

# Uncertainties in the SAUNA measurements - calibration revisited

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- Uncertainties in detection efficiencies and interference correction ratios estimated by earlier procedures were too large.
- Calibration procedure
  - $\Rightarrow$  Covariance between the calibration parameters.
- A more robust curve fitting and enhanced uncertainty estimates were implemented.
- Resulting calibration parameters generally have smaller uncertainties.
- Impact of covariance on the uncertainties on Xe-isotopic ratios were studied.

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### P3.2-237

### **Overestimated uncertainties**

Large uncertainties are delivered by the efficiency calibration program Xeff2. Example from a calibration measurement on a SAUNA III:

| #b-gEfficiency |    |          |          |
|----------------|----|----------|----------|
| XE-135         | 2  | 0.560483 | 0.129624 |
| XE-133         | 3  | 0.664781 | 0.164466 |
| XE-133         | 4  | 0.618714 | 0.098366 |
| XE-131m        | 5  | 0.633318 | 0.084492 |
| XE-133m        | 6  | 0.564280 | 0.075282 |
| XE-133         | 7  | 0.220611 | 0.035146 |
| XE-133         | 8  | 0.036837 | 0.005913 |
| XE-133         | 9  | 0.164964 | 0.026294 |
| XE-133         | 10 | 0.439025 | 0.069888 |
|                |    |          |          |

Large uncertainties are delivered by the efficiency calibration program Xeff2. Used scaling of uncertainties overestimates uncertainties in fit parameters. Calibration procedure involves extracting input parameters that are correlated.

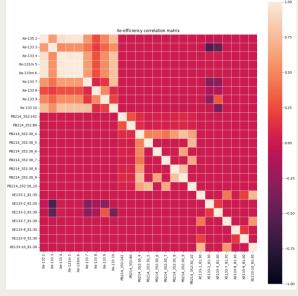
## **Monte-Carlo uncertainty propagation**

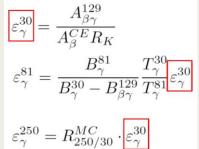
What are the real uncertainties in the efficiency calibration?

Uncertainties in the efficiency calibration are statistical in nature. Have they been propagated correctly? Uncertainties in the efficiency calibration assessed by resampling the calibration measurement data:

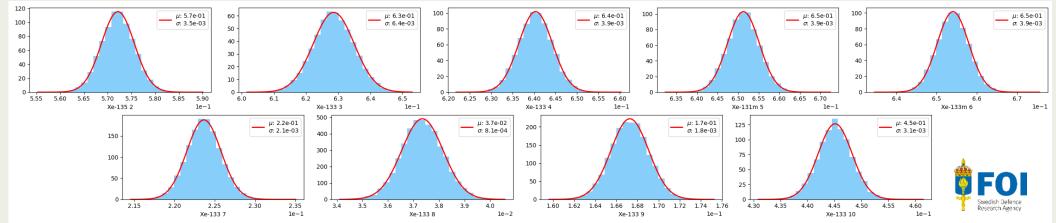
- 1. Each measurement is resampled by assuming  $N_{ij}^{\beta\gamma} \sim Pois\left(N_{ij}^{\beta\gamma}\right)$
- 2. Repeat full calibration procedure for each sampled measurement to obtain samples of efficiency.
- 3. Estimate covariance matrix from the samples.

## **The Correlation Matrix**





The ROI efficiencies are strongly correlated.
Correlation between



input parameters must be accounted for.

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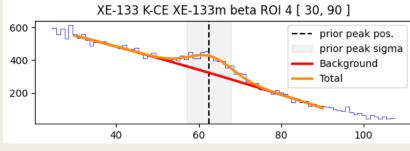
## **Peak Fitting with a Bayesian Prior**

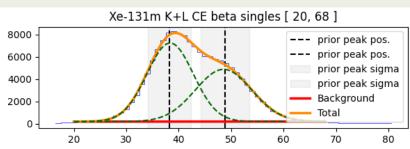
Peak fitting could be difficult in Xeff2. Lot's of human intervention needed for convergence.

- · Small peak on large background.
- Close lying peaks.

Position and width known from energy and shape peak calibration.

- Use Bayesian prior.
- Refactor curve fitting procedures.





## **Uncertainty propagation**

Approximate the efficiency calculation as

$$f(x) \approx f(x_0) + J_f(x - x_0)$$

The efficiency covariance matrix is then

$$\boldsymbol{\Sigma}_f = \boldsymbol{J}_f \boldsymbol{\Sigma}_{x} \boldsymbol{J}_f^T$$



Analytical calculation of the Jacobian is cumbersome and prone to human error. Here implemented instead using scipy.optimize.approx fprime.

The input covariance must still be derived. We are using covariance from:

1. Parameters extracted from one and the same peak fit.

Ex: A,  $\mu$ ,  $\sigma$  from one peak in a spectrum

Counts from (partially) overlapping ROIs. Ex: Counts in ROI 4 and ROI 5 & 6

## Comparison

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|            |   |        |  |

| #b- | -gEfficiency |    |          |          |
|-----|--------------|----|----------|----------|
| XE- | -135         | 2  | 0.560483 | 0.129624 |
| XE- | -133         | 3  | 0.664781 | 0.164466 |
| XE- | -133         | 4  | 0.618714 | 0.098366 |
| XE- | -131m        | 5  | 0.633318 | 0.084492 |
| XE- | -133m        | 6  | 0.564280 | 0.075282 |
| XE- | -133         | 7  | 0.220611 | 0.035146 |
| XE- | -133         | 8  | 0.036837 | 0.005913 |
| XE- | -133         | 9  | 0.164964 | 0.026294 |
| XE- | -133         | 10 | 0.439025 | 0.069888 |
|     |              |    |          |          |

### New:

| #b-gEfficiency |    |          |          |
|----------------|----|----------|----------|
| XE-135         | 2  | 0.572191 | 0.003292 |
| XE-133         | 3  | 0.628564 | 0.006463 |
| XE-133         | 4  | 0.640481 | 0.003798 |
| XE-131m        | 5  | 0.651405 | 0.003687 |
| XE-133m        | 6  | 0.654175 | 0.003736 |
| XE-133         | 7  | 0.223729 | 0.002175 |
| XE-133         | 8  | 0.037357 | 0.000818 |
| XE-133         | 9  | 0.167295 | 0.001837 |
| XE-133         | 10 | 0.445229 | 0.003333 |
|                |    |          |          |

Uncertainties correspond nicely to the Monte Carlo propagation results.

