



Next Generation Microchannel Plate Detectors for High Spatial and Temporal Resolution

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> Contents

- Characterisation of the TORCH 16x96 Multi Anode MCP-PMT
 - **Pulse shape**
 - **Crosstalk – count rate and gain**
 - **Single Photon Timing Jitter**
 - **CASE studentship with University of Warwick – Alex Davidson**
- Development of a Charge Sharing Photon Timing/Imaging Detector
 - **Instrument a capacitively coupled multi-anode readout using TOFPET2d ASIC**
 - **Time-over-threshold discriminator, non-linear charge measurement**
 - **Can we avoid complicated calibration using Neural Networks?**
 - **Can we even do it faster?**
 - **Royal Commission for 1851 Industrial Fellowship with University of Leicester – Amelia Markfort**

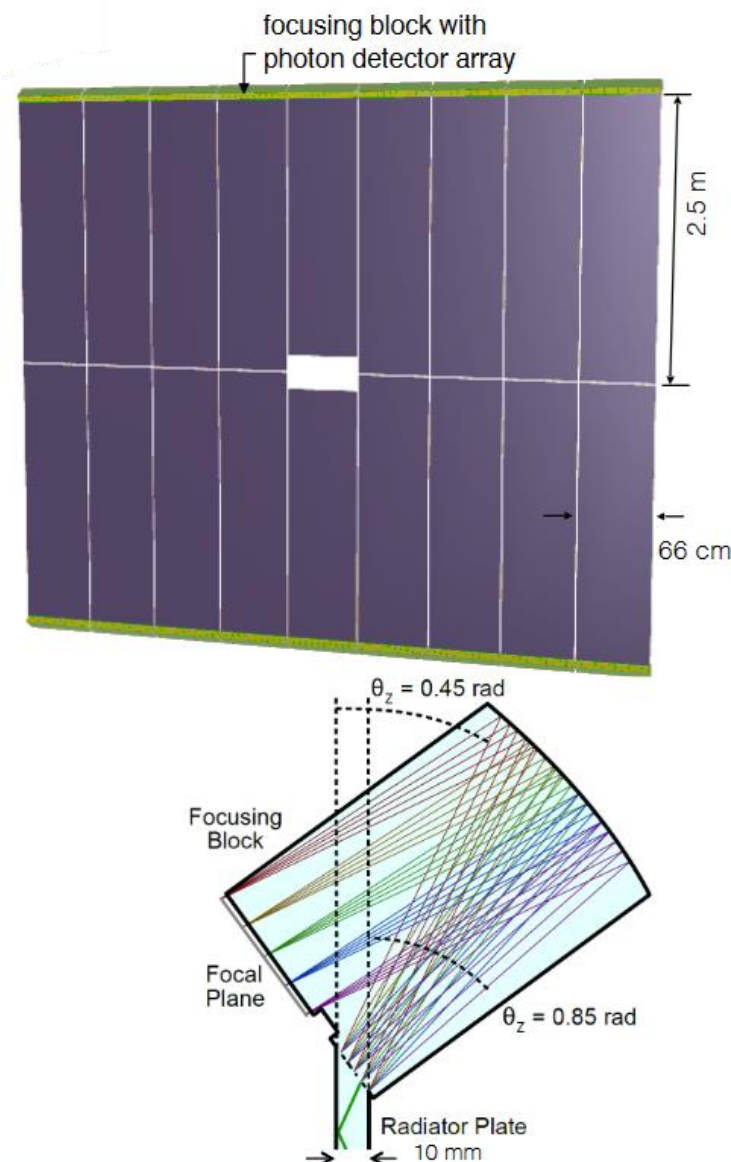
TORCH MCP Detector

High density multi-anode MCP Photon detector characterisation

> TORCH Cerenkov Particle ID Detector Concept

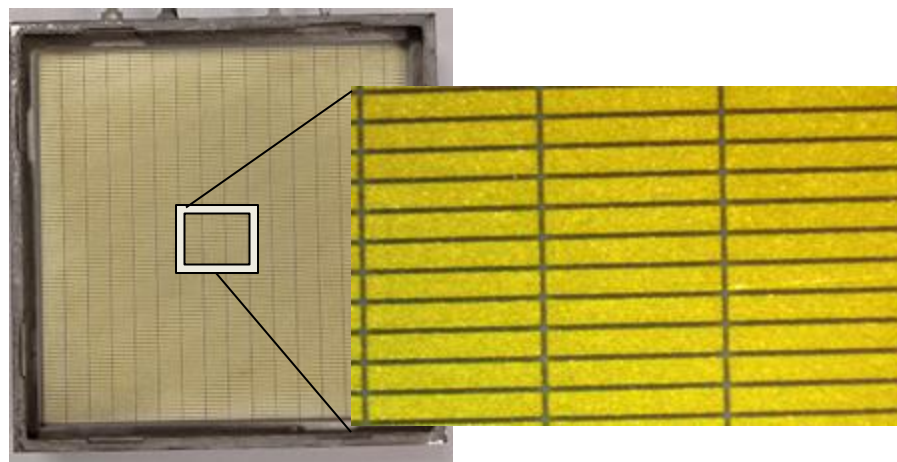
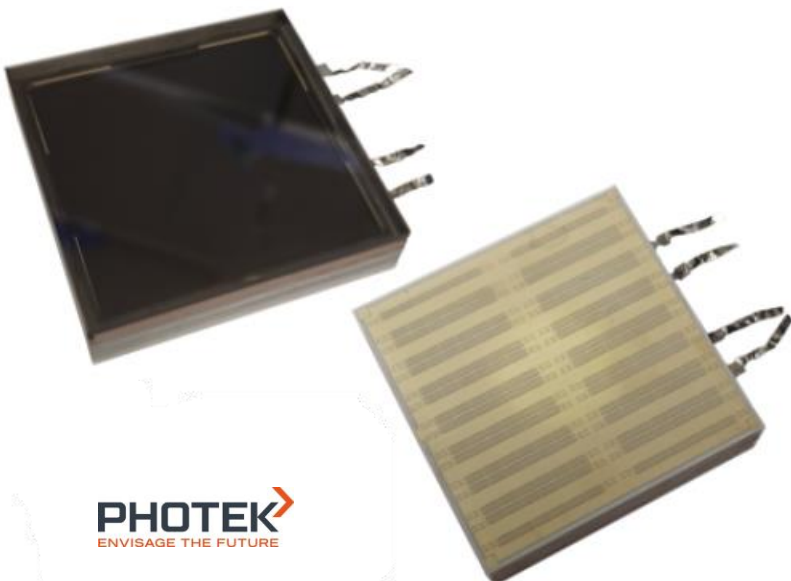
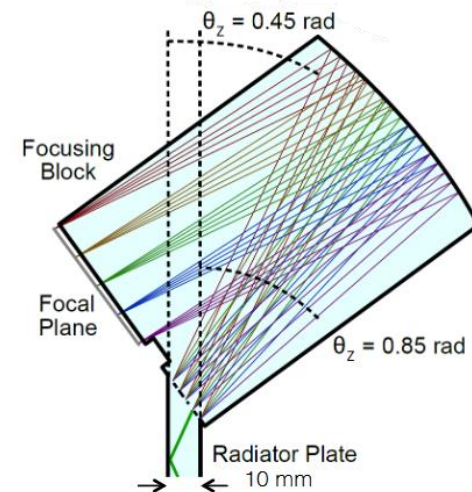
- Exploit prompt production of Cherenkov light in an array of fused-silica bars to provide timing.
- Cherenkov photons are propagated to detector plane via total internal reflection from the quartz surfaces.
- Cylindrical focussing block, focusses the image onto a detector plane with highly segmented photon detectors.
 - ▶ Used to correct for chromatic dispersion.
- Large area detector required to cover the full LHCb acceptance ($5 \times 6 \text{m}^2$).

For more details on the TORCH concept see [\[NIM A 639 \(1\) \(2011\) 173\]](#)

