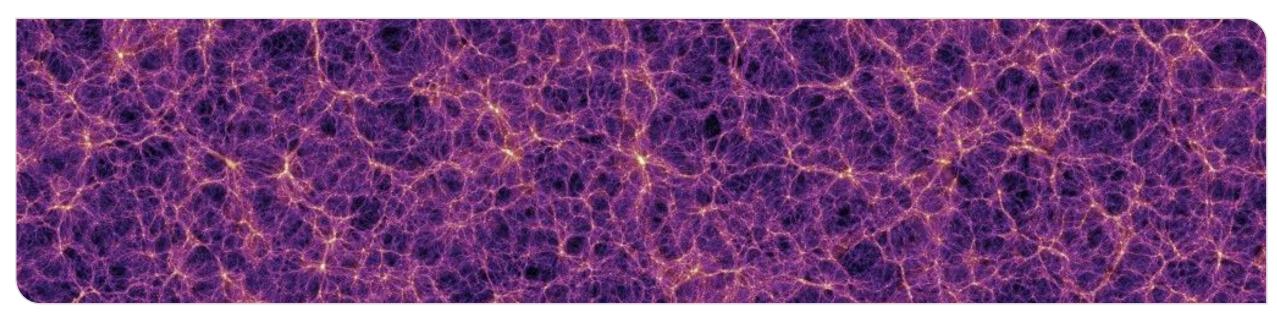


# **Update on MMC analysis**

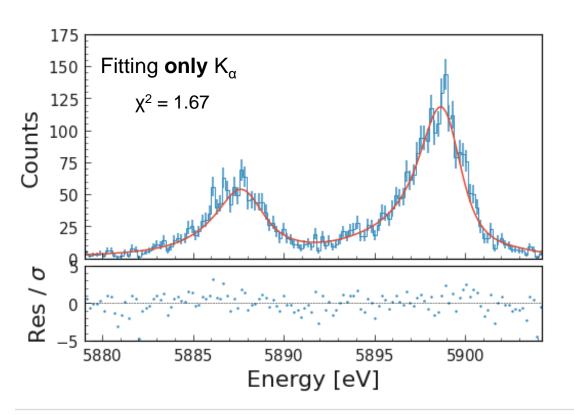
Francesco Toschi
DELight meeting, 08.08.2023

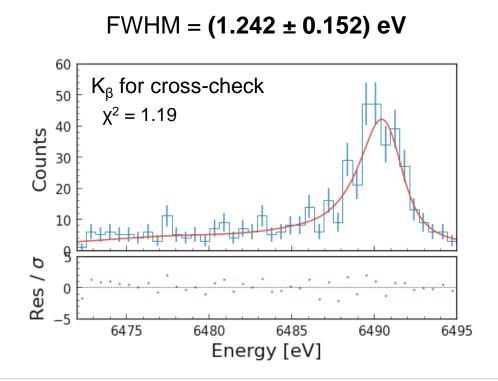


### Where were we?



- Athermal phonons escape modeled as error function
- Minimizing chi-square with binned fit

