

PAUL SCHERRER INSTITUT



Michael Heiss :: **M**uon **I**nduced **X**-ray **E**mission :: Paul Scherrer Institute

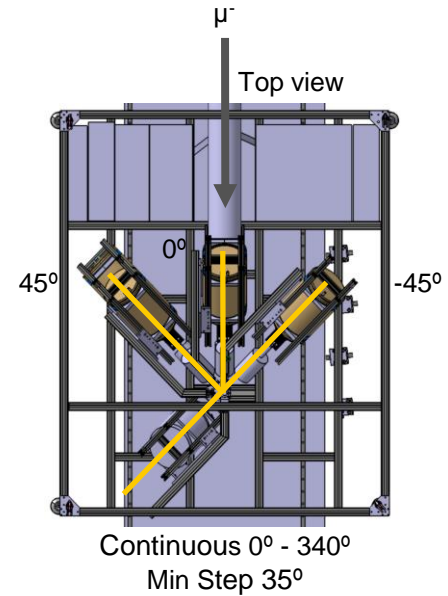
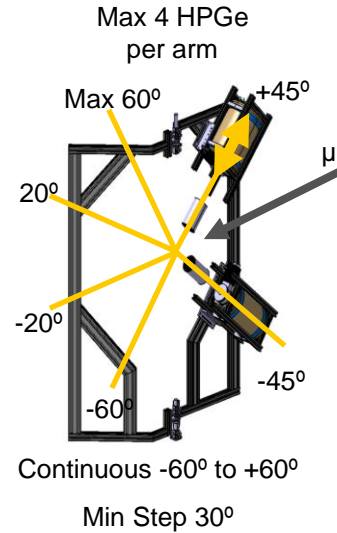
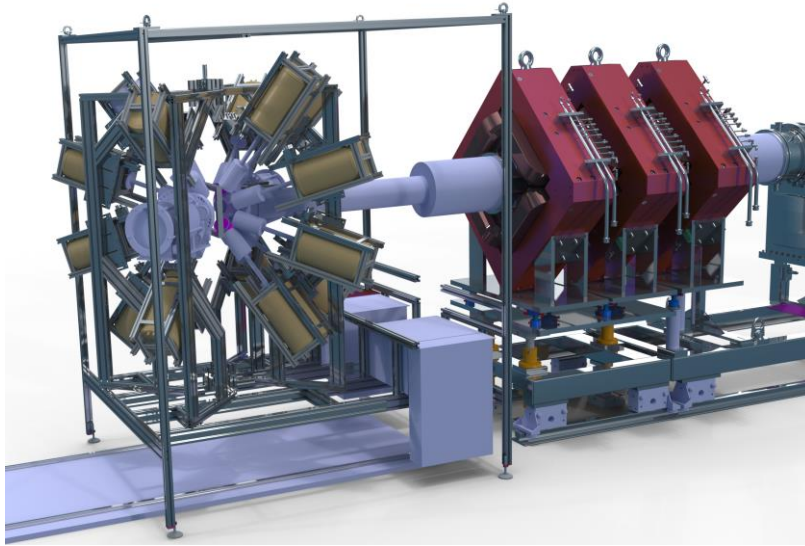
# From Raw Data to Physics Results and where Machine Learning enters the picture

SAMURAI Kick-Off Meeting, 6.9.2023



# Raw Data

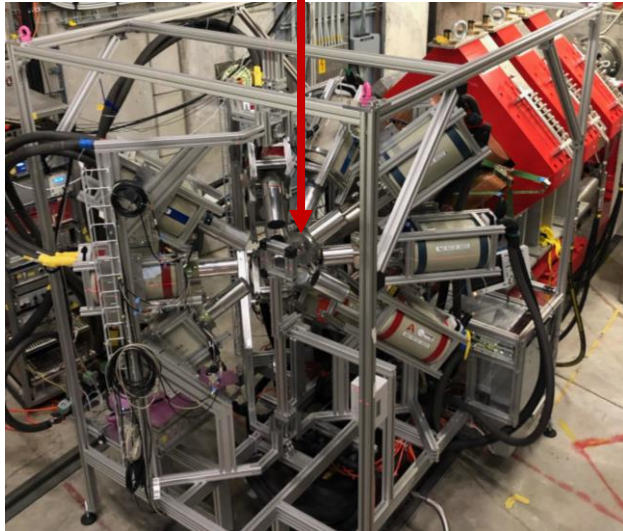
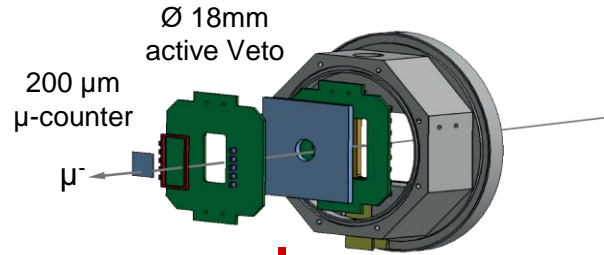
# The GIANT Setup