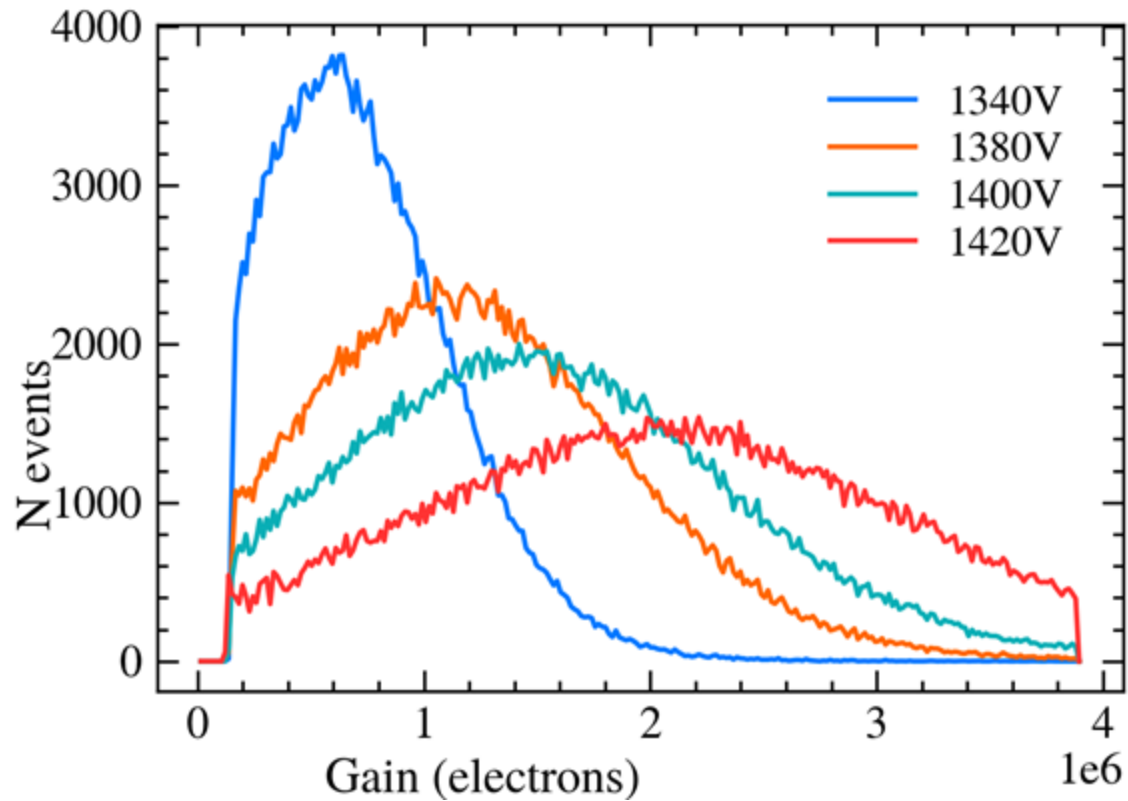


➤ TORCH 16×96 Multi-anode MCP Detector

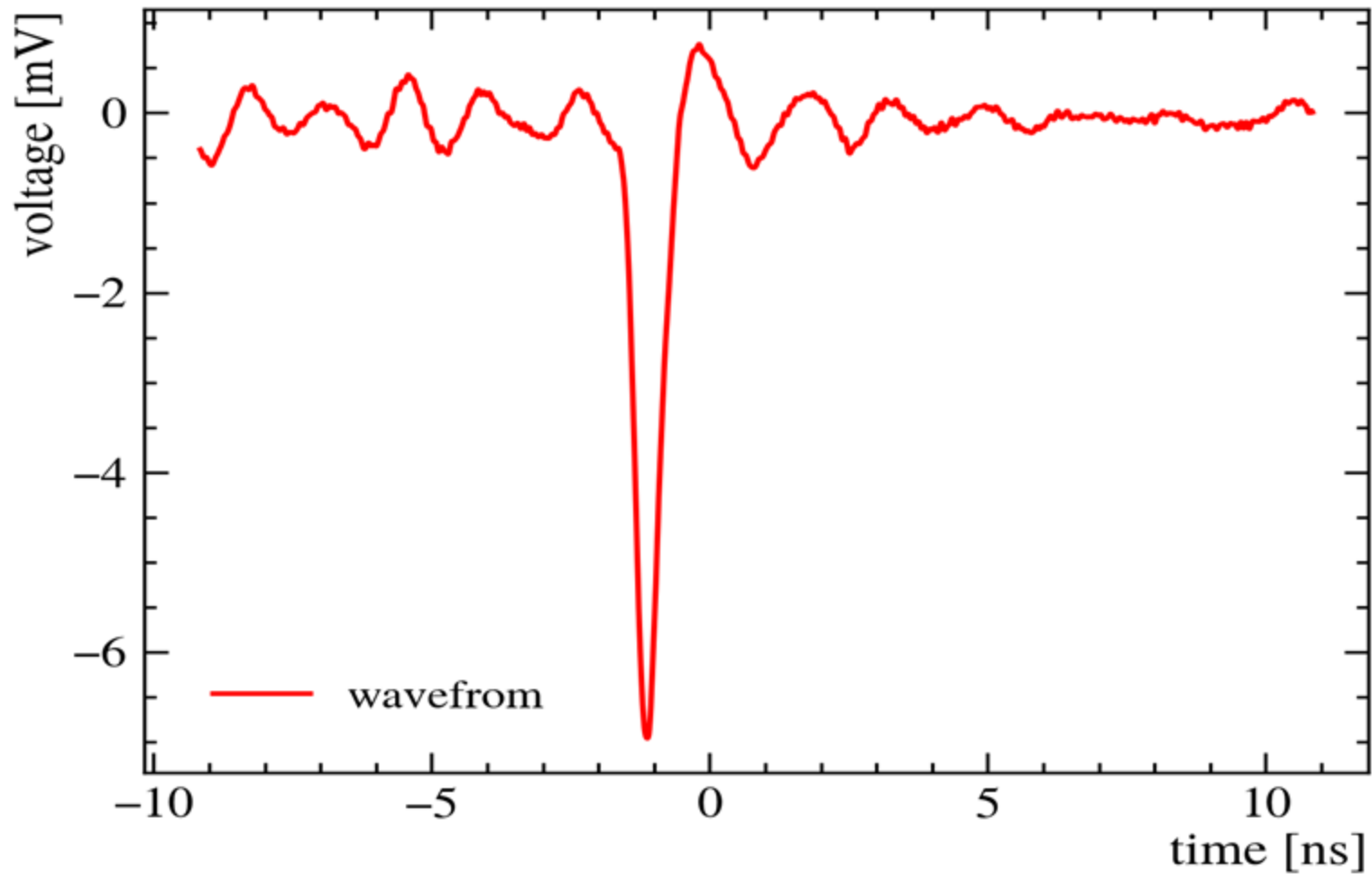
- 16×96 Anode layout, 53×53 mm² active area
- Fine pitch required in one direction
 - 552 μm pitch
 - Measure angle – Cerenkov/chromatic aberration correction
 - Also reduces occupancy for high rates
- In coarse direction 3312 μm pitch



Global Gain Measurement – All pixels connected



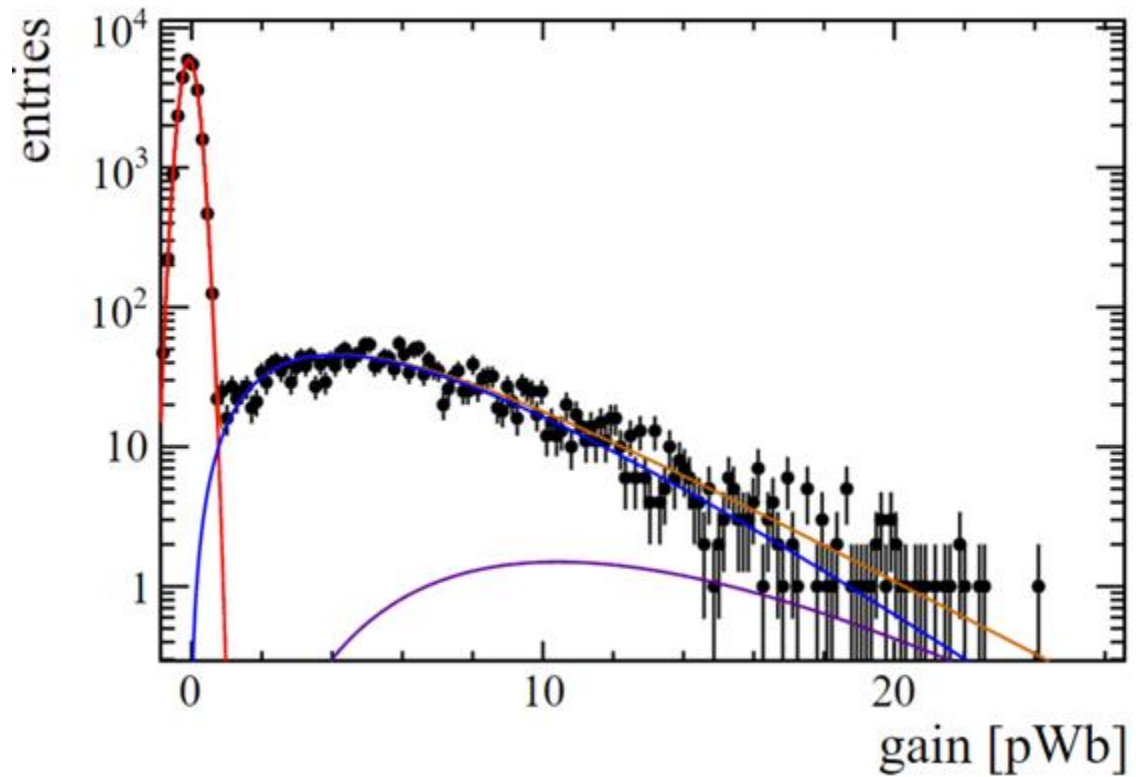
➤ Single Pixel – Average Single Photon Pulse Shape



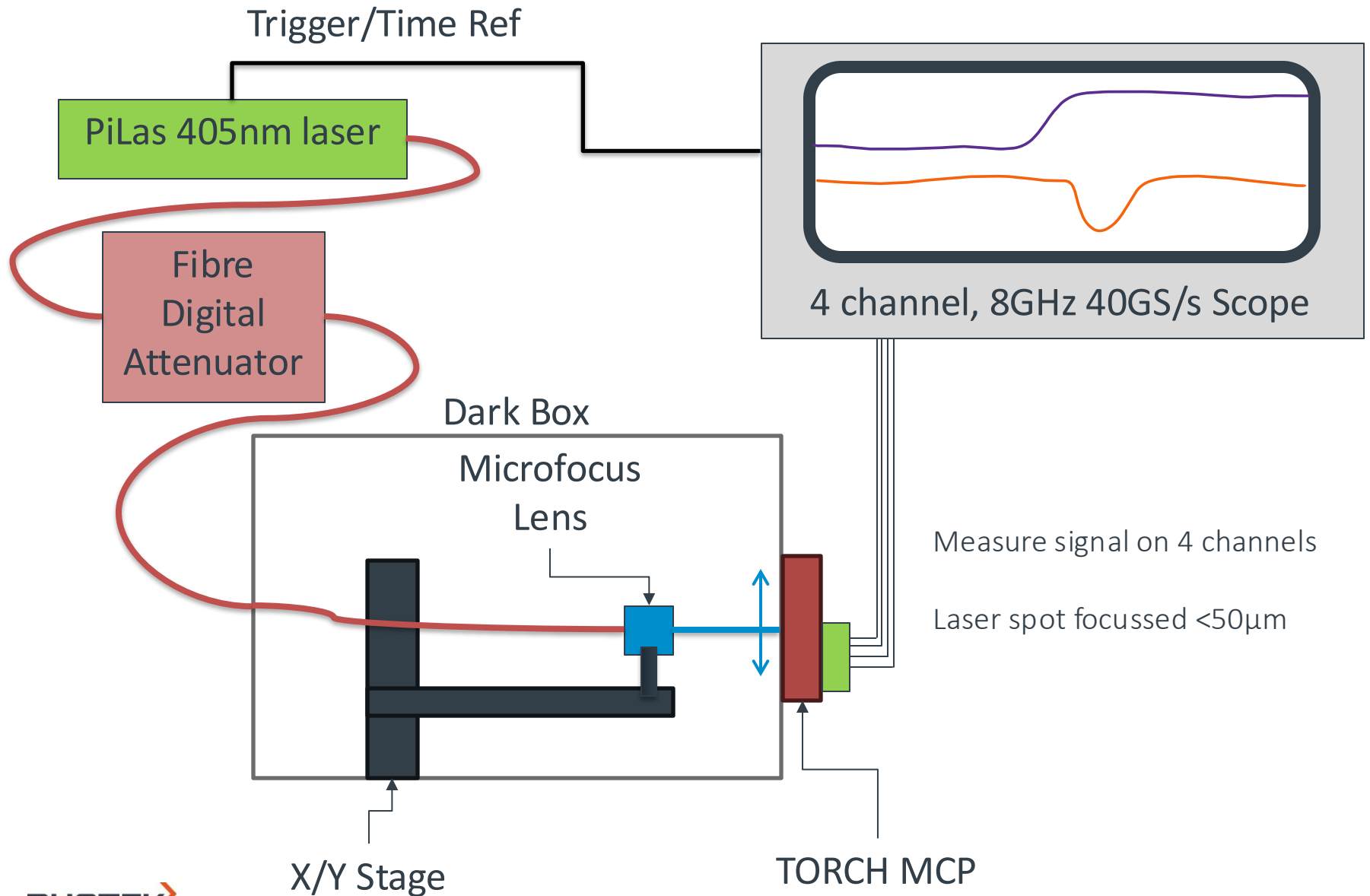
NB: Ripple due EM pickup from laser

➤ Single Pixel Gain – Pulse Area (on oscilloscope)