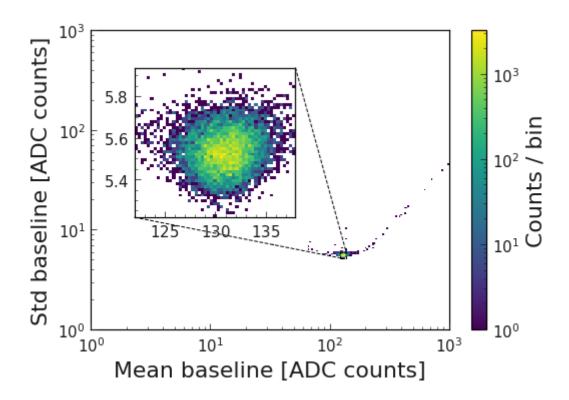




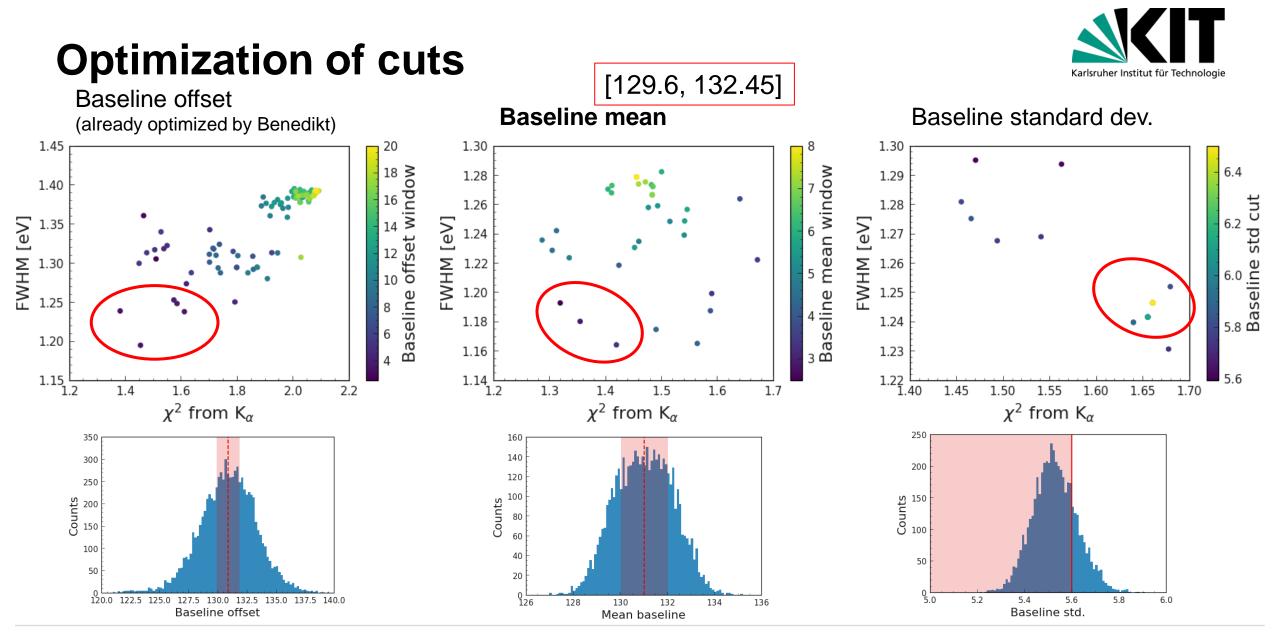
### **Optimization of cuts**



- "Basic" cuts throwing away very bad traces,
- Quality cuts selecting best traces based on baseline.



```
mask mean baseline = df['mean baseline'] < 138 #132.45
mask mean baseline &= df['mean baseline'] > 122 #129.6
mask_baseline_offset = df['baseline_offset'] > 129
mask baseline offset &= df['baseline offset'] < 133
mask std baseline = df['std baseline'] < 6
mask temperature = df['temperature'] > 3e4
mask\_amplitude = (df['OF\_ampl_0'] > 0)&(df['OF\_ampl_1'] > 0)
mask mean = df['mean'] < 20000
mask chi2 TF = df['TF chi2'] < 1000
mask rise time = df['rise time'] > 0.00001
```



## Fit systematics



- Checking the fit for different initial values and binning → systematics
  - Define 2000 initial values (distribution based on experience)
  - Minimize chi-square of K<sub>α</sub> using different binnings
- KAs = np.random.uniform(6, 15, N\_params) SIGMA 0 = np.random.uniform(0.1, 0.7, N params)E2s = np.random.normal(3.26315e-06, 2e-6, N params) E1s = np.random.normal(0.532509, 0.3, N params)

ESC K = np.random.uniform(0, 0.15, N params)

- Estimate chi-square of K<sub>β</sub>
- Repeat for:
  - looser and tighter cuts of mean\_baseline
  - fit with and without escaping factor