# GIANT

Germanium Array for Non-destructive Testing

# GIANT: Detector Setup



## • Detectors

- Scintillator with Silicon Photomultiplier readout
- Muon Counter
- Veto Detector

*Start Time Only!*

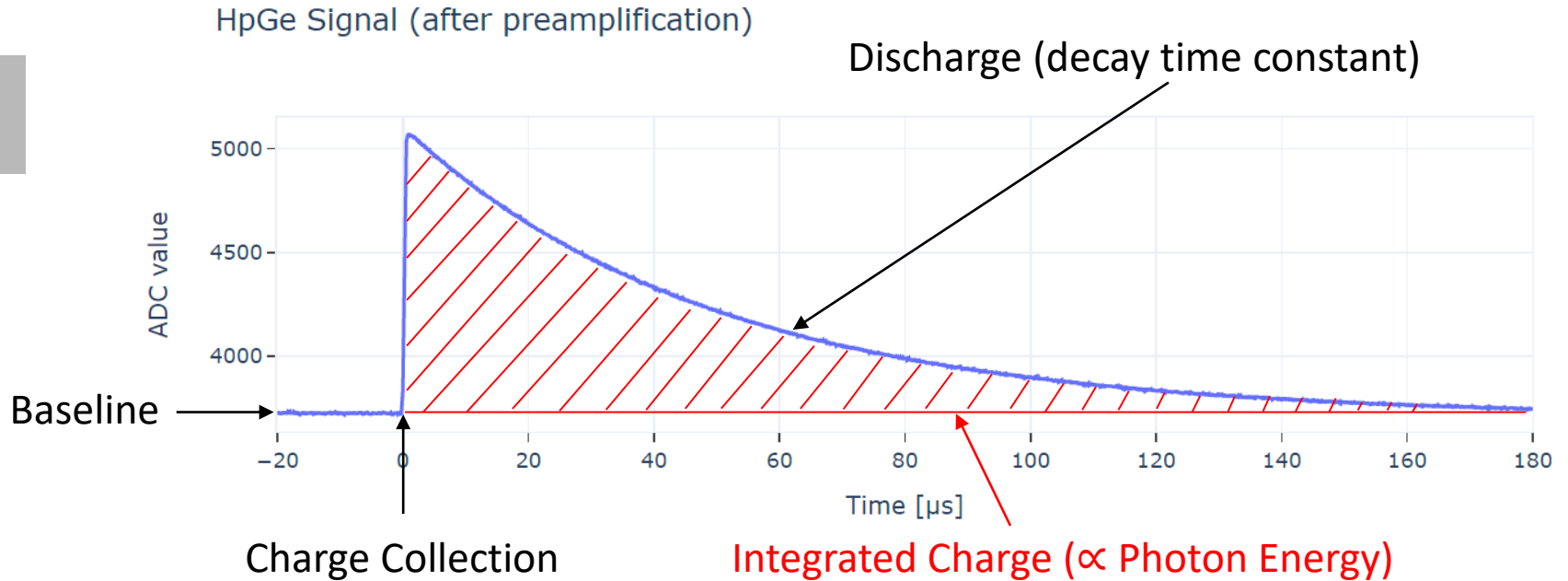
- High Purity Germanium detectors
  - currently 11 detectors (up to 30 total)
  - various types (slightly different response)
  - shared with multiple other experiments

- Silicon Drift Detectors
  - testing phase – currently 2 detectors
  - good candidate for low energy, low noise
  - similar signal structure

*Future Addition to Project?*

- MIDAS DAQ system (PSI development)
  - ingests and stores VME digitizer data
  - efficient binary format (easily convertible to more standard formats)
  - online preanalysis allows status monitoring
  - includes slow control (liquid nitrogen cooling, beam monitoring, etc.)
- Digitizer: Struck SIS3316-250-14 VME
  - 16 channels with 2V or 5V selectable dynamic range
  - 14-bit resolution with 250 MSps sampling rate
  - trapezoidal filter with decay time correction implemented on FPGA
  - fast optical readout
  - (partial) waveform readout for original and filtered waveforms
    - we usually store  $\sim 400$  samples ( $\sim 1.4\mu\text{s}$  per event)
    - allows for offline baseline correction
  - multiple devices chainable (external clock available)