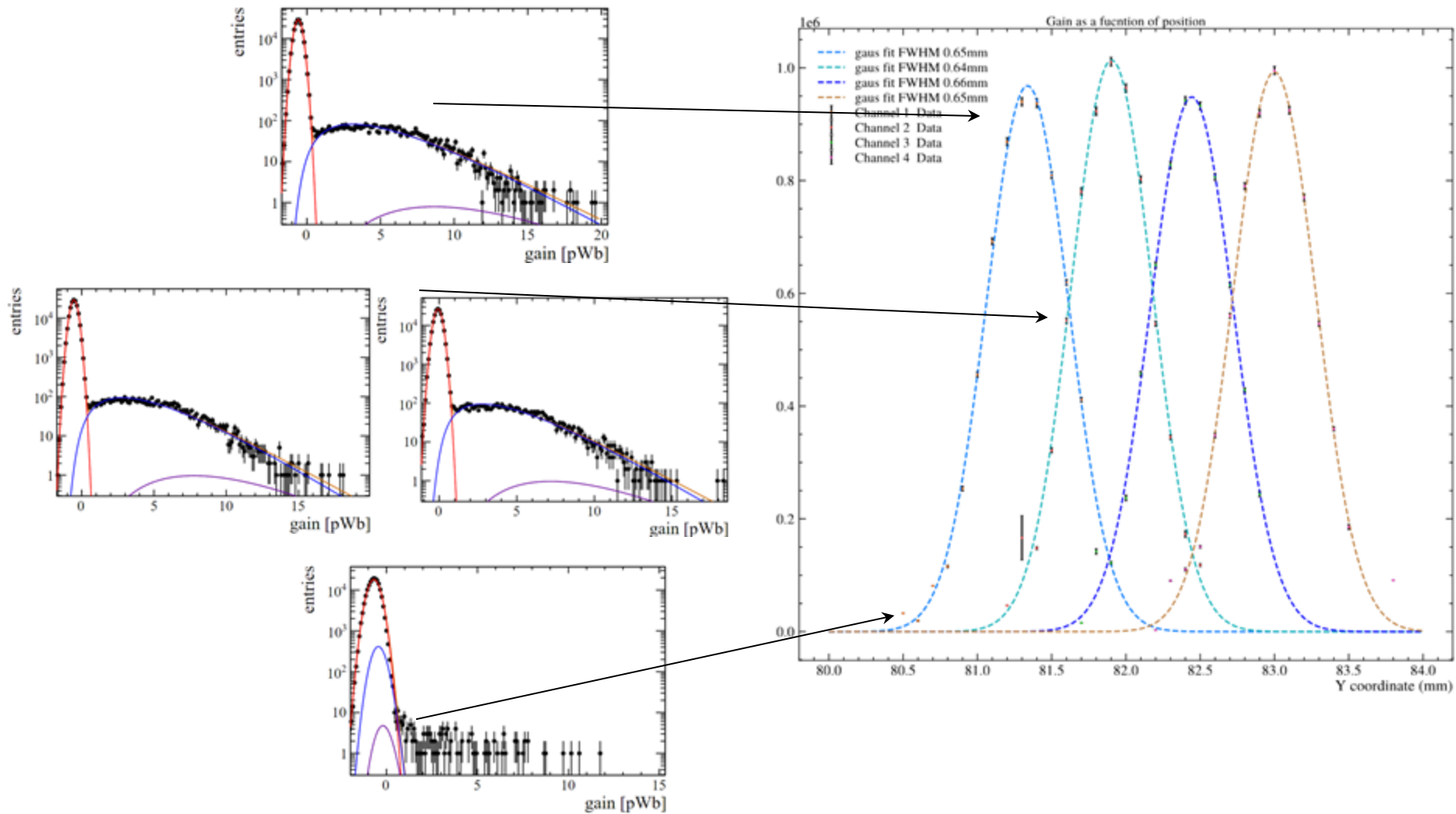
- Trigger on each laser trigger
- Integrate pulse area for single photon events, on a single pixel
- Histogram to produce Pulse Height Distribution (PHD) to characterise gain
- Gaussian (noise floor) + multi-Polya fit to find gain
 - 0 photon
 - 1 photon
 - 2 photon
 - Etc....



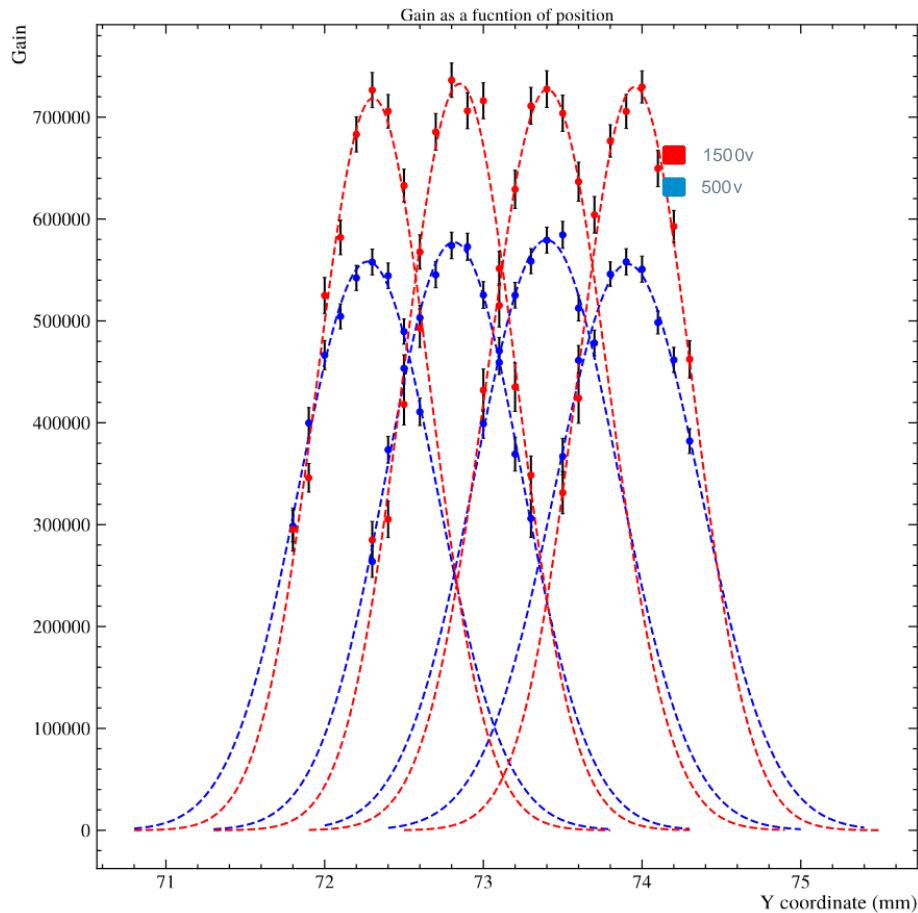
Single Photon Scanning - Crosstalk/Timing Measurement setup



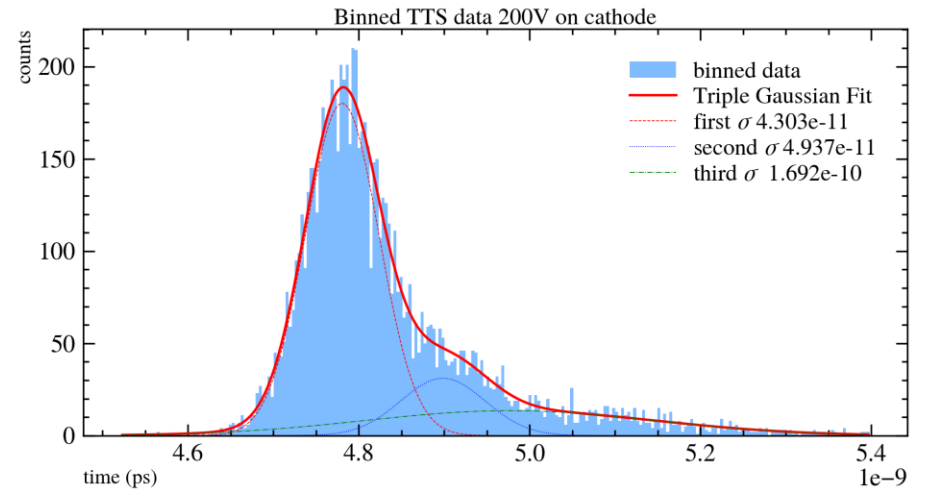
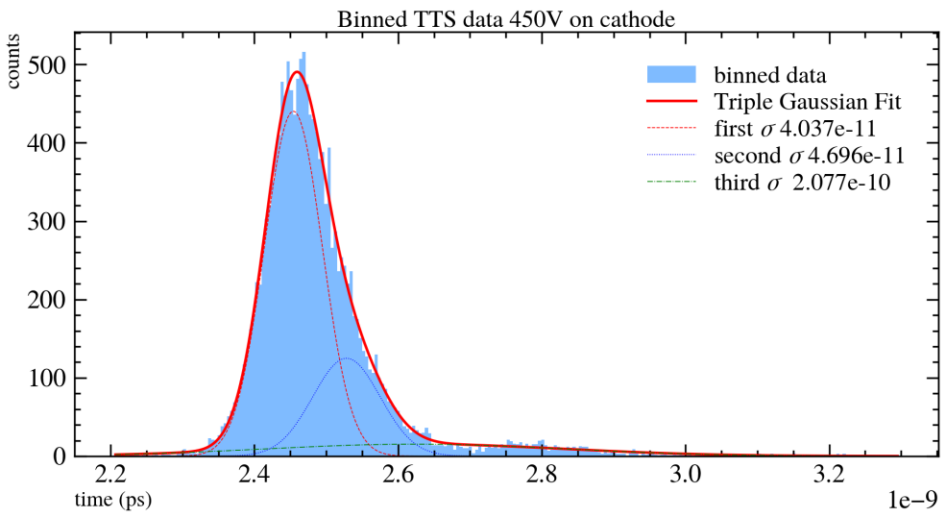
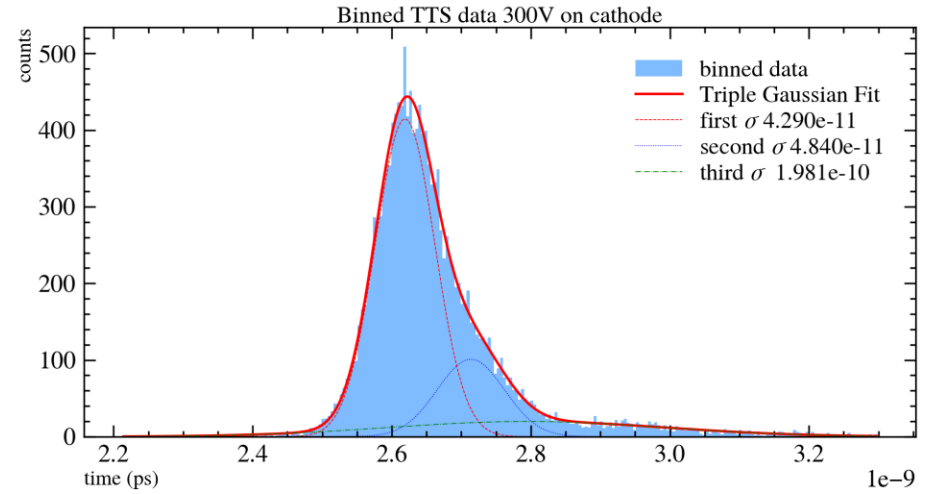
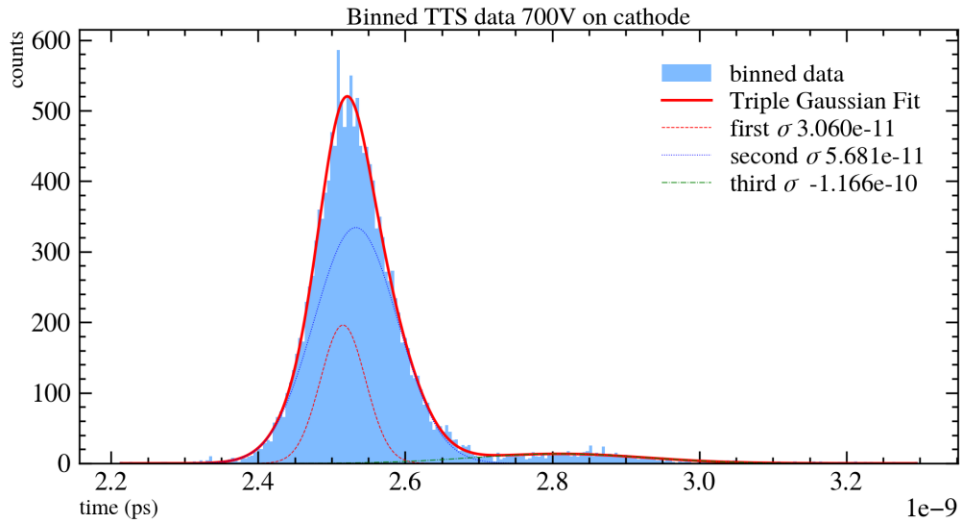
➤ Scope Measured – Gain crosstalk