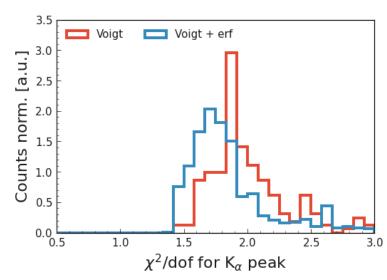
## Fit systematics – loose cut



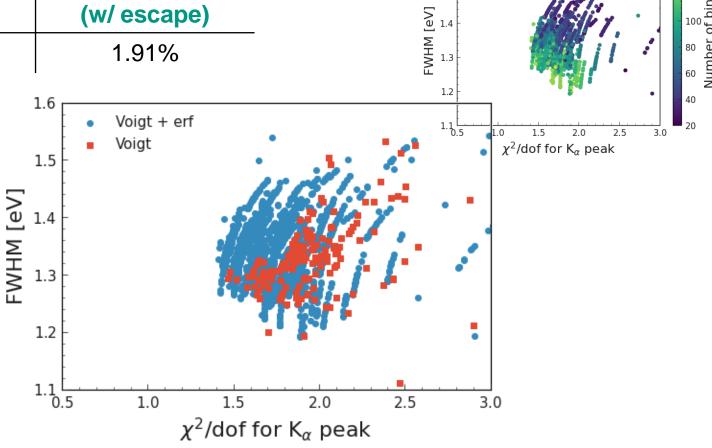
Voigt + erf

1.5

How many fits failed?	Voigt (w/o escape)	Voigt + erf (w/ escape)
Acceptance	0.37%	1.91%



χ² preference for Voigt+erf



## Fit systematics – loose cut



Voigt + erf

How many fits failed?	Voigt (w/o escape)	Voigt + erf (w/ escape)	1.5	140 120 sulg to
Acceptance	0.37%	1.91%		80 Anmper
3.5 Voigt Voigt + erf  2.5 Voigt 1.0 Voigt + erf $x^2/dof $ for $x^2/dof $ for $x^2/dof $	2.5 3.0 eak	1.6 Voigt + erf Voigt  1.5 Voigt  1.7 Voigt  1.8 Voigt  1.9 Voigt	1.0.5	$\sum_{i.0}^{1.5} \sum_{i.0}^{2.0} \sum_{i.0}^{2.5} \sum_{i.0}^{3.0} $

1.0

1.5

 $\chi^2$ /dof for K<sub> $\alpha$ </sub> peak

2.0

2.5

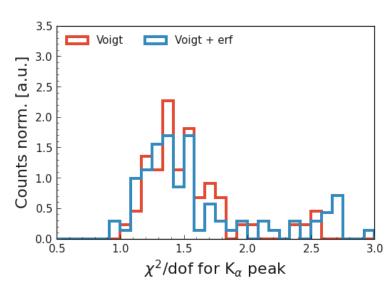
3.0

χ² preference for Voigt+erf

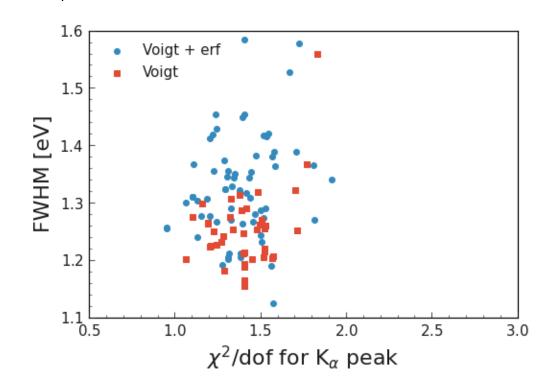
## Fit systematics – tight cut

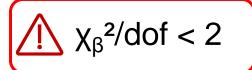


How many fits failed?	Voigt (w/o escape)	Voigt + erf (w/ escape)	
Acceptance	0.31%	0.34%	

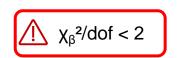


No clear preference: losing statistics, losing information?



