

Data handling, evaluation and unfolding methods for radionuclide spectrometry based on low-temperature calorimetric detectors

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Metallic Magnetic Calorimeters (MMCs) are energy-dispersive low-temperature detectors that are particularly suitable for radionuclide spectrometry over wide energy ranges and with high energy resolution. To obtain a high-resolution beta spectrum with enough statistics to allow a shape analysis, a large number of decays,

in the order of 10⁶, needs to be recorded. A continuous digital data stream for pulse signal detection at high sampling rates may then generate data sets of up to one terabyte in size. This contribution presents a database approach that enables sensible pulse data handling and sorting as well as spectrum evaluation. For higher energy spectra (~ $E_{max} \ge 300$ keV) an unfolding algorithm may be applied. The procedures were implemented in Python and EGSnrc, respectively.



Unfolding algorithm



"divide-and-conquer" approach [5]:

1) Monte Carlo simulation of N single-energy bins (inputs) in absorber geometry over the entire energy interval:



Correction of spectrum distortions

The proposed unfolding algorithm has several advantages [5]:

- the true spectrum shape is not needed as an input.
- it reduces the simulated measured ℓ^1 -error (sum of absolute residuals) \geq 82% at an energy bin width of 1 keV.
- the method is pleasingly parallel and was implemented on a multicore computer cluster.

Simulated examples for ${}^{36}Cl$ ($E_{max} = 709.5$ keV):



2) Use pulse height distributions i.e. N outputs from step 1) to construct a response matrix estimate $S_{N,m}$ columnwise, where m is the number of simulated events/bin. This results in an NxN nonnegative triagonal matrix which is invertible with prob. 1.

3) Calculate the unfolded histogram by multiplying $S_{N,m}^{-1}$ (response) matrix inverse) with the measured histogram.

estimate of the true histogram:

 $\boldsymbol{h}_{N.n}^{true} \cong \boldsymbol{h}_{N.m.n}^{algo} = (\boldsymbol{S}_{N,m})^{-1} \boldsymbol{h}_{N,n}^{meas}$



5 keV bin width, N=142 energy bins, $m=1.10^5$ sim. single-energy bin events/bin

Simulated measured (red) and unfolded histograms (green) in a 4π /on a 2π absorber (top row/bottom row) along with the simulated true histograms. The bremsstrahlung skew is visible in the residual plots (right column) and the correction provided by the unfolding algorithm is substantial. The number of decays in the true ³⁶Cl histogram was $n=2.5\cdot10^6$.

Remark: Energy escape may be reduced by using bilayer absorbers [3,4] that can be further combined with unfolding methods.

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References:

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