➤ Increasing rear field – MCP output to anode



	@500V	@1500v
Channel 1	1.02 mm	0.84 mm
Channel 2	1.00 mm	0.84 mm
Channel 3	1.06 mm	0.90 mm
Channel 4	1.07 mm	0.85 mm



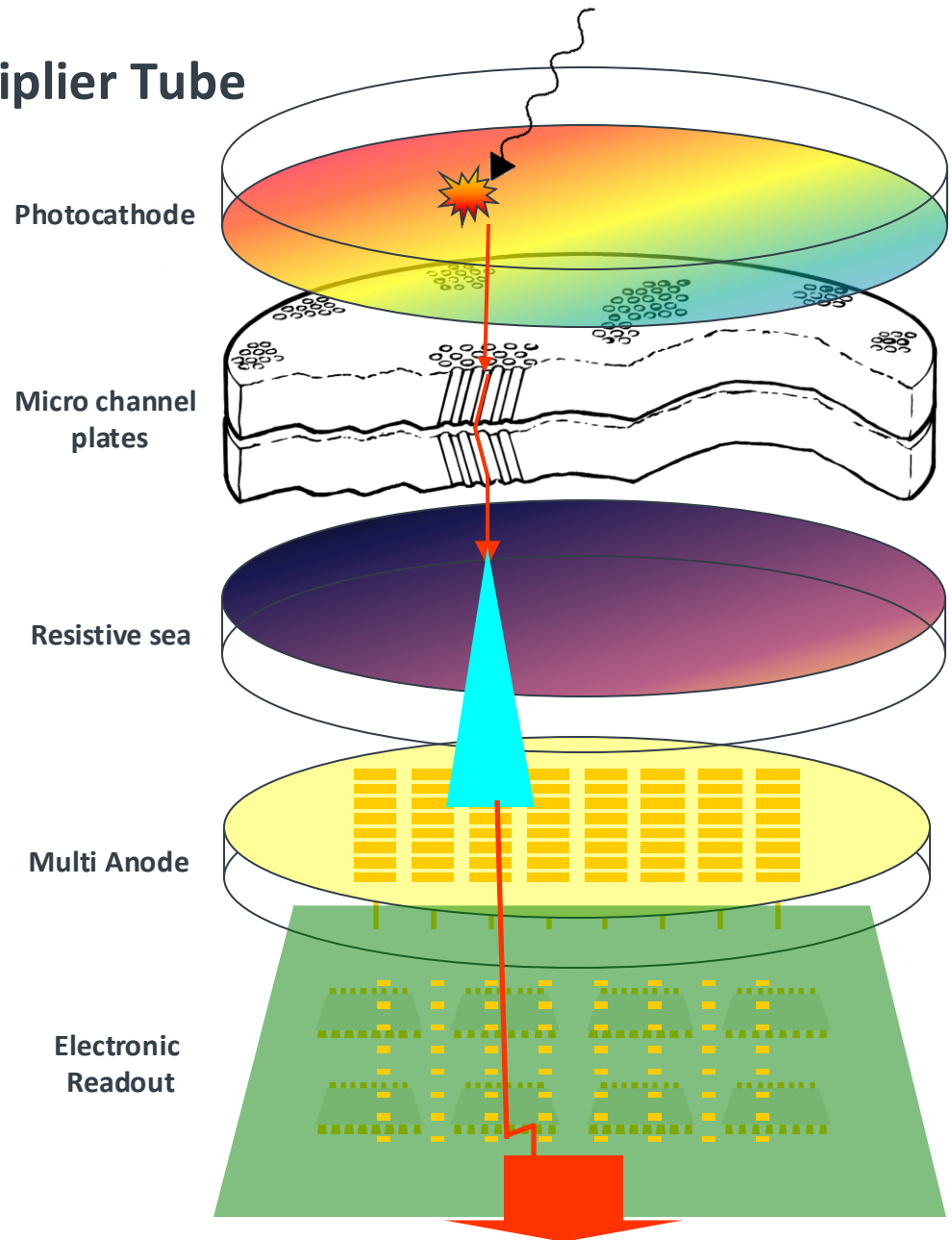
Charge Sharing Photon Timing/Imaging Detector

Neural network imaging for ToT charge measurement

➤ Microchannel based Photo-Multiplier Tube

Multi- Anode Photon Multiplier Tube

- Photocathode to convert photon to an electron
 - Photoelectron accelerated toward micro-channel Plate (MCP)
- MCP amplifies photoelectron
 - Electron enters pore
 - Hits wall produces secondary electrons
 - Multiple “bounces” creates gain (up to a factor of 10^7)
- Collected by a readout
 - Capacitively coupled to Multi anode → Multi-channel current pulse → Tofpet2d Electronics



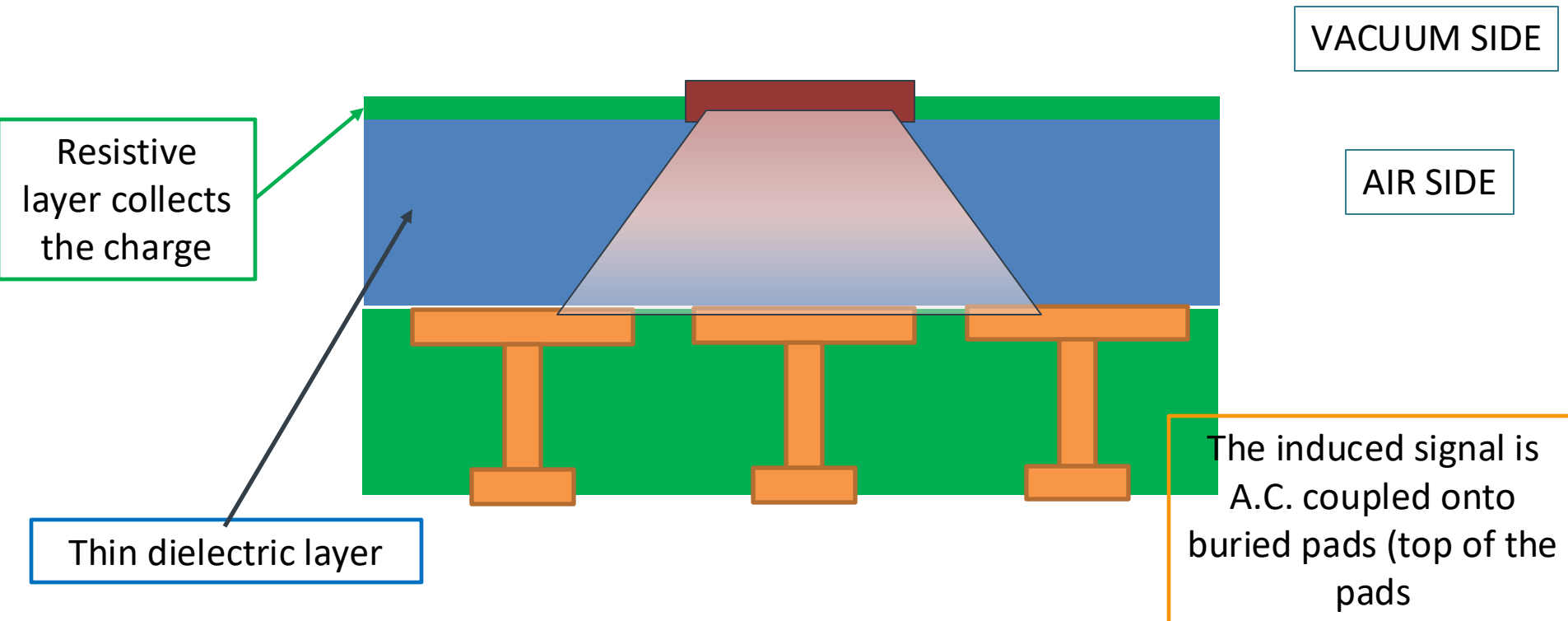
> Next generation of Auratek PCS: Charge sharing