# High Purity Germanium detector signal

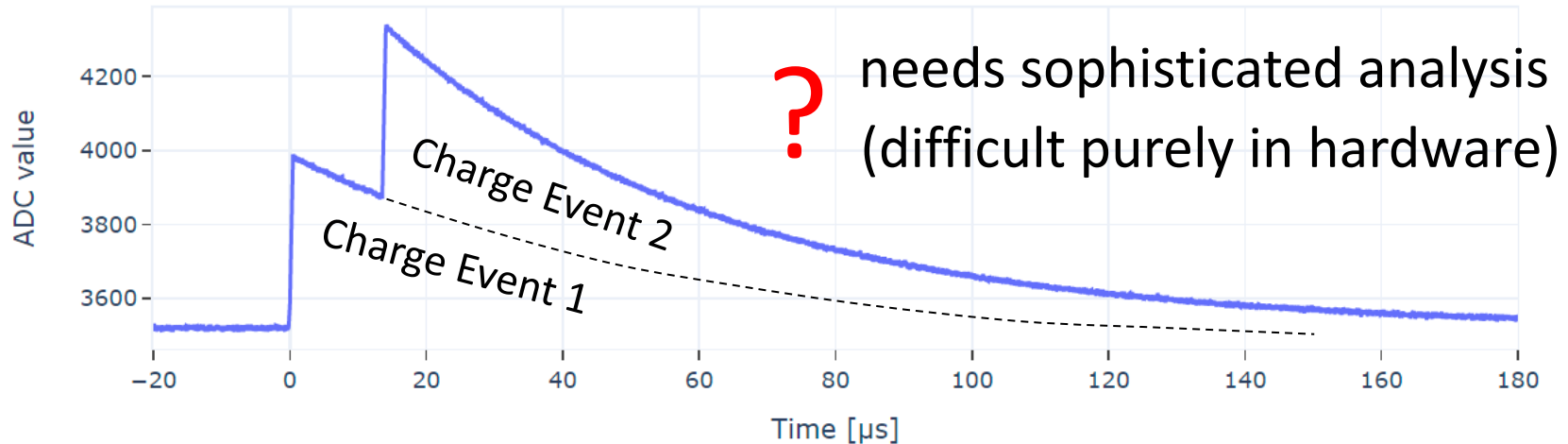


Potentially “best” method

at least in terms of energy resolution (measurement precision)

# High Purity Germanium detector signal

But what about pileup – multiple photons at once?



And what about storage?

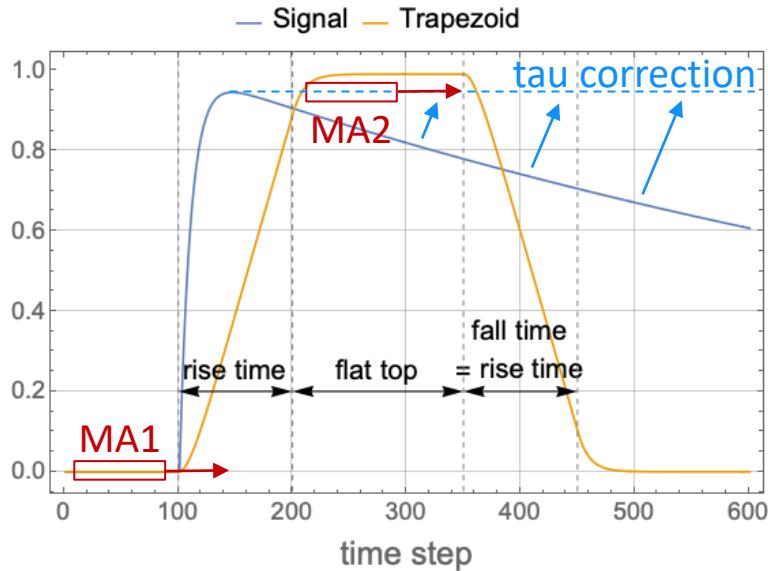
14 bit x 50.000 samples  $\approx$  1 MB per event

