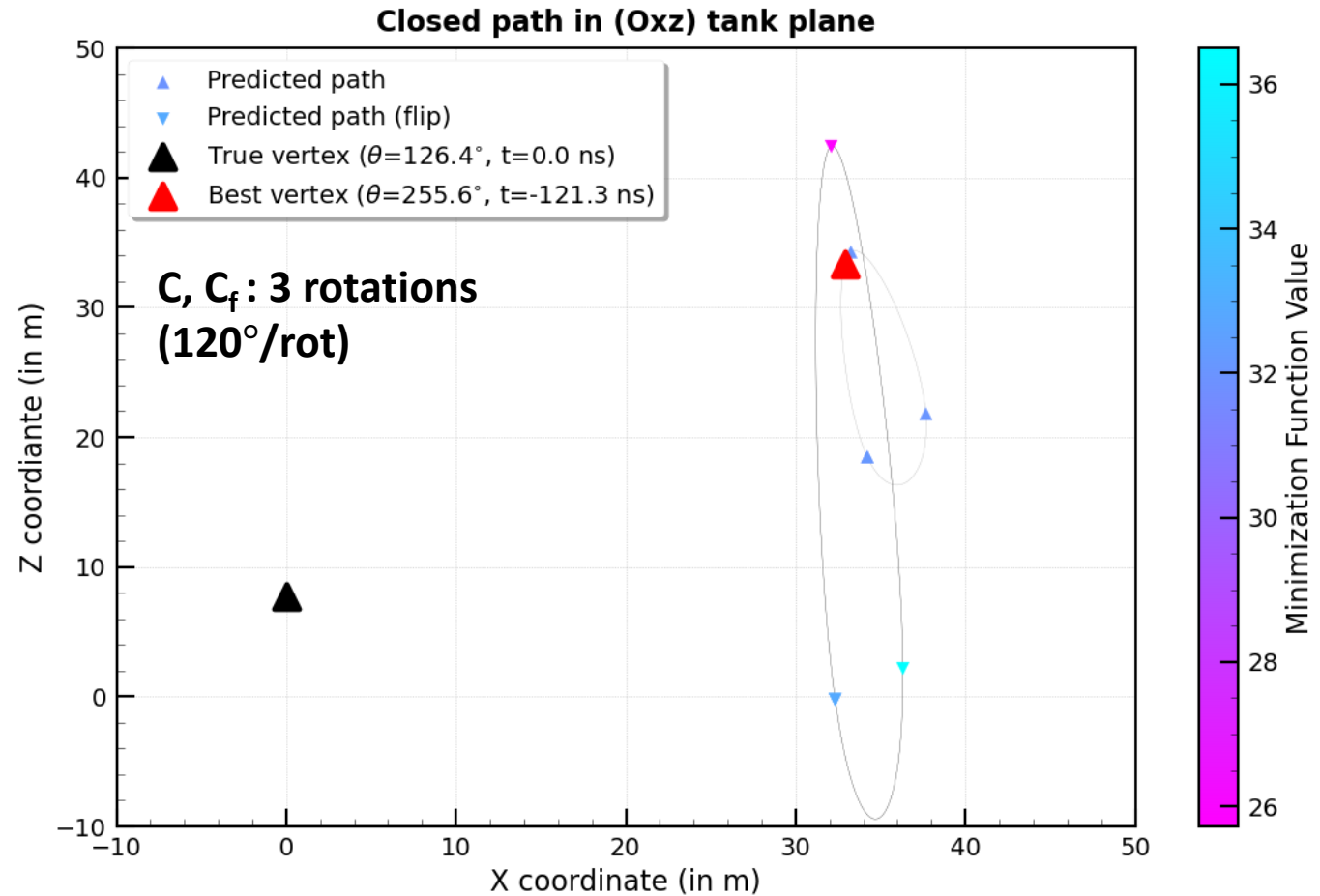
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- $\mathcal{C}=g(C)$ and $\mathcal{C}_f=g(C_f)$ in \mathcal{P} are continuous: **proof by computation on a grid**
- Best candidate vertex = minimize a continuous function f defined in \mathcal{P} i.e. over \mathcal{C} and \mathcal{C}_f