Electronic output:

Each pixel has its own ToT discriminator

- Digitally time stamp
- Charge measurement via Time over threshold

Capacitively Couple MCP charge cloud to multiple readout pads

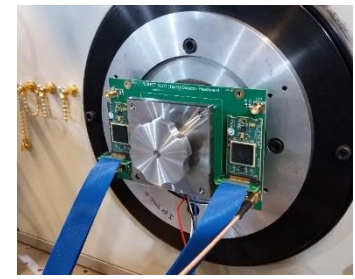


➤ Experimental data testing empirical detector data

For training data perform a fine pitch scan of the detector area, using pulse laser

Photons are detected, and produce multi channel events in TOFPET2d

Further processed to cluster photon events coincident with laser



Example of electronics set up outside the dark box

A single charge distribution:

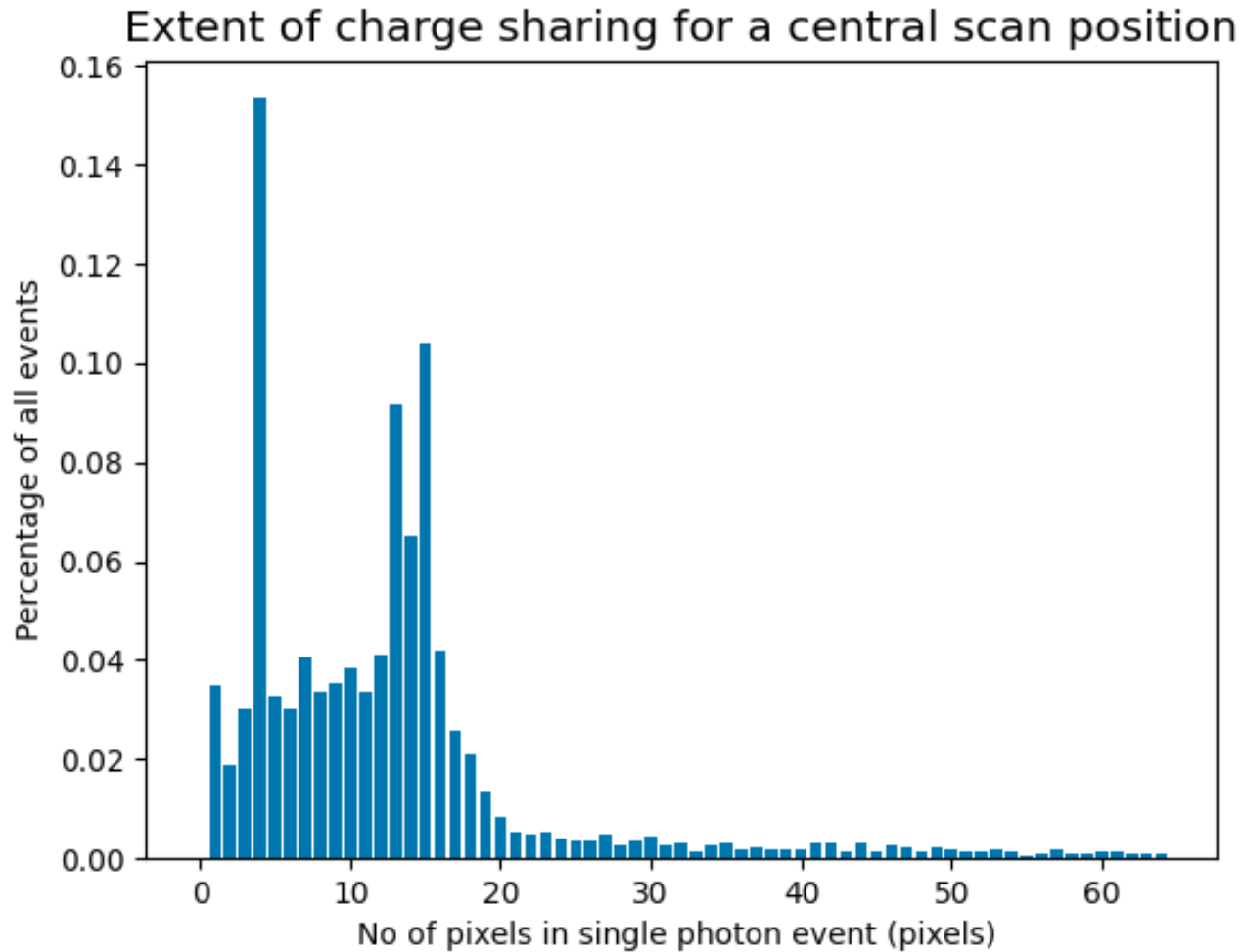
What the neural network sees as input:

X (in pixels)	Y (in pixels)	TOT (charge)	Time 2 (in ps)	Time 1 (in ps)
X ₁	Y ₁	TOT ₁	T ₁	T _{laser}
X ₂	Y ₂	TOT ₂	T ₂	T _{laser}
X ₃	Y ₃	TOT ₃	T ₃	T _{laser}
X ₄	Y ₄	TOT ₄	T ₄	T _{laser}
X ₅	Y ₅	TOT ₅	T ₅	T _{laser}
X ₆	Y ₆	TOT ₆	T ₆	T _{laser}

x 10,000
for a scan
position

Expected output: X_{laser} (mm), Y_{laser} (mm), T_{laser}

➤ Charge Sharing – Number of TOFPET2d channels over threshold



➤ Analytically spatial reconstruction of empirical detector data

- The Convolutional neural network is trained a single sample representing a charge distribution giving information:

- $64 \times (X_{p'}, Y_{p'}, Q_{p'})$

- to give outputs of:

- X_γ and Y_γ .

Each $(X_\gamma$ and $Y_\gamma)$ represents the spatial photon coordinates for each charge distribution

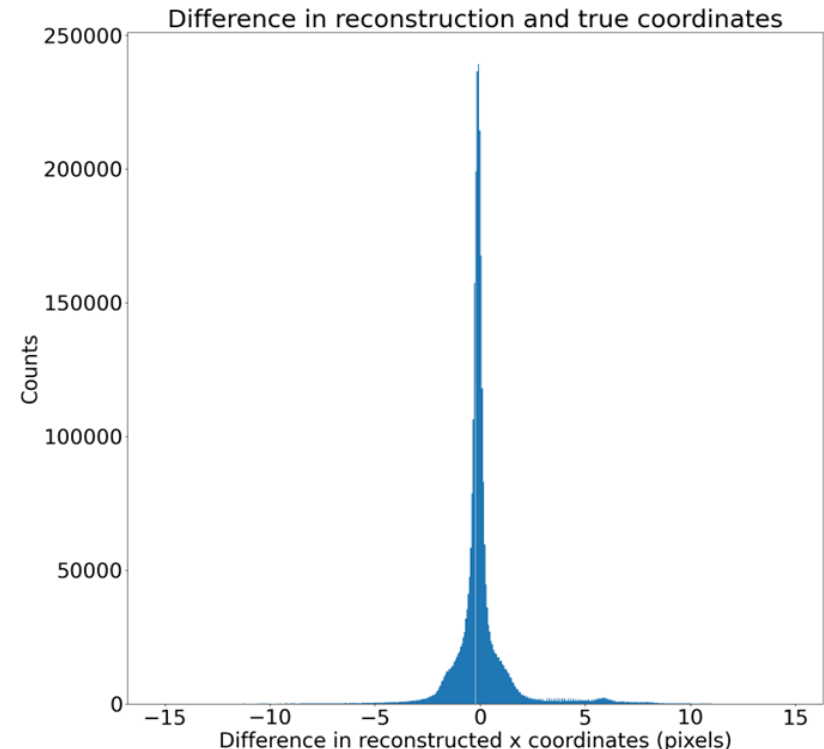
Each $(X_{p'}, Y_{p'}, Q_{p'})$ represents a packet of data from a single anode pad.

CENTROIDING ALGORITHMIC METHOD

$$X_\gamma = \sum \frac{x_p q_p}{\sum q_p},$$

$$Y_\gamma = \sum \frac{y_i q_p}{\sum q_p},$$

$$T_\gamma = \bar{t}_i$$

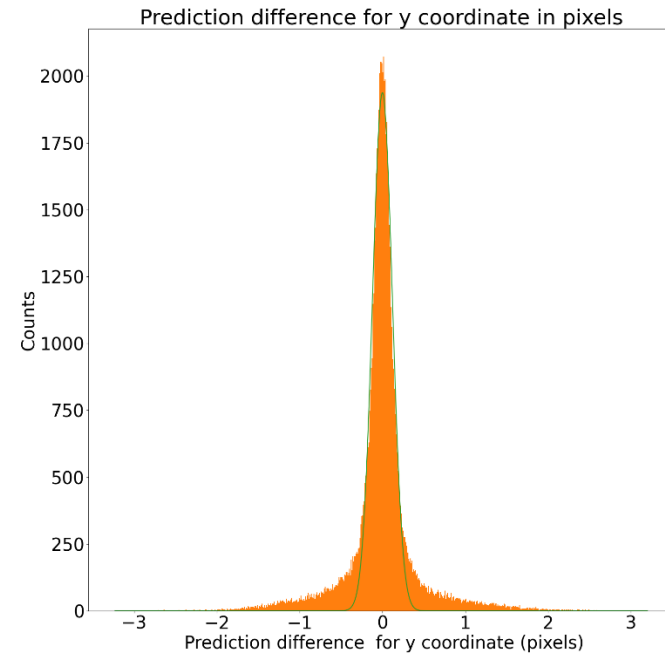
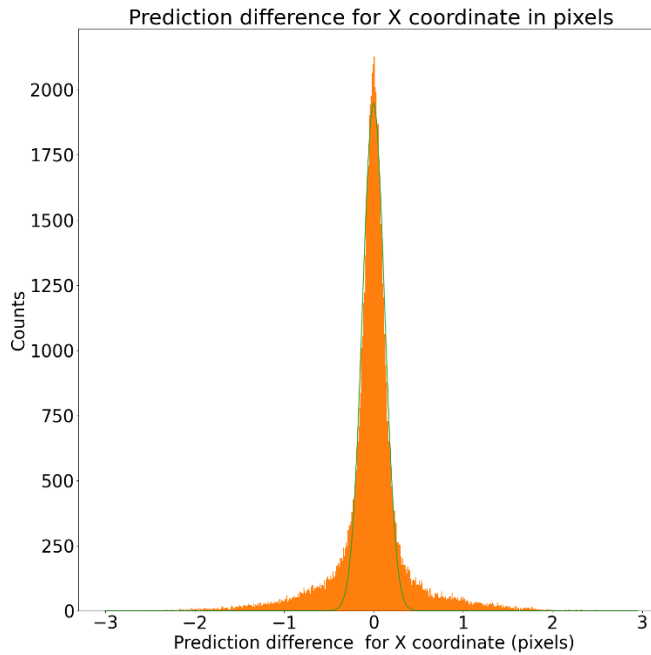
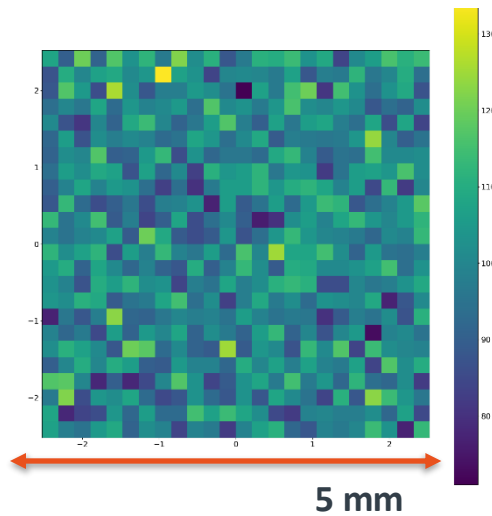