**Tens of TB per Measurement!**

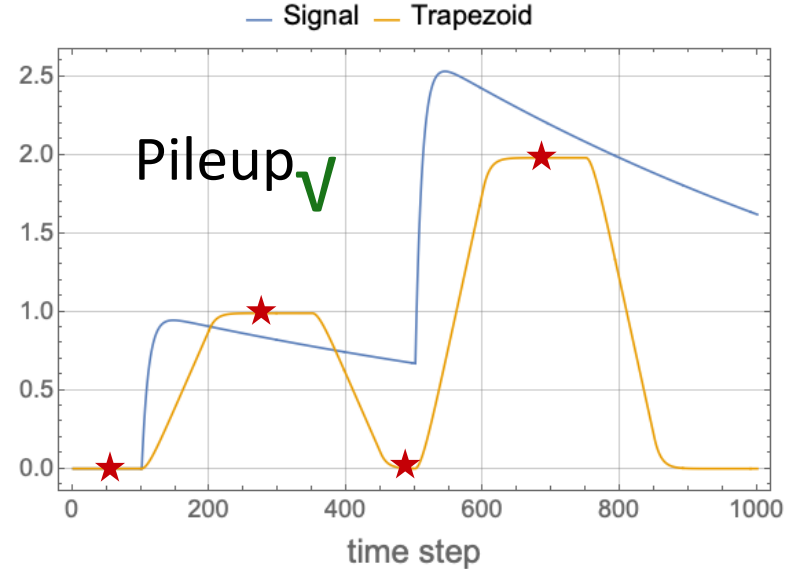


# High Purity Germanium detector signal

## Good alternative – Trapezoidal Filter



Easy to implement in  
hardware (FPGA)

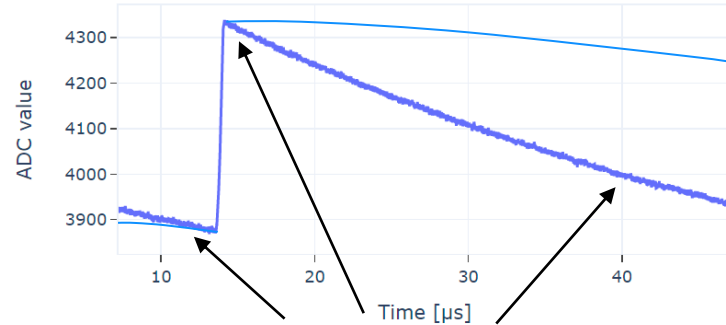
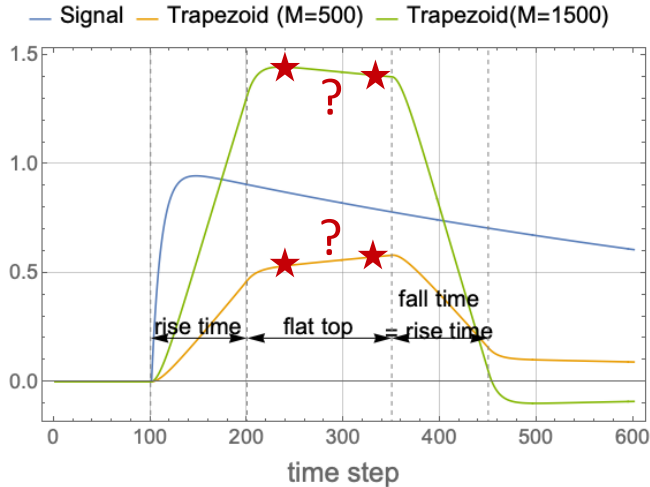


And what about storage?

14 bit x 4 values → few GB per sample

# High Purity Germanium detector signal

## Disadvantages of Trapezoidal Filter



multiple decay constants

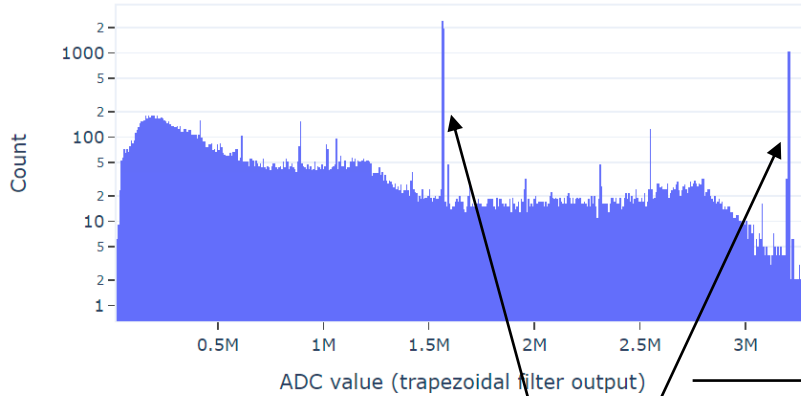
“Optimal” filter parameters strongly depend on detector