$$f(\theta) = \sum_{PMT(\theta)} -\log(P(t_{PMT(\theta)} - t_{VTX} - TOF))$$

- TO PROVE: In the slice \mathcal{P} , best candidate vertex \sim map true event



Note: slight variation of points due to Bayesian layers ; use mean GNN output over 100 iterations for each points, $\sigma(\text{mean}) \approx 10\%$

Note2: the model used here had a bug, could not patch it in time for the meeting, that's the reason why best vertex \neq map true event

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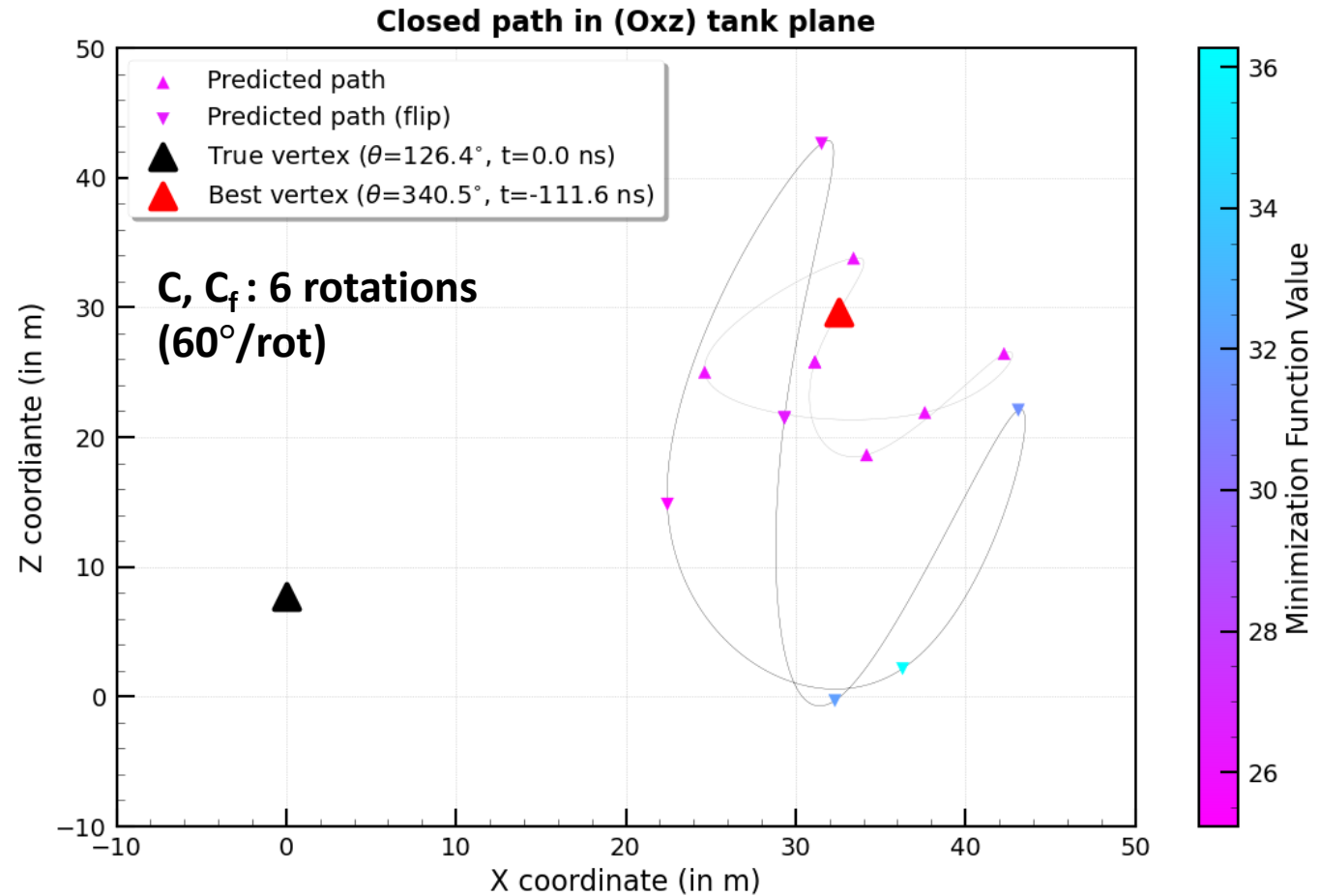
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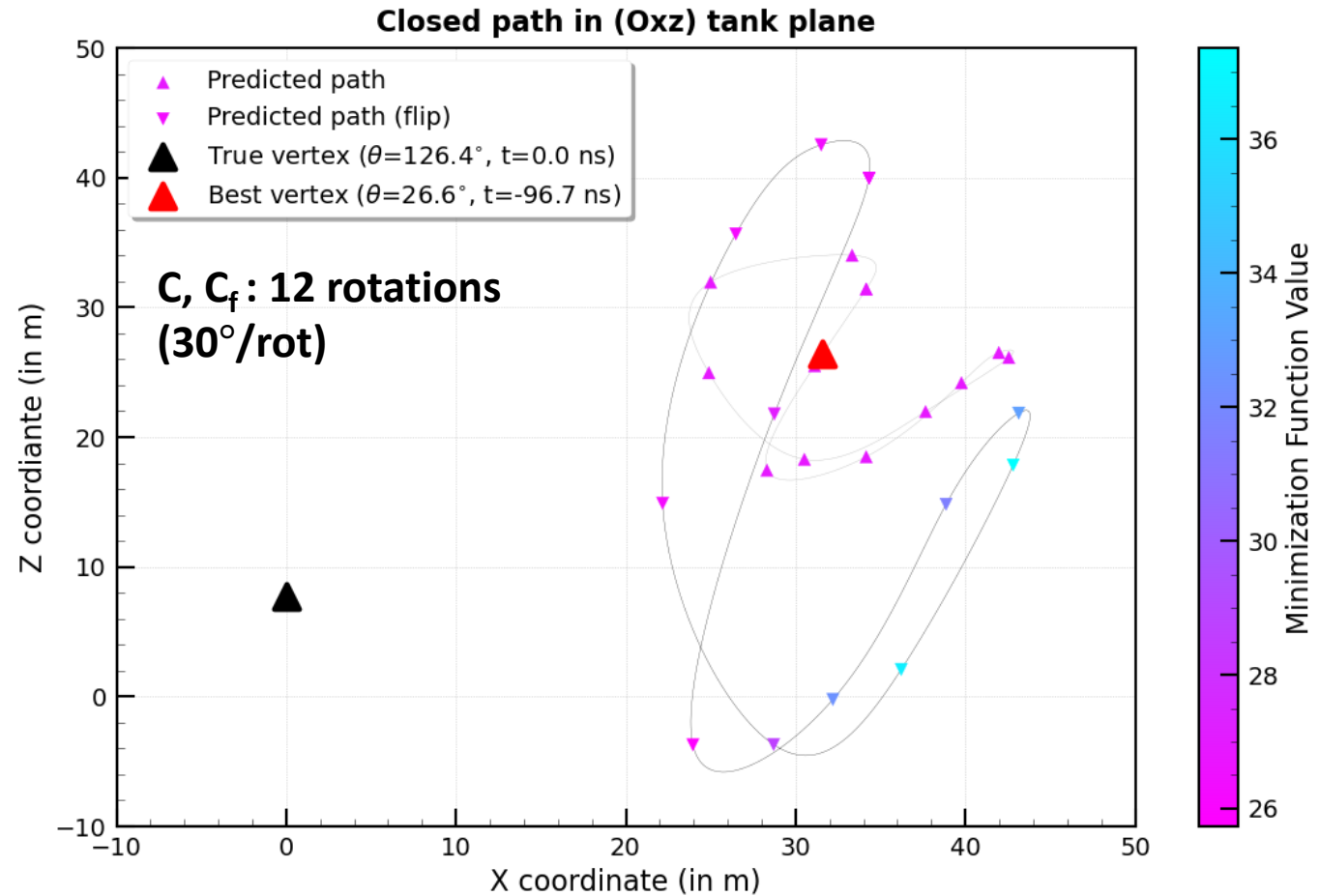