➤ Reconstructing X and Y, convolutional NN

Central data scan:

Active area: -2.5mm to 2.5mm

Counts of samples per position in test data



Difference in reconstruction and empirical coordinates

	Reconstruction error (Pixels)	
	Full Width Half Maximum	Root Mean Squared Error
X/Y error in reconstruction (CNN)	0.26	0.11
X/Y error in reconstruction analytical	3.42	1.46

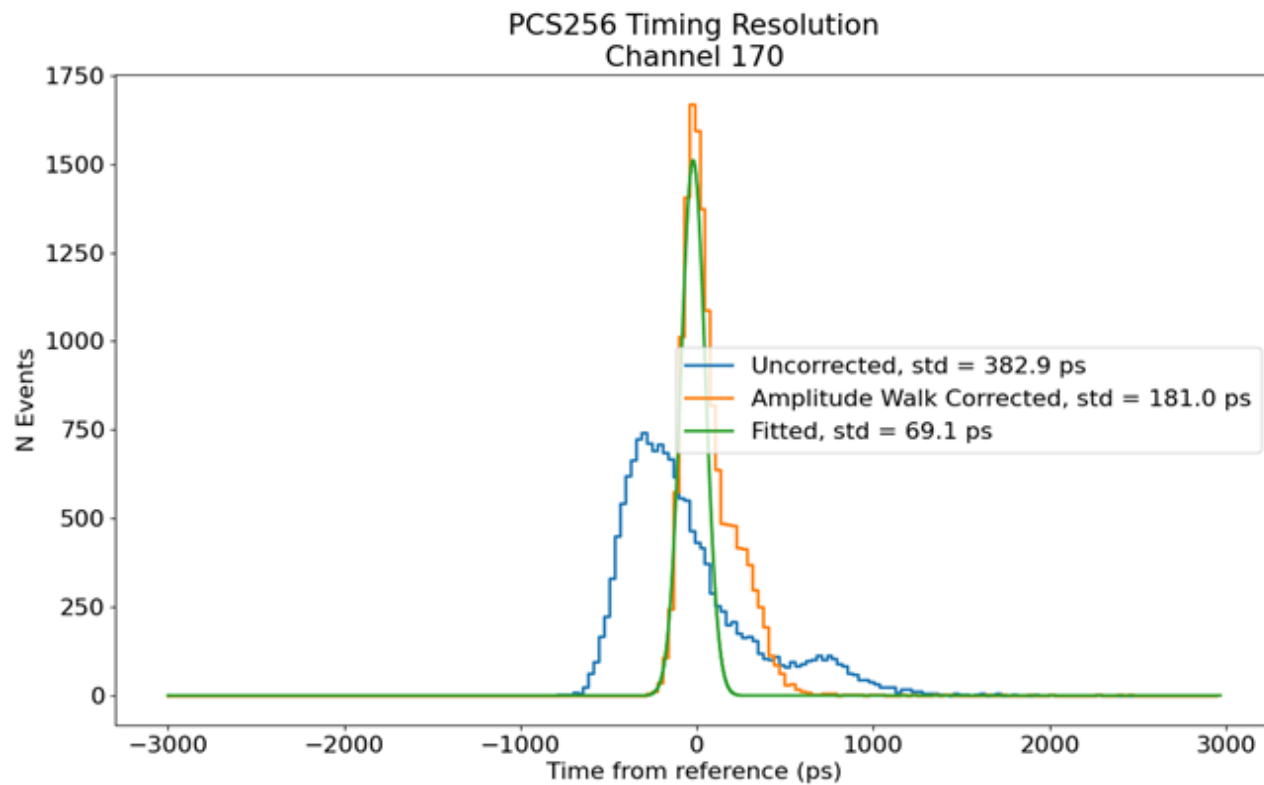
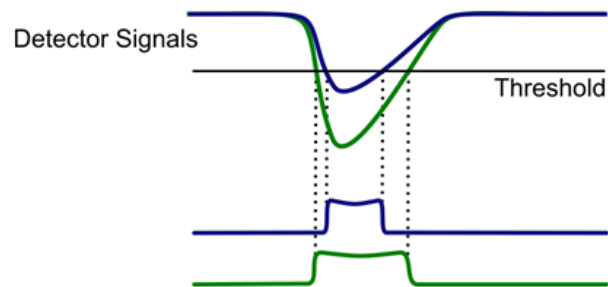
RMS of a top hat distribution:

$$Rms = \frac{1}{\sqrt{12}} Pixel_{width},$$

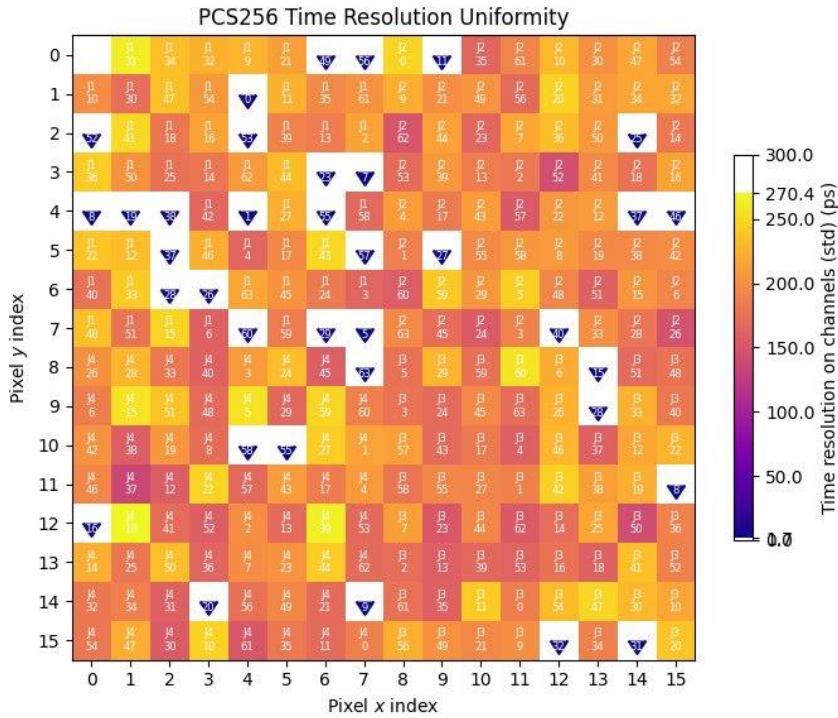
$$Rms = \frac{1}{\sqrt{12}} \times 1.656 = 0.478$$

Characterisation of Resistive Sea PMT

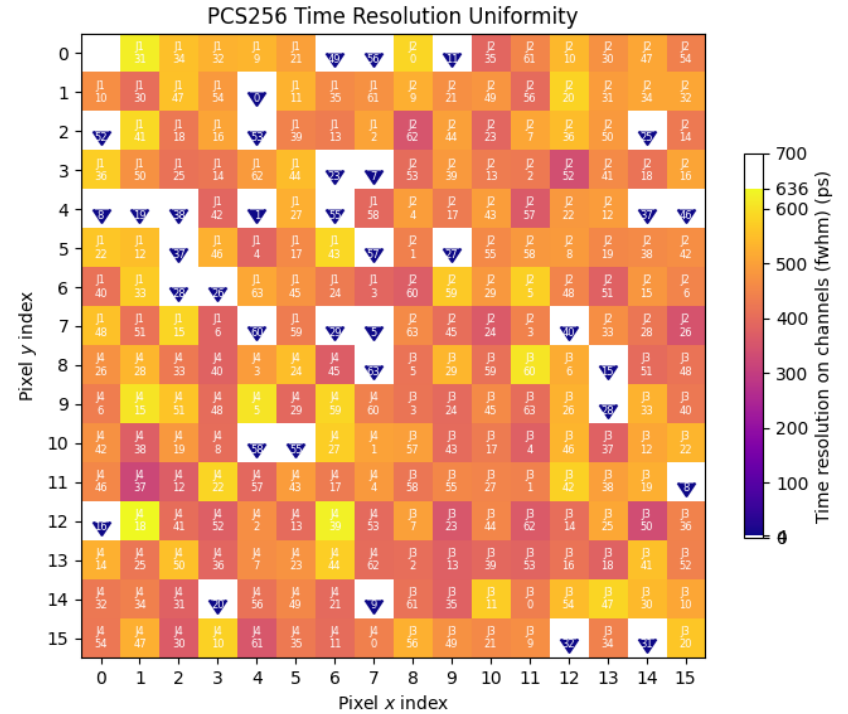
Amplitude walk calibration:



Total RMS Uniformity



Data resolution = 170.1 ps (\pm 74.1 ps)



Data resolution = 399.6 ps (\pm 174.1 ps)

To Conclude

- First prototype of TORCH 16x95 MAPMT built, characterisation ongoing
- Promising cross talk and timing performance
- Proof of principle Neural Network developed for charge sharing ToT imaging photon detector showing potential
- Need to demonstrate with real world images/timing data

With Thanks:
University of Warwick & TORCH Collaboration
University of Leicester, Maths & Physics Departments