Incorrect tau correction can introduce baseline error

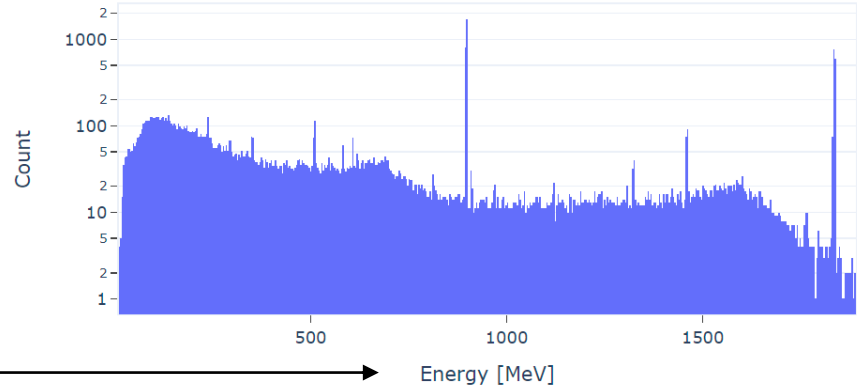
# Physics Results

# Energy Spectrum (Y88 radioactive source)

Histogram of Photon Energies



Energy Spectrum



## Energy Calibration

**Radiation:** Decay mode: Electron Capture

**Major Positrons:**

Max E (MeV)	Avg E (MeV)	# per 100 dis
0.755	0.355	0.2

Max. Beta Range in air 290 cm or 9.51 ft  
 Max. Beta Range in water 0.31 cm

**Major Gammas:**

E (MeV)	# per 100 dis
0.898	93
1.836	99
2.734	0.6

Avg. gamma E = 1.060 MeV

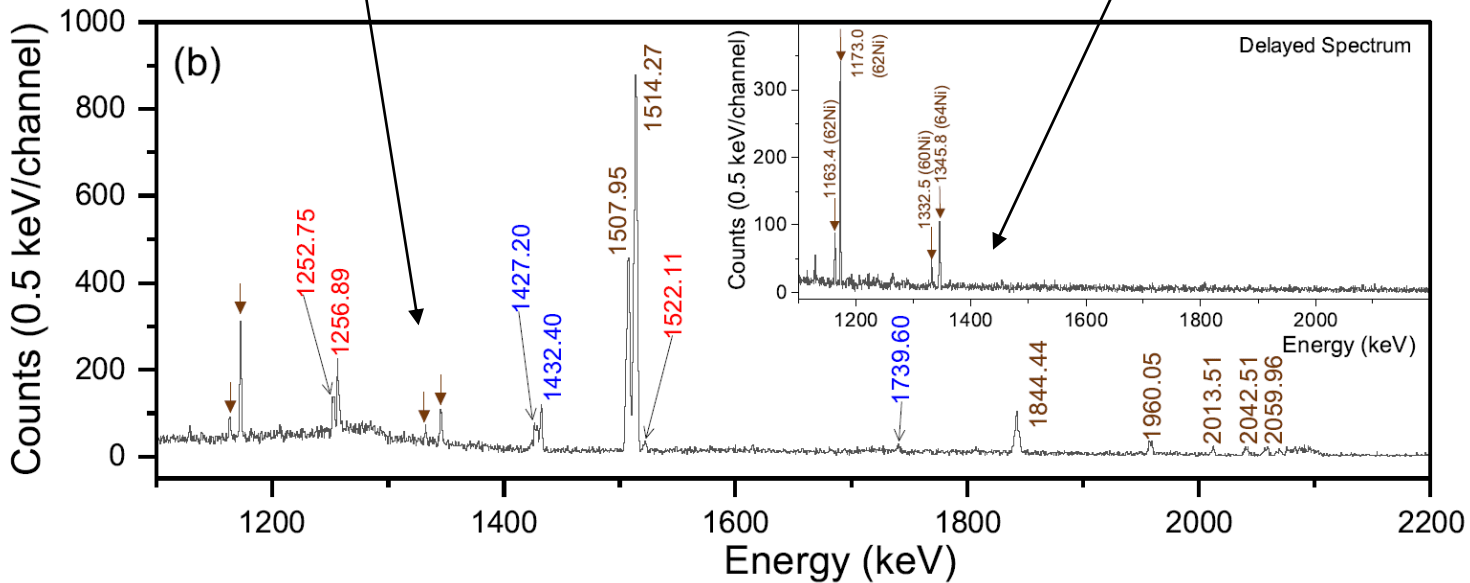
Proper calibration for every detector:  
 many lines over wide range  
 using non-linear corrections