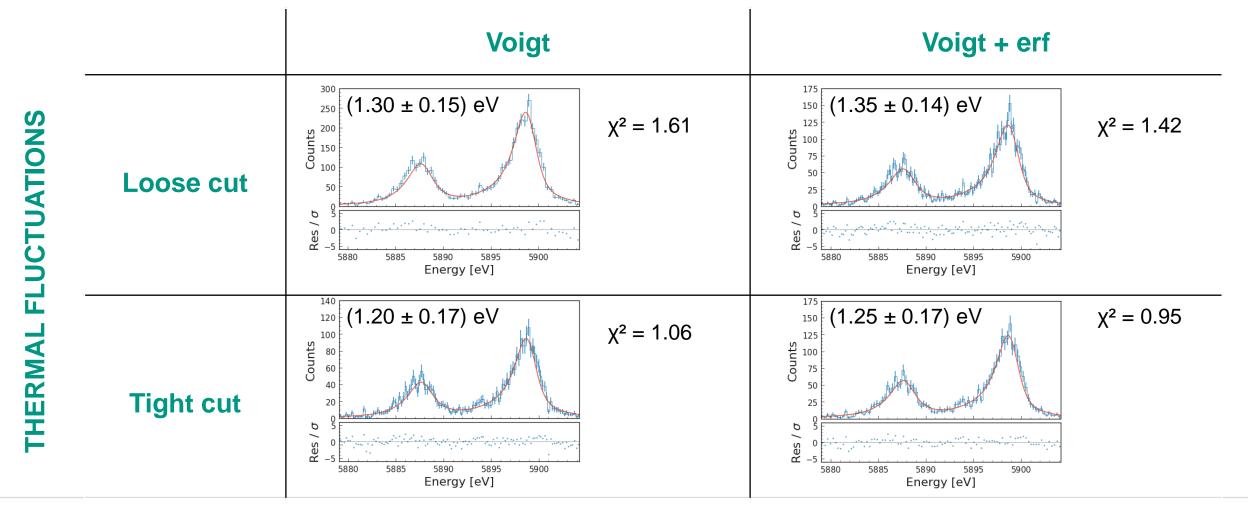


### Picking minimum chi-square

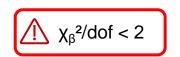




#### ATHERMAL PHONONS ESCAPE

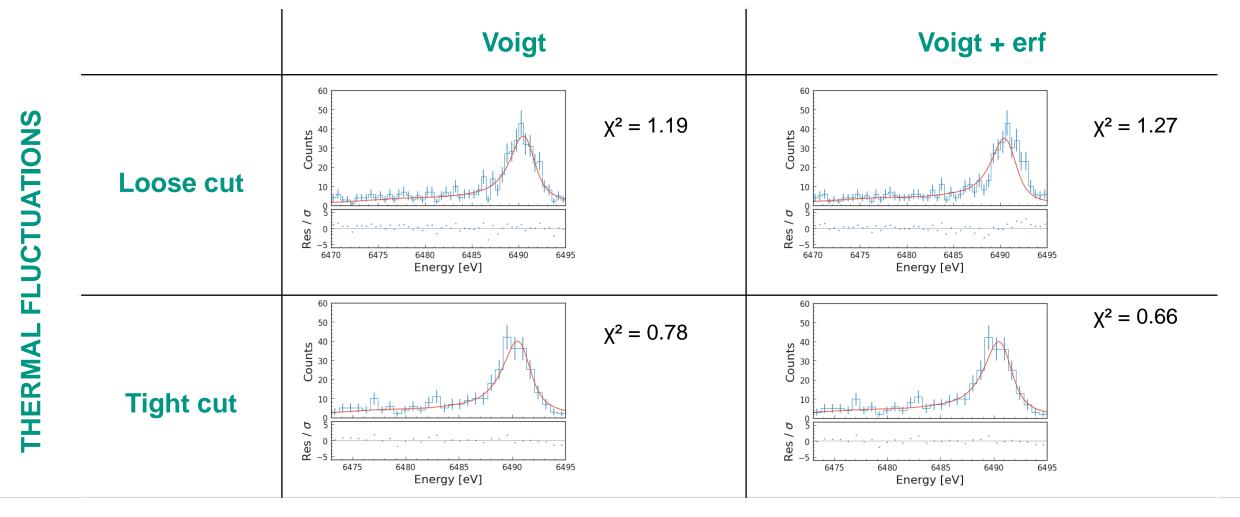


### Picking minimum chi-square





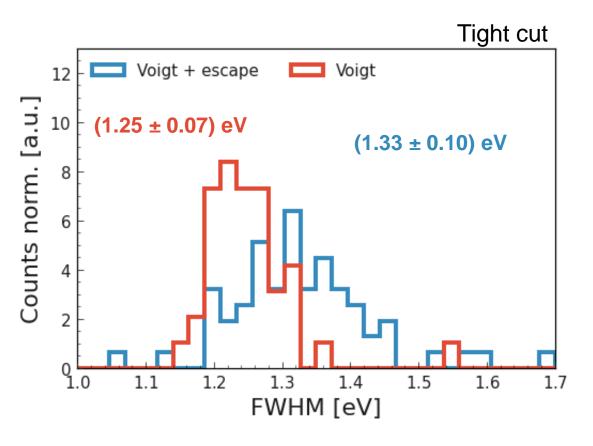
#### ATHERMAL PHONONS ESCAPE

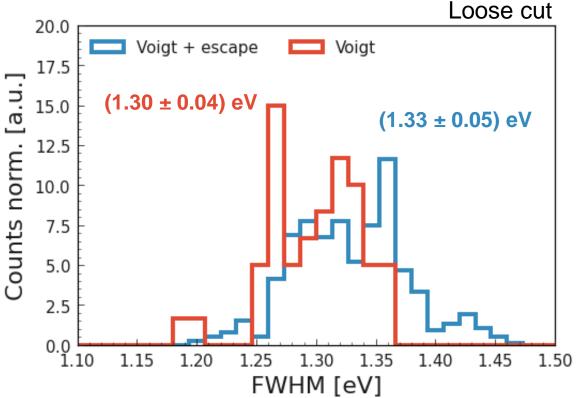


### **FWHM** distribution



 $\chi_{\beta}^2 / \text{dof} < 2 \text{ and } \chi_{\alpha}^2 / \text{dof} < 2 \text{ (similar results with stricter } \chi_{\beta}^2 / \text{dof} < 1.2)$ 





### **Baseline resolution**



After talking with Sebastian, he pointed that it is possible to calculate the optimal energy resolution for a given OF (thermal equilibrium).

#### Time domain

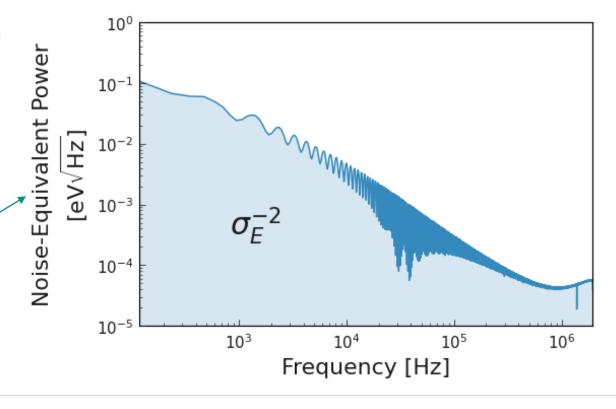
$$S(t) = aA(t) + n(t)$$
  $\longrightarrow$   $\tilde{S}(\omega) = a\tilde{A}(\omega) + \tilde{n}(\omega)$ 