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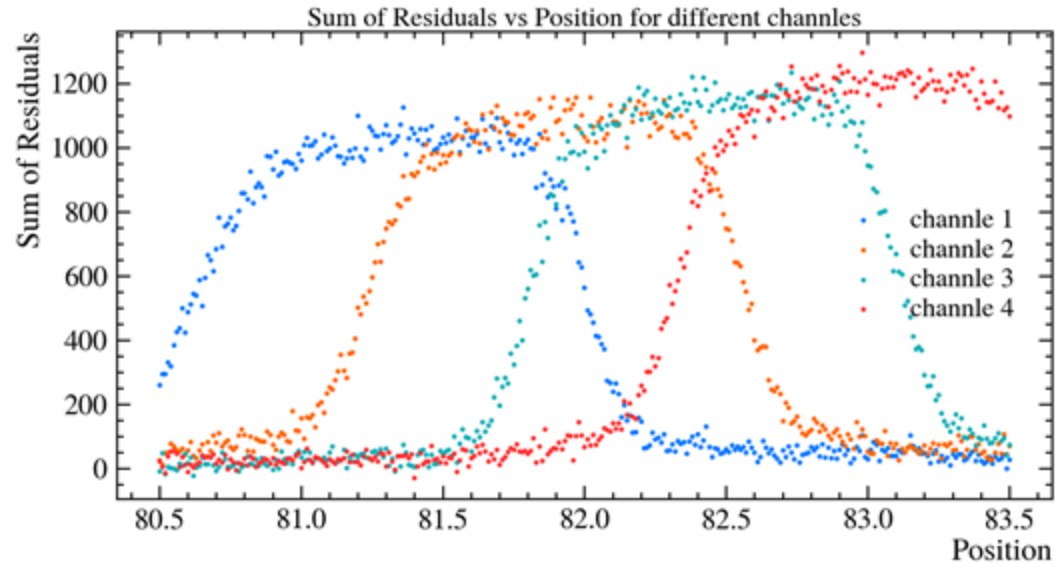
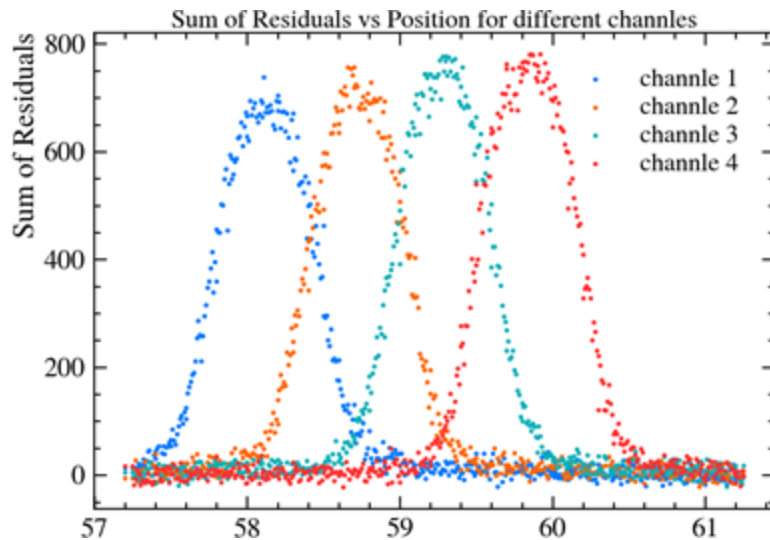
PHOTEK 
ENVISAGE THE FUTURE

Backup Slides



➤ Crosstalk Using residuals

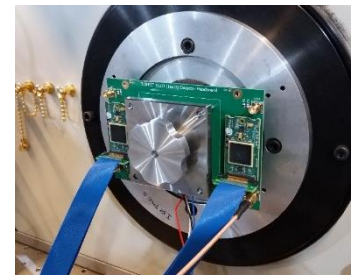
- Measure crosstalk using residuals of PHD fit, i.e. number of events outside noise pedestal (e.g. 5σ and 10σ)



Experimental data testing empirical detector data

Preliminary reconstruction of X, Y and T values of photon position

A pulsed 0.2 mm laser with a wavelength of 638 nm was triggered at a rate of 300 kHz completing a coarse 0.25 mm step 28 mm² circular scan of the detector's active anode area.



Example of electronics set up outside the dark box

A single charge distribution:

What the neural network sees as input:

X (in pixels)	Y (in pixels)	TOT (charge)	Time 2 (in ps)	Time 1 (in ps)
X ₁	Y ₁	TOT ₁	T ₁	T _{laser}
X ₂	Y ₂	TOT ₂	T ₂	T _{laser}
X ₃	Y ₃	TOT ₃	T ₃	T _{laser}
X ₄	Y ₄	TOT ₄	T ₄	T _{laser}
X ₅	Y ₅	TOT ₅	T ₅	T _{laser}
X ₆	Y ₆	TOT ₆	T ₆	T _{laser}

Values of order 10¹³

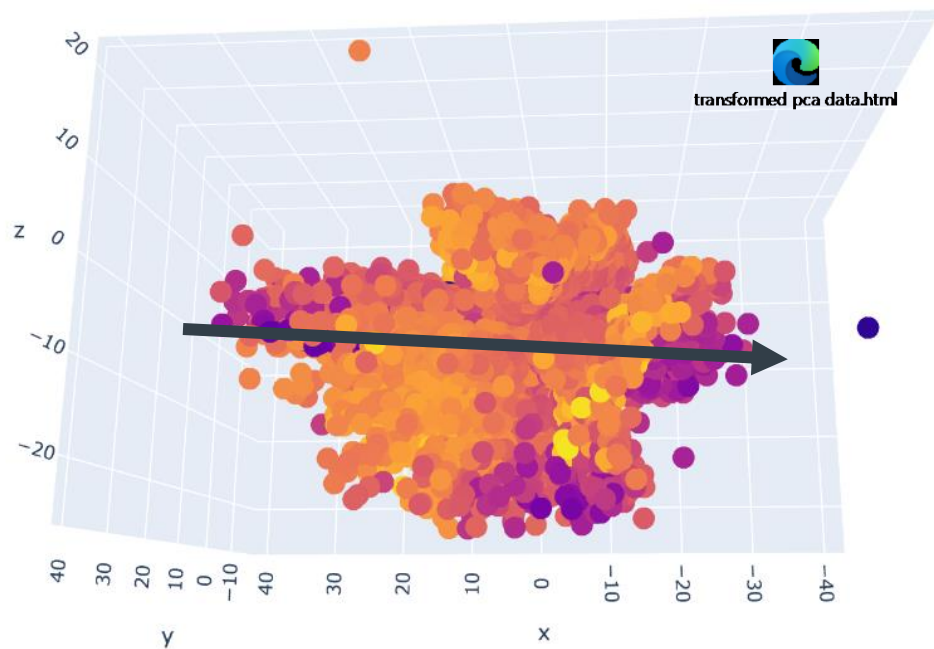
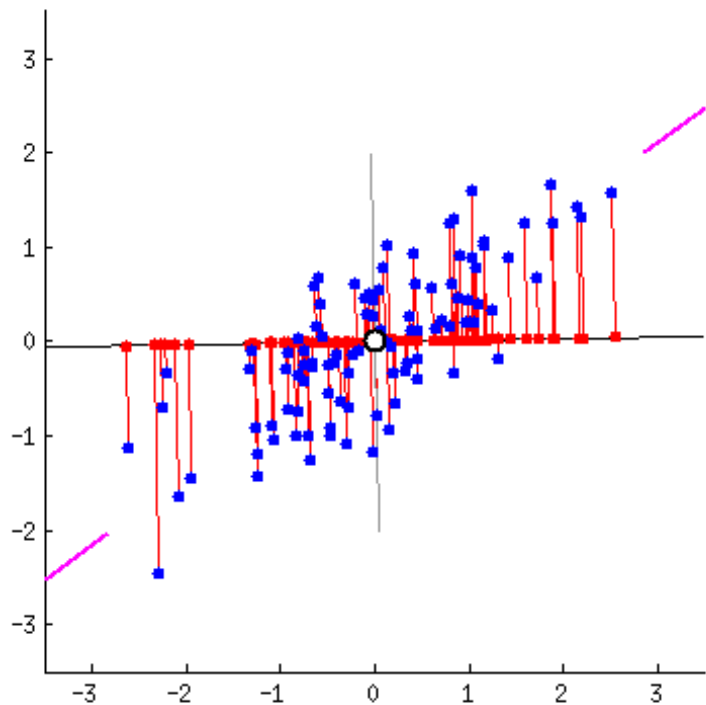
x 10,000 for a scan position

Expected output: X_{photon} (mm), Y_{photon} (mm), T_{laser}

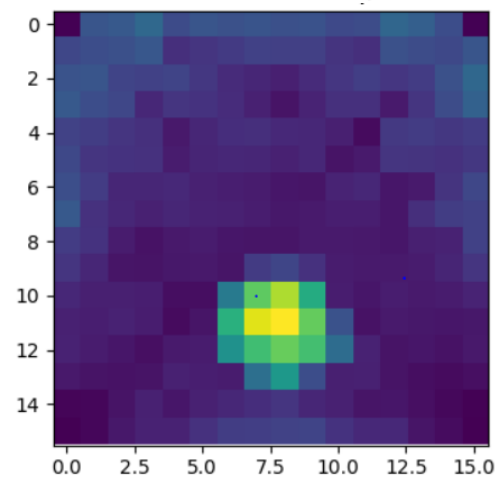
Traversing the latent space

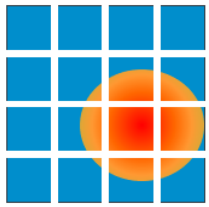
What does the latent space look like in my work?

PCA analysis



Expected result/current work:





➤ Reconstructing the photon position using a Python Monte Carlo trained NN

Convolutional Neural Network to reconstruct photon spatial coordinates X and Y

- The CNN is trained a single sample representing a charge distribution giving information:

- $16 \times (X_p, Y_p, Q)$

Each (X_p, Y_p, Q) represents a packet of data from a single anode pad.

- to give outputs of:

- X_y and Y_y .