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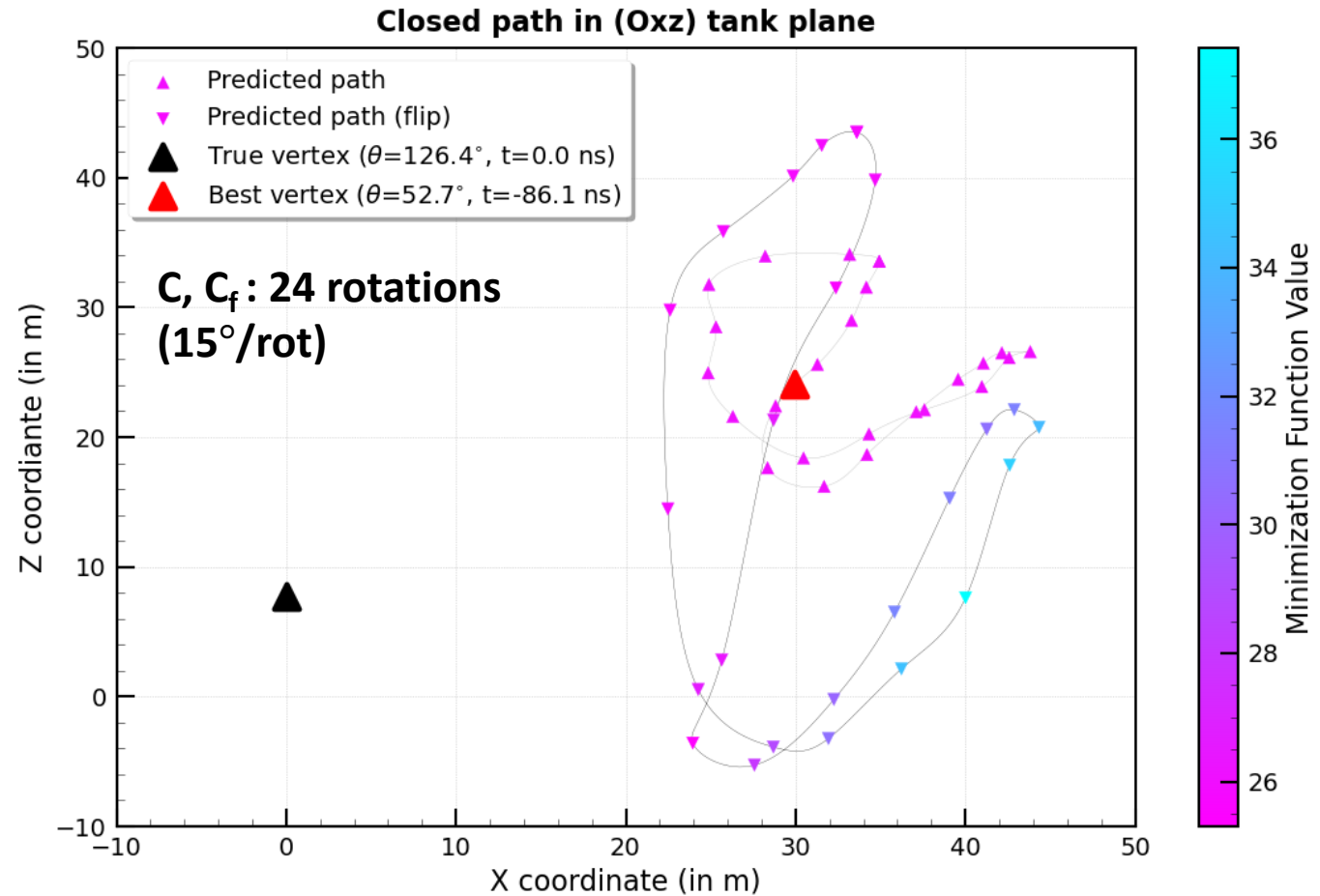
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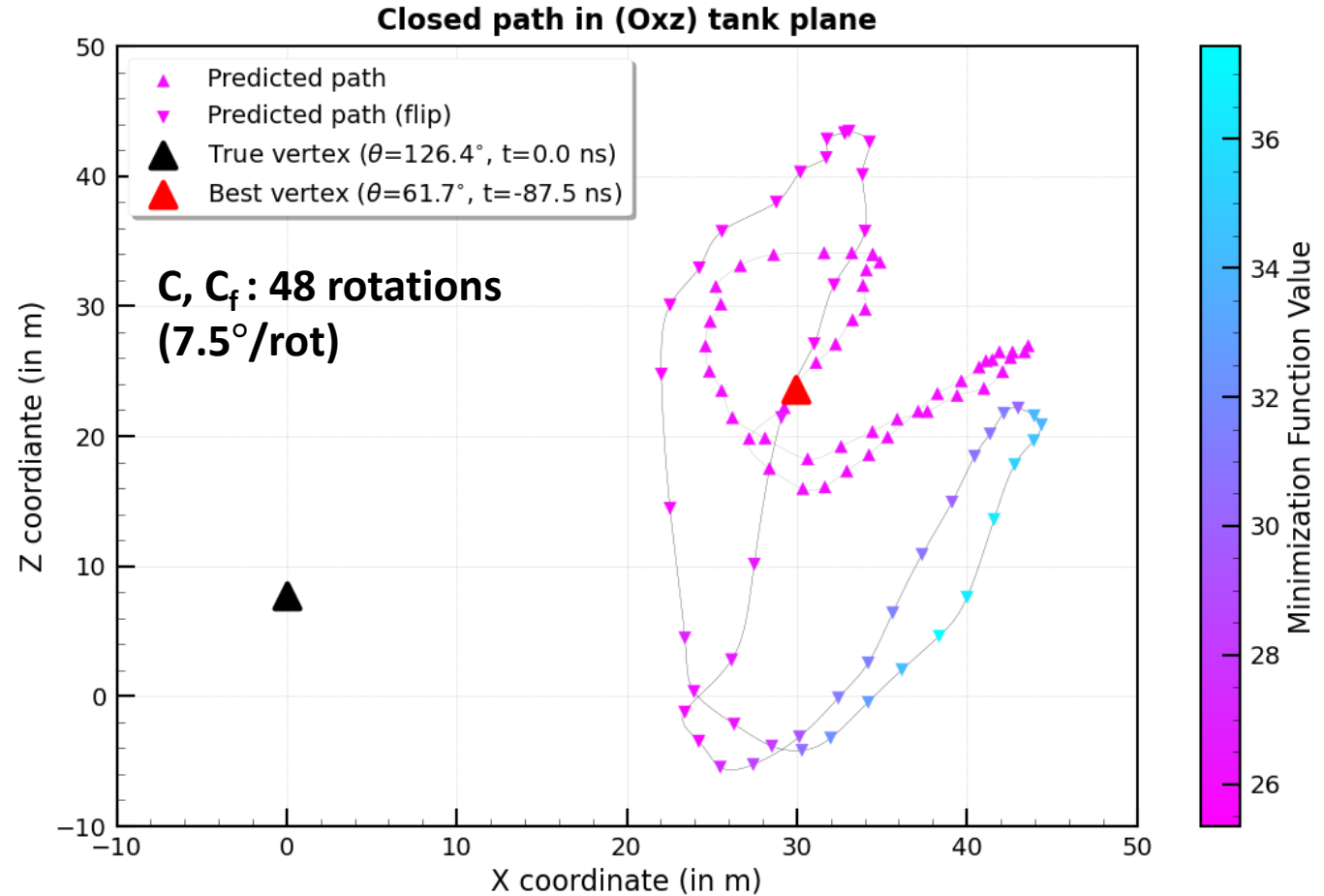
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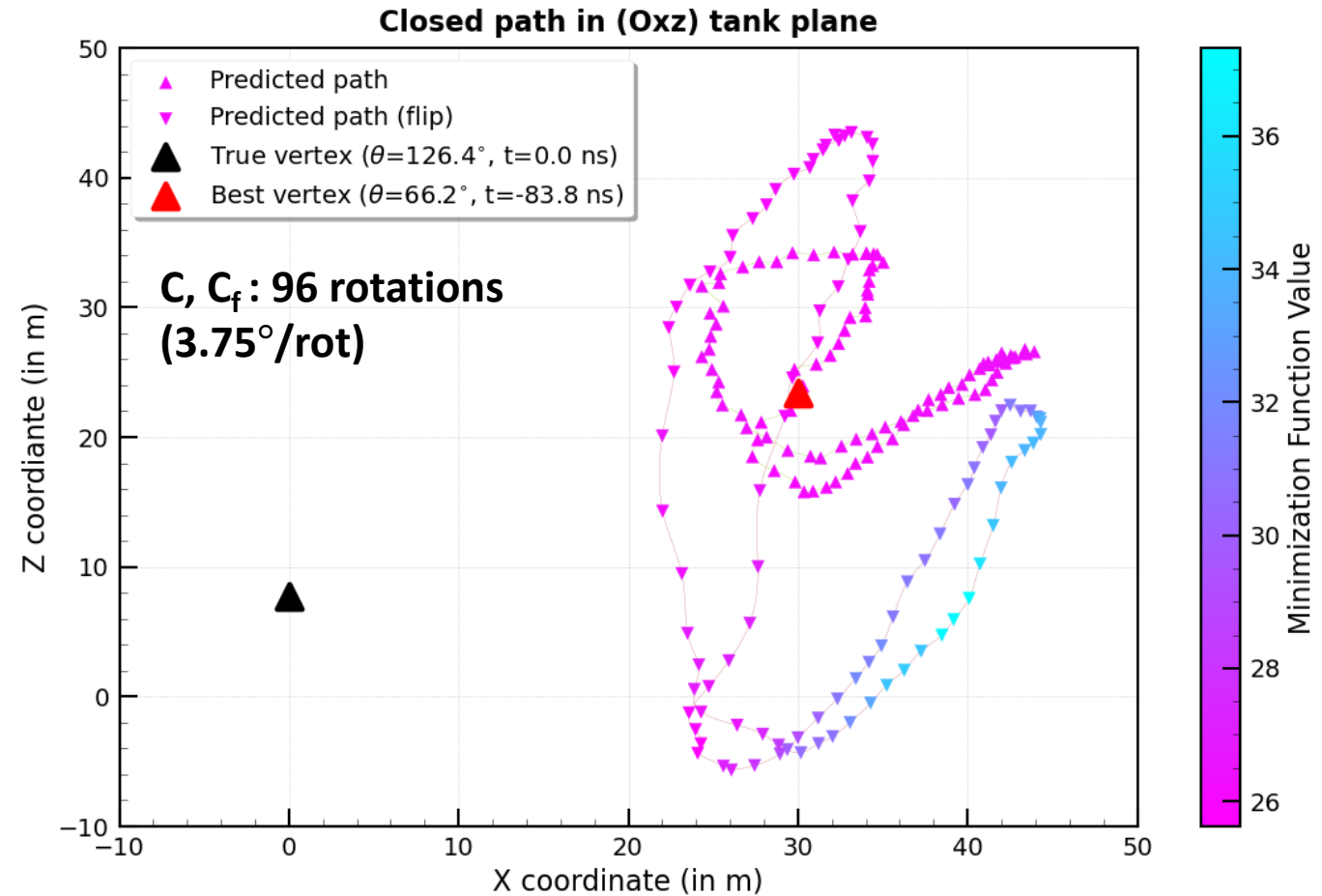