# Muonic Energy Spectrum including timing

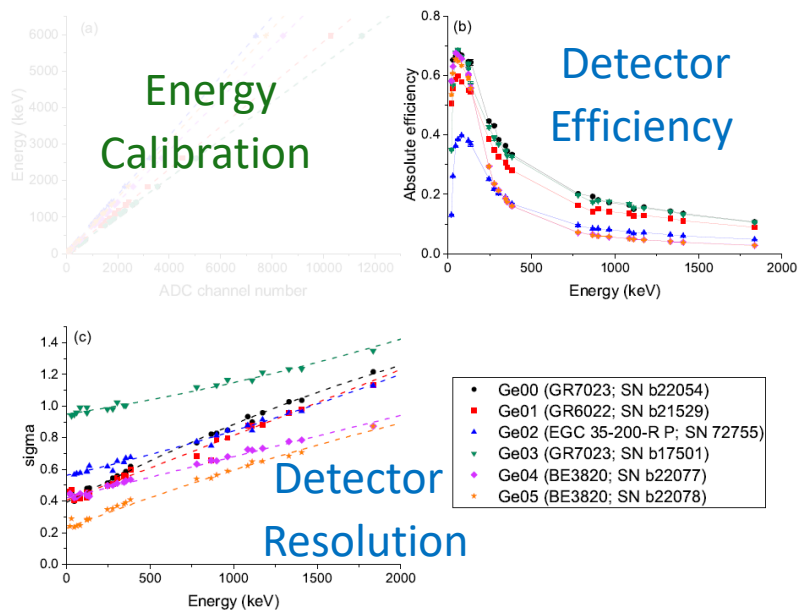
X-ray cascade:  
prompt, <100ns delay

Nuclear Capture and  
subsequent decay:  
delayed, >100ns delay

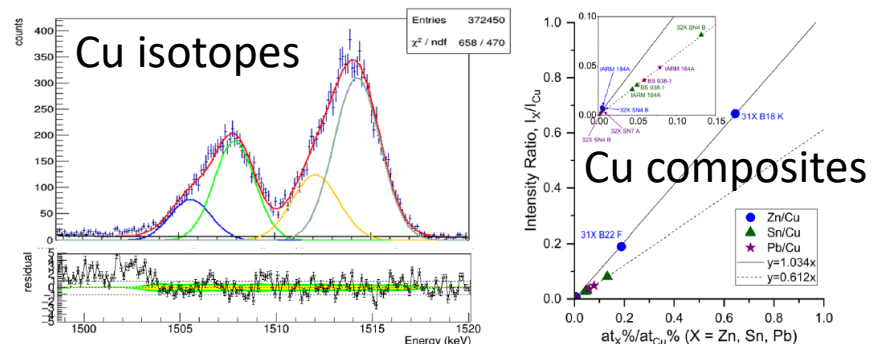


# Quantitative Results

- So far only qualitative results → Which elements / isotopes do we find in a sample?
- Quantitative results are much harder to obtain!
  - *Theory results on relative intensities are currently very unreliable*



Reference Samples allow for relative measurements



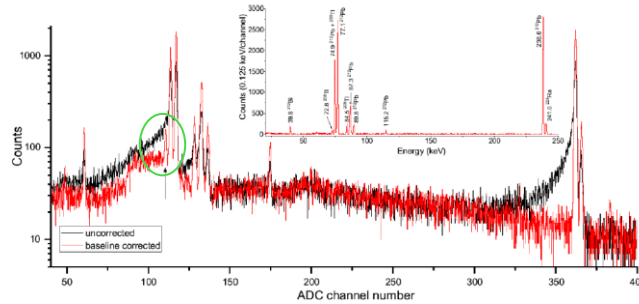
*Attenuation of X-rays and gammas depends on exact material composition, density and penetration depth of the muon!*

# Machine Learning



# Baseline correction

- Unoptimized trapezoidal filter settings and pile-up can lead to reduced energy / timing resolution and “missing” peaks