Each $(X_y$ and $Y_y)$ represents the spatial photon coordinates for each charge distribution

CENTROIDING ALGORITHMIC METHOD

$$X_Y = \frac{\sum x_i q_i}{\sum q_i},$$

$$Y_Y = \frac{\sum y_i q_i}{\sum q_i},$$

$$T_Y = \bar{t}_i$$

➔ SIMULATION OF REALISTIC INPUTS AND OUTPUTS

Training samples used by the CNN are of the following format:

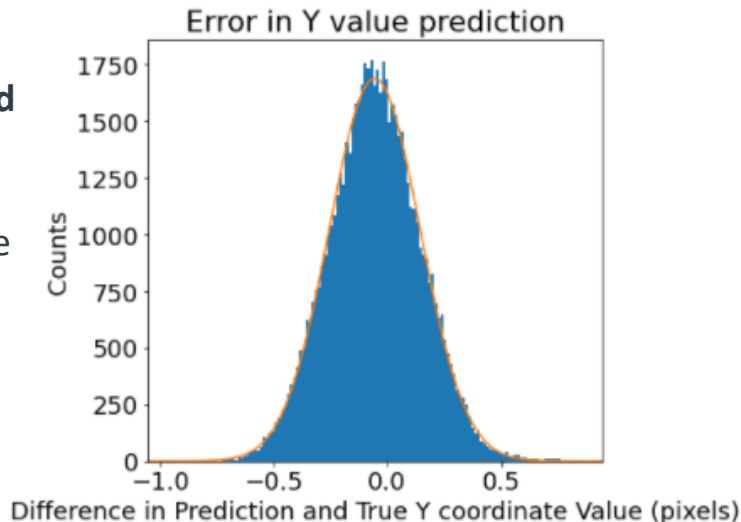
$$16 \times (X_p, Y_p, ToT(Q \pm N(ToT, \sigma ToT)), T_{measured})$$

to give outputs of:

$$X_y, Y_y \text{ and } T_y (T_{measured} - AWC(ToT)).$$

RESULTS

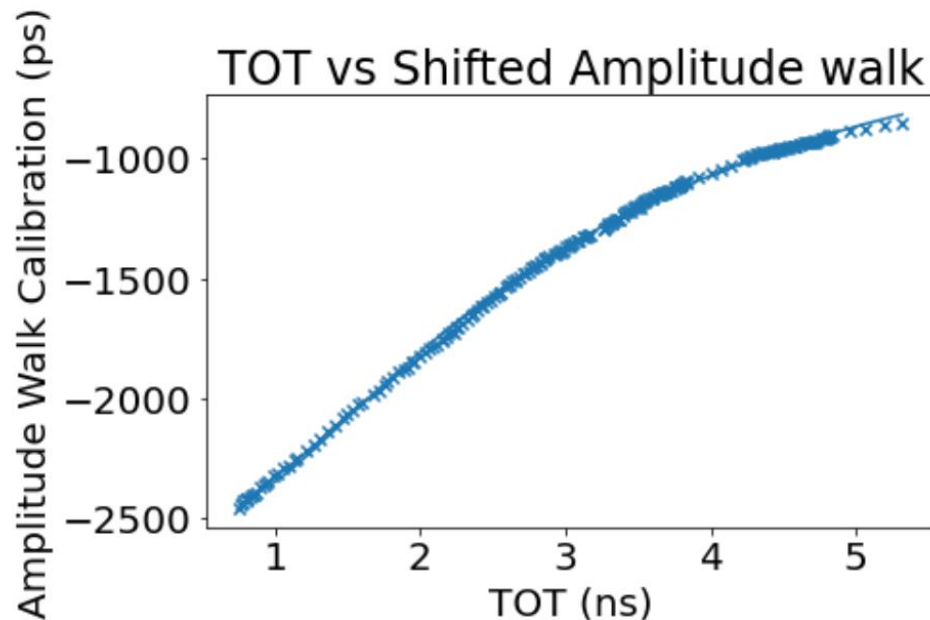
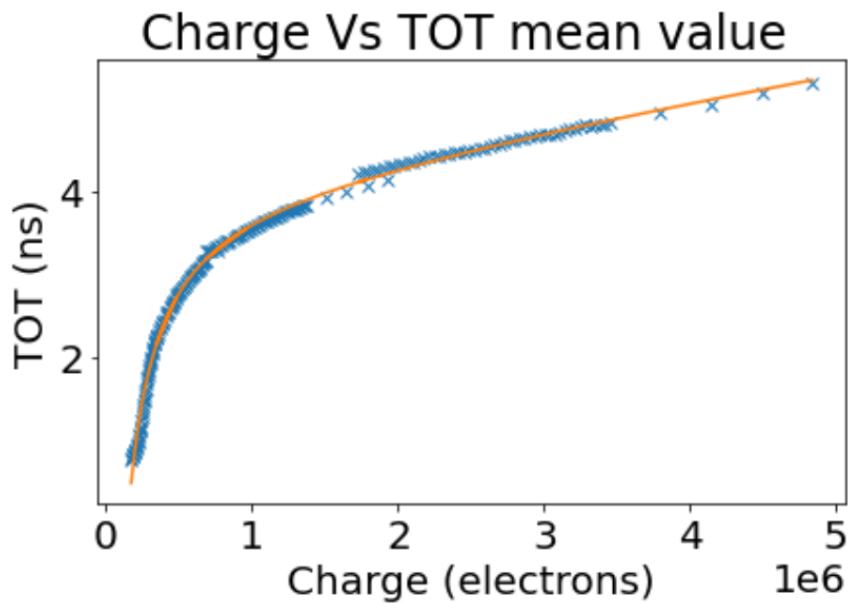
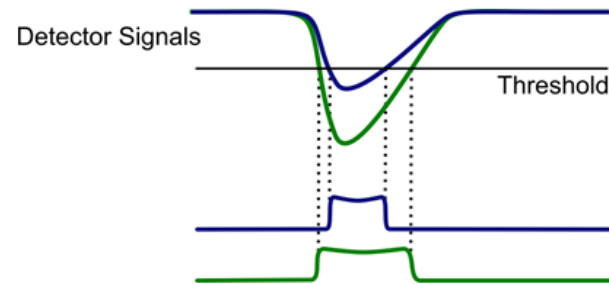
Reconstruct X and Y to an RMS of 0.202 pixels, demonstrating we can increase the spatial resolution by just over 4 times!



Experimental method for calibrations

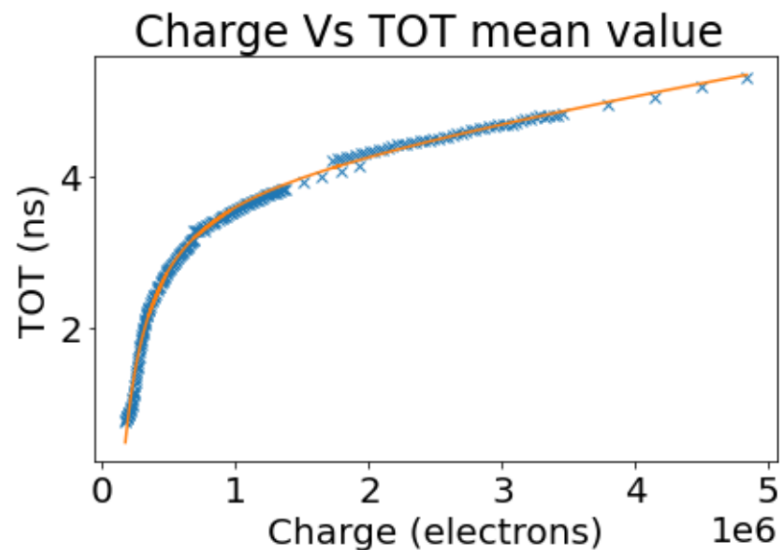
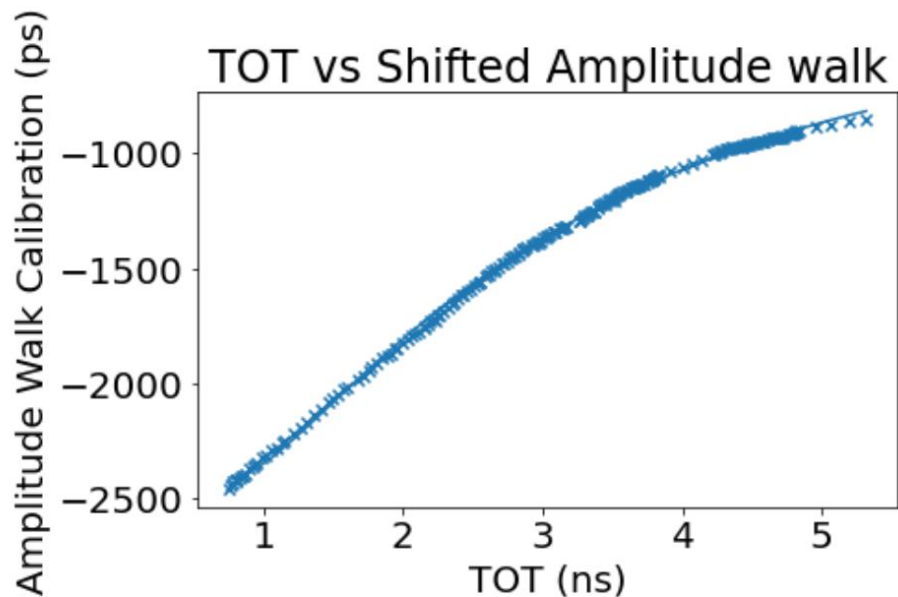
Methodology

- To calibrate for **charge**, **ToT** and T_{γ} :
- TOFPET2d electronics do not give a direct measure of charge, instead it **measures Time-over-threshold (ToT) using a time discriminator technique**, this is related to charge collected on the anode
- a fast voltage step (1 ns rise time) produces a charge injection to each TOFPET2d channel through a 2.2 pF capacitor, emulating the MAPMT (Multi-anode Photo Multiplier Tube) output current pulse.

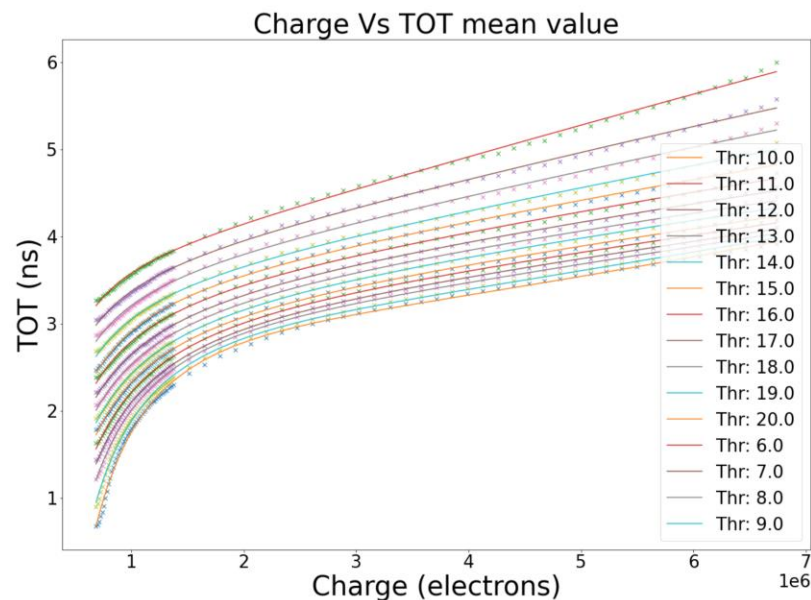


➤ Calibration results

Charge, Time-over-Threshold and time of event calibrations



Future work...



> Training the VAEgan

Training cycle