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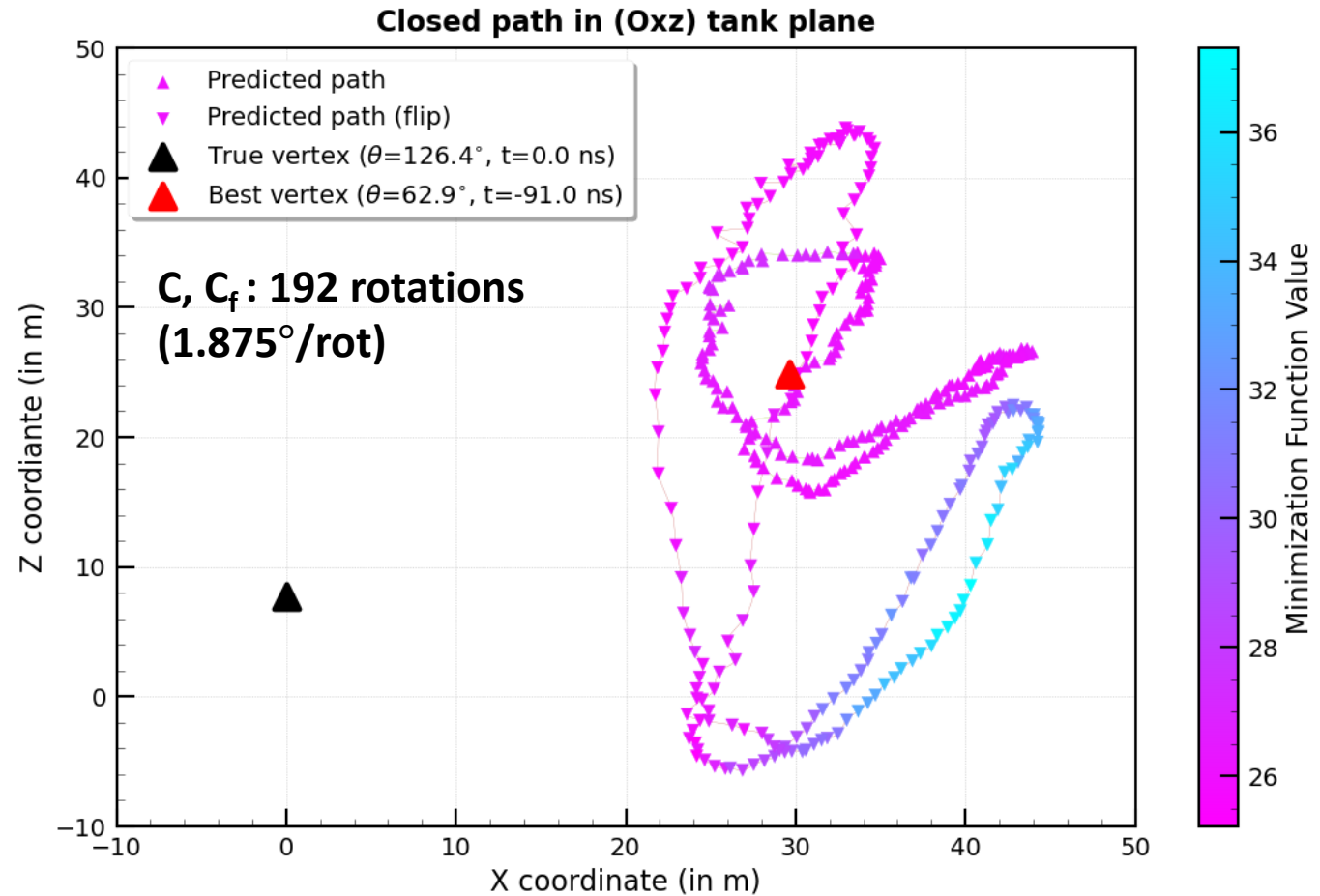
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3D → 2D vertex regression :

- Continuous behavior of mapping functions g and f checked
- Algorithm ready for 2D vertex regression + f -minimization in \mathcal{P}
- Waiting for CC-IN2P3 to run jobs

Next steps, 3 directions:

- **Optimization of resolution:** hyper parameters, GPU time, # rotation for C and C_f , # of interpol. points in $\mathcal{C}=g(C)$ and $\mathcal{C}_f=g(C_f)$, # of Bayesian iterations
 - Maybe to be done by interns in April if resolution + GPU time are satisfactory for first analysis ?
- **Development:** Add direction & energy (most of the algorithm is written, comparison + optimization needs to be done)
- **Analysis:** neutron classification, rework SKYNET-class for WCTE (use mPMT) + benchmark with Antoine version
 - WCTE neutron tagging analysis: mean path in Gd-loaded water (+ is resolution possible ?), tagging accuracy