$$\tilde{S}(\omega) = a\tilde{A}(\omega) + \tilde{n}(\omega)$$

$$\chi^2 = \sum_n \frac{\left(\tilde{S}_n - a\tilde{A}_n\right)^2}{J_n}$$
, where  $J_n = \langle \tilde{n}_n^2 \rangle$ 

$$\frac{\partial \chi^2}{\partial a} = 0 \to \hat{a} = \frac{\sum_n \frac{\tilde{A}_n^* \tilde{S}_n}{J_n}}{\sum_n \frac{|\tilde{A}_n|^2}{J_n}}$$

$$\sigma_a^2 = -\frac{1}{2} \left( \frac{\partial^2 \chi^2}{\partial a^2} \right)^{-1} = \left( \sum_n \left( \frac{|\tilde{S}_n|^2}{J_n} \right)^{-1} \right)^{-1}$$



**Optimum Filter** 

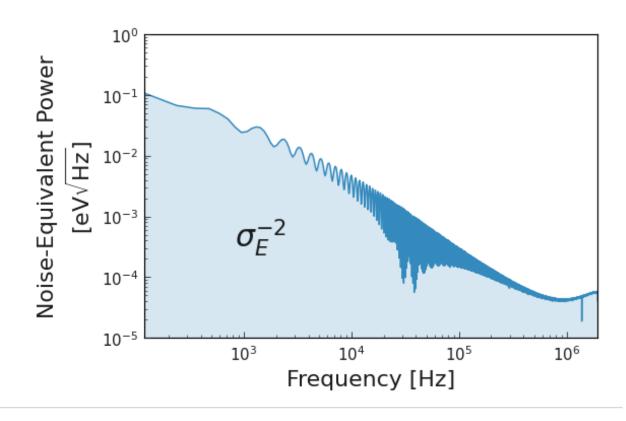
### **Baseline resolution**



After talking with Sebastian, he pointed that it is possible to calculate the optimal energy resolution for a given OF (thermal equilibrium).

### $\Delta E_{FWHM} = 1.09 \text{ eV}$

- Based on Krantz thesis, the theoretical resolution is 0.64 eV (FWHM), but no active temperature stabilization during data taking (T ~ 7 mK), hence possible impact of temperature fluctuations.
- Considering a noise PSD + template using the same cut, we get same resolution → maybe T fluctuations do not impact PSD/templ.?



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### Summary of discussion with Sebastian



- Tighter baseline = less temperature fluctuations
  - This indicates that we are affected by T fluctuations
  - We cannot distinguish between athermal phonons escaping and T fluctuations, indeed chi-square shows no preference between the two models
- For the paper show the entire story (different fits with different cuts), as it shows that we are affected by T fluctuations
  - For DELight we need to have better temperature control/stability

# **Back-up slides**

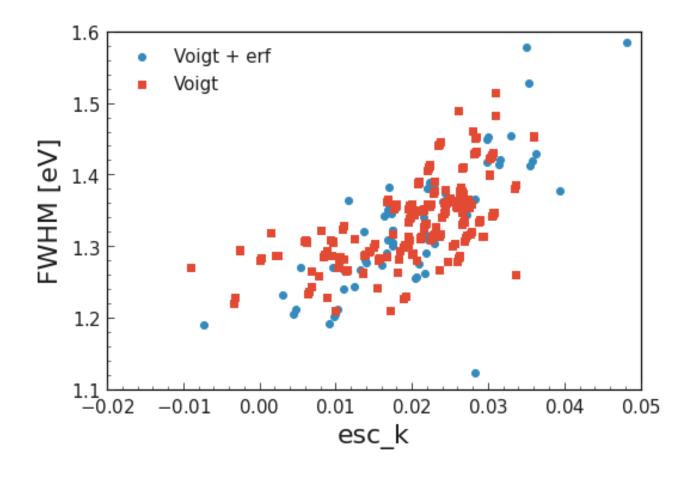




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# Correlation FWHM – esc\_k





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