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# Correlation Effects in Gamma Spectroscopy Efficiency Calibrations and their Impact on Activity and Uncertainty Quantification

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## Introduction

- ◆ Ongoing effort to improve upon & modernize key analysis functions in Canberra's Genie 2000 gamma spectroscopy software
  - ▶ Objective: Consistently provide improvements in algorithms that deliver more accurate, more complete, and more traceable analysis results.
  - ▶ Utilize of new technologies and computing capabilities
  - ▶ Maintain legacy compatibility with existing algorithms
- ◆ Presented today: Correlation effects and impacts on uncertainty
  - ▶ Prototype results; Currently undergoing physics validation testing



## Why Correlation Effects?

- Application: High Precision Gamma Spectrometry Laboratory Measurements
- Enhanced internal uncertainty propagation to address correlations
  - ▶ More accurate uncertainty quantification (UQ)
  - ▶ Defendable, mathematically rigorous treatment
  - ▶ High confidence in reported uncertainty values
- Enhancements focus on two key analysis stages:
  - ▶ Efficiency Calibration
  - ▶ Nuclide Identification with Interference Correction



## Terminology

- **Correlation** means the errors on two random variables are related to one another
- **Covariance** is a quantitative measure of the degree of relatedness

$$\text{cov}(X_1, X_2) = \langle \delta X_1 \cdot \delta X_2 \rangle \equiv \iint \delta X_1 \delta X_2 \text{Pr}(X_1, X_2) dX_1 dX_2$$

$$\delta X \equiv (X - \langle X \rangle)$$

- The **Correlation Coefficient** is a *relative* measure of covariance, between -1 and +1

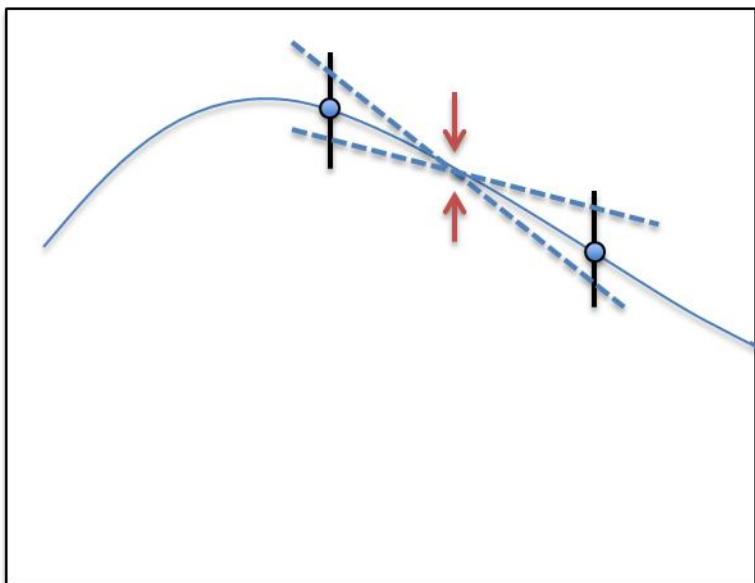
$$\phi_{1,2} = \frac{\text{cov}(X_1, X_2)}{\sigma(X_1)\sigma(X_2)}$$

- **Correlations are important when calculating the uncertainty (variance) of a function of random variables:**

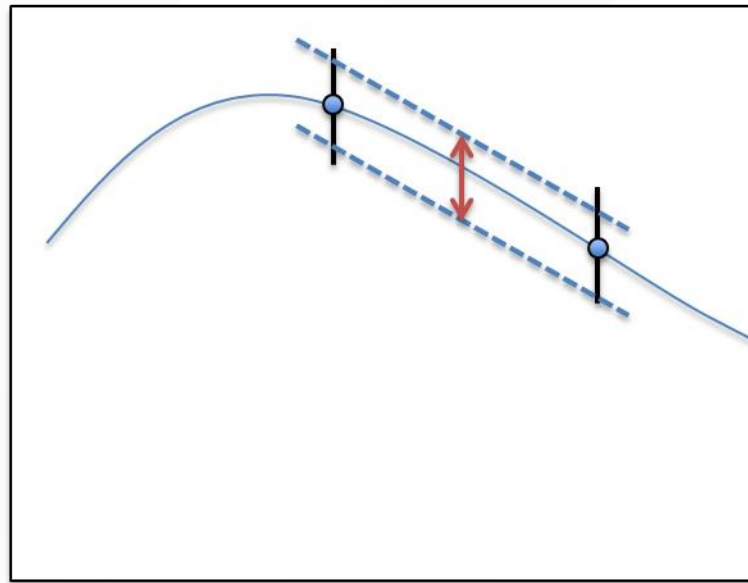
$$\sigma^2(f(X_1, X_2)) = \left(\frac{\partial f}{\partial X_1}\right)^2 \sigma^2(X_1) + \left(\frac{\partial f}{\partial X_2}\right)^2 \sigma^2(X_2) + 2 \frac{\partial f}{\partial X_1} \frac{\partial f}{\partial X_2} \text{cov}(X_1, X_2)$$



## An intuitive way to think about the effects of covariance



Uncorrelated data vary independently, leading to smaller uncertainties in between data points



Correlated data tend to vary together – both up or both down – leading to larger uncertainties in between data points



## Background

- ◆ Measurement uncertainty can be underestimated when covariance effects are not accounted for
  - ▶ Measured efficiency calibrations with multi-line nuclides
  - ▶ Weighted Mean Activities of multi-line sample nuclides
  - ▶ Interference correction
- ◆ These are all instances of Parameter Estimation (Least Squares Fitting)
  - ▶ Combining independent measurements reduces uncertainty
  - ▶ Inappropriate when the data are correlated
- ◆ Current Options for Genie users:
  - ▶ Apply an overestimated measurement Sample Uncertainty
  - ▶ “Empirical”, “Interpolated” efficiency models use interpolated uncertainty propagation
  - ▶ These may result in a greater than necessary activity uncertainty
- ◆ **To directly address correlations, we start to modify the Genie fitting routines used by the algorithms (such as the least squares fitting)**



## Least Squares Fitting with Correlations

- To account for correlated input data, Genie algorithm chi-squared calculation is modified
- $1/\sigma^2$  weighting factors are replaced with the inverse covariance matrix of the data  $V^{-1}$ :

$$\chi^2 = \sum_{i=1}^N \frac{(y_i - \sum a_m f_m(x_i))^2}{\sigma_i^2} \quad \text{becomes} \quad \chi^2 = (\vec{y}(\vec{x}) - \mathbf{F} \cdot \vec{a})^T \cdot \mathbf{V}^{-1} \cdot (\vec{y}(\vec{x}) - \mathbf{F} \cdot \vec{a})$$

where

$$\mathbf{V} = \begin{pmatrix} \sigma_1^2 & \text{cov}(y_1, y_2) & \text{cov}(y_1, y_3) \\ \text{cov}(y_2, y_1) & \sigma_2^2 & \text{cov}(y_2, y_3) \\ \text{cov}(y_3, y_1) & \text{cov}(y_3, y_2) & \sigma_3^2 \end{pmatrix}$$

- **Next Step: How to calculate the covariance of the input data**



## Correlations in the Measured Efficiency Data

- ▶ The Efficiency calibration fits a function to a set of measured “efficiency triplets”  $\{E, \varepsilon, \sigma(\varepsilon)\}$

$$\varepsilon_i = \frac{N_i}{I_{ia}A_\alpha T}, \quad \sigma(\varepsilon_i) = \varepsilon_i \left\{ \frac{\sigma^2(N_i)}{N_i^2} + \frac{\sigma^2(I_{ia})}{I_{ia}^2} + \frac{\sigma^2(A_\alpha)}{A_\alpha^2} \right\}^{1/2}$$

- ▶ Net peak area  $N_i$ , gamma intensity  $I_{ia}$ , Source Activity  $A_\alpha$ , Measurement Live Time  $T$
- ▶ Indices  $i$  and  $\alpha$  specify energy and nuclide respectively

Calibration  
Measurement  
(statistical Unc  
on this peak)

From the  
Nuclide Library

From the  
Source  
certificate

- ▶ Knowing the true correlations is complex and challenging
- ▶ Assumption made: Any gamma lines from a common nuclide will be considered 100% Correlated





## Correlated Efficiency Calibration Engine

- ◆ A modified version of the G2K efficiency calibration algorithm was prototyped in Python
  - ▶ **Calculates correlations between measured efficiency data using Measured Efficiency Triplets and Nuclide Library data**
  - ▶ Modified Least-Squares fitting routine correctly incorporates data correlation matrix
  - ▶ Outputs Efficiency Model Parameters and Parameter Covariance Matrix, which are stored in spectral data file
- ◆ No need for user to calculate or enter correlations
- ◆ Parameter Covariance Matrix data is available to be used in downstream analysis, if needed
  - ▶ Used to calculate uncertainties of calculated efficiency values
  - ▶ Used to calculate correlations between calculated efficiencies
  - ▶ Parameters are always correlated, even if input data isn't!
- ◆ Algorithm tested using simulated data (shown next)



## Efficiency Calibration Examples – Testing Approach

- ◆ Efficiency Calibrations were created using simulated data, both with and without accounting for effects of data correlations.
- ◆ Three calibration standards with differing degrees of correlation,
  - ▶ 3% ( $1\sigma$ ) Activity Uncertainty on each nuclide

Nuclide	Energy
Am-241	59.5
Cd-109	88.0
Co-57	122.1
Ce-139	165.9
Sn-113	391.7
Cs-137	661.7
Mn-54	834.8
Y-88	898.0
	1836.0
Co-60	1173.2
	1332.5

“Mixed Gamma”