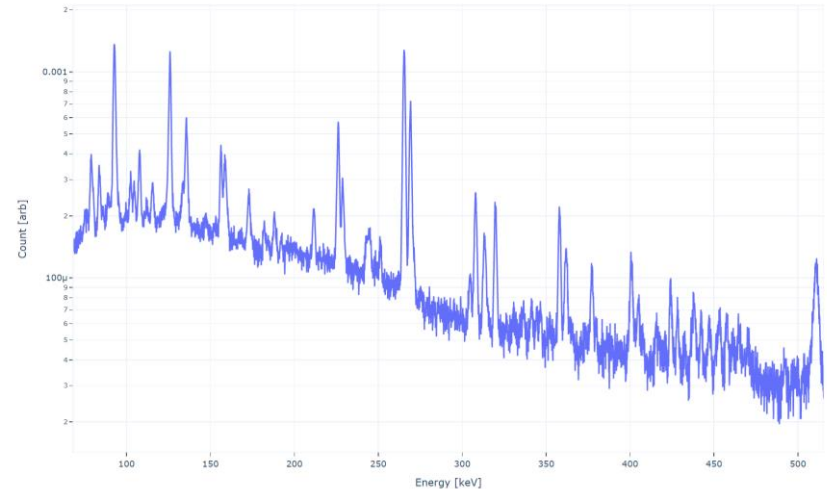


- Machine Learning Model could be used to
  - efficient optimization of trapezoidal filter parameters, e.g. reinforcement learning
  - replace trapezoidal filter by advanced algorithm based on neural network model
  - train using clean set of events with full waveforms
  - implement resulting model directly on FPGA?

*low to medium priority*

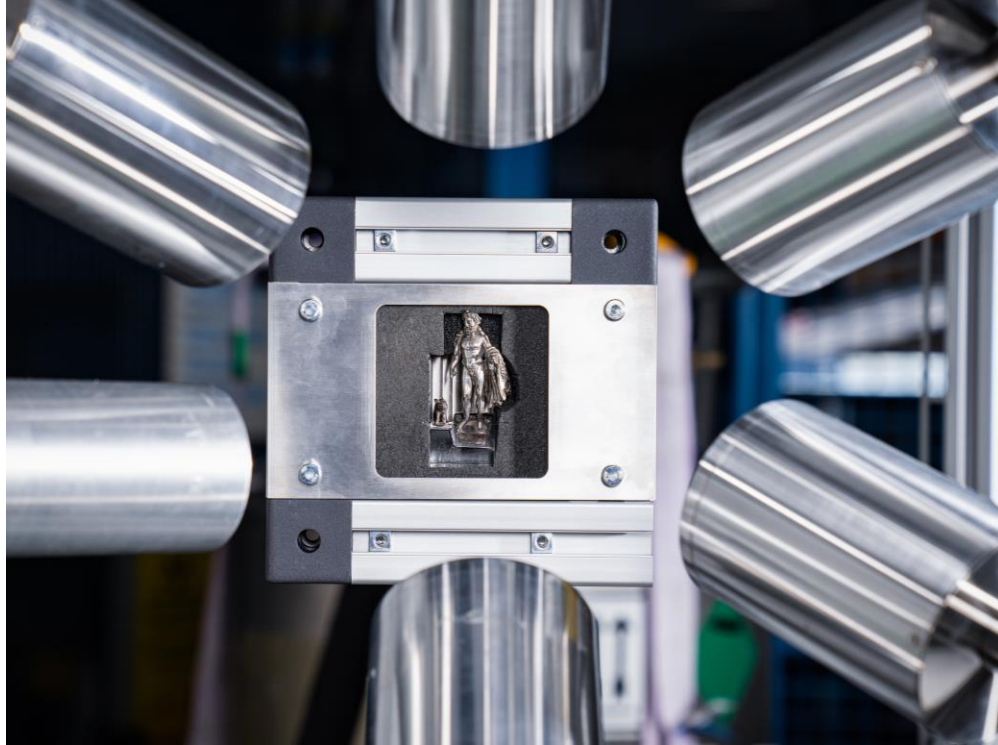
# Peak identification

- **Currently, the peak identification is most serious analysis bottleneck!**
  - has to be done by experienced MIXE scientist (usually not by the user)
  - very time-consuming manual labor
  - overlapping peaks can make ID ambiguous
  - can not be done online → usually no ad-hoc decisions regarding the measurement
- Machine Learning Model could be used to
  - directly identify relevant peak candidates based on a dictionary learning approach
  - give probabilities for candidates based on full spectrum (not only single peaks)
  - empower users to analyze data without domain knowledge
  - allow for “online” (within minutes) analysis of spectra



