M. Borghesi, D. Becker, D. Bennet, M. De Gerone, R. Dressler, M. Faverzani, M. Fedkevych, E. Ferri, J. Fowler, G. Gallucci, . Gard, F. Gatti, A. Giachero, G. Hilton, J. Mates, A. Nucciotti, ... Origo, G. Pessina, C. Reintsema, D. Schmidt, D. Swetz, . Ullom, L. Vale

	The Watson software
	bioct oriented programming. Written in pythen, numpy and scipy for
f	ast (yes, even with python) and robust array and matrix operations.
• E	Easy to read, easy to fix code.
• 0	GUI with QT5 for handy day to day operations.
• 0	Data are saved in hdf5. 🔶 Hierarchial, filesystem-like data format
	Dile-up discrimination
● T ⊦	Two techniques were studied with simulated data: Wiener Filter and DSVP (*)
● A d fi	FIR Wiener Filter of order <i>N</i> is used to leconvolve the response function of the detector rom the signal. WE coefficients
	$x[n] = \sum_{i=0}^{N} a_i w[n-i] \longrightarrow \\ \hline \\ \hline \\ \\ \hline \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\$
● V t D n	Ve developed a novel technique, called DSVP (Discrimination hrough Singular Vectors Projections). It exploits the Singular Value Decomposition, PCA and Multiple Linear Regression to identify as nany undesireble events as possible in a dataset.
	Expected time resolution τ_R $1.5 < \tau_R < 2 \ \mu s$
	for signals with $O(10\mu s)$ of RT \longrightarrow Sub-sample time resolution! and 2 μs of sampling time.
• E	Both the algorithms match the specification of the HOLMES experiment ($ au_R < 3 \ \mu s$
	(*) <u>https://doi.org/10.1140/epjc/s10052-021-09157-x</u>
(Amplitude evaluation
۰T	The energy of the pulse is estimated through optimal filtering for amplitude evalu Optimal filter function Pulse in frequency domain
	$E = C \times \int H(\omega)S(\omega)d\omega$ $H(\omega) = K \frac{S_{avg}(\omega)}{N(\omega)}$ Noise power spec The average pulse is evaluated with an iterative procedure, selecting only a su and correcting the time-jitter between pulses.
• T d	The pulse heights produced by optimal filtering are sometimes subject to an unwe lependence on the exact arrival time of the energy deposition in the detector, rel This bias is reduced by smoothing the signals with a moving average of k poin k is choose as the value which minimize the "entropy" l of the histogram
• V n	Ve are also testing a procedure to evaluate the energies of two overlapping signan nodified version of the optimum filter. Dead time reduction!
C	Event tagging and correct
• T 	The pulses are classified before their parameters are extracted. Coincidence signals, cross-talk, noise events, fast signals (*)
• (*	*) Fast signals are pulses which have a large slew rate \rightarrow the 10 ⁴
	Probability of a fast signals \propto Rise Time of the pulse
	Time Jitter Image: Time Jitter Energy of the pulse (!) Image: Time Jitter
•T s e	This signals are identified and corrected by our algorithm. If they were imply discarded, the energy spectrum would be distorted \rightarrow systematic $\stackrel{10^{-1}}{_{-0.0}}$ error.

Signal processing and data analysis for the HOLMES experiments



In order to reach the goal sensitivy, HOLMES will use 1024 Transition Edge Sensors, each implanted with an activity of 300 Hz, that will record about 10e13 events in three years.

Although ~ 110 MBq of purified Ho163 is avaliable, we haven't yet implanted the Ho in the TES abosrbers. We are currently testing the detectors fabrication steps and the analysy routines.

The TES microcalorimeters for HOLMES must fulfill stringent requirements in terms

The analysis of the pulses from microcalorimeters requires great care, because their excellent intrinsic energy resolution can hardly be achieved whitout an accurate analysis. This contribution is an overview of the essential part of

erc

• The detector gain is not stable during the measurament time. This trend is due to the oscillation of the bath temperature/voltage bias.

pretrigger mean (Baseline)...

... however, the slope of this function depends also on the signal energy.

 $m = m(A_0)$

pulse amplitude at time 0

- First, the function $m(A_0)$ is approximated to a spline of degree 1, using the peaks of known energies as reference.
- •Then, the amplitude spectrum is divided into an arbitrary number of intervals. In order to correct the gain drift for all the signals, a different value of *m* will be used for each interval.

Automatic multi-peak calibration.

- The position of each peak is evaluated with a gaussian fit + non-uniform background.
- Second grade or one grade polinomial for calibration function is enough.

Frequentist and bayesian parameter estimation.

Maximum-likelihood or minimum chi-square with MINUIT as minimizer.





• Detector energy resolution is very close to the one expected from the NEP (4.8 eV).

19th International Workshop on Low Temperature **Detectors (LTD)** July 19 - 29, 2021 National Institute of Indards and Technology

Thermal drift corrections

The drift be corrected by removing the dependence of the signal amplitude on the





Parameters estimation

Hamiltonian MCMC with STAN.

Powerful for multi-dimensional parameters space.