*very high priority*

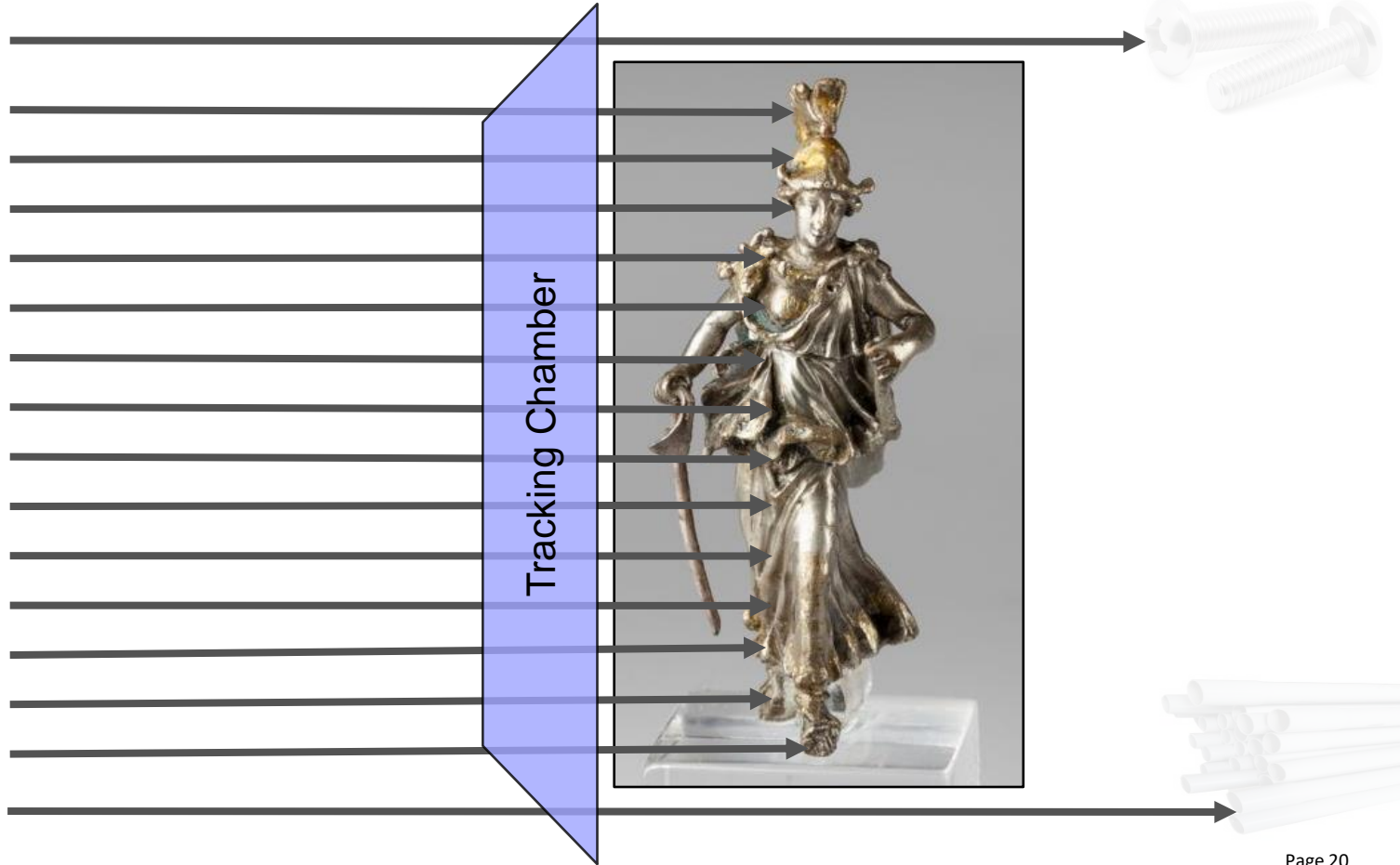
Thank you for your attention!



Any Questions, Comments or Suggestions?

# Bonus: Tracking

μ



# Twin GEM-TPC Tracking chamber

- Twin Time-Projection-Chamber
  - 1D strip readout – 1024 channels in total
  - X (cluster on strips) & Y (drift time)
- Currently track is given by 2 points
  - simple average of spatial charge distribution (X)
  - simple average of temporal charge distribution (Y)
- Machine Learning Model could be used to
  - recover angle information within each TPC
  - determine quality of the track
  - increase detector resolution
  - could be implemented e.g. as (convolutional) neural network, trained with simulation (existing)

*low priority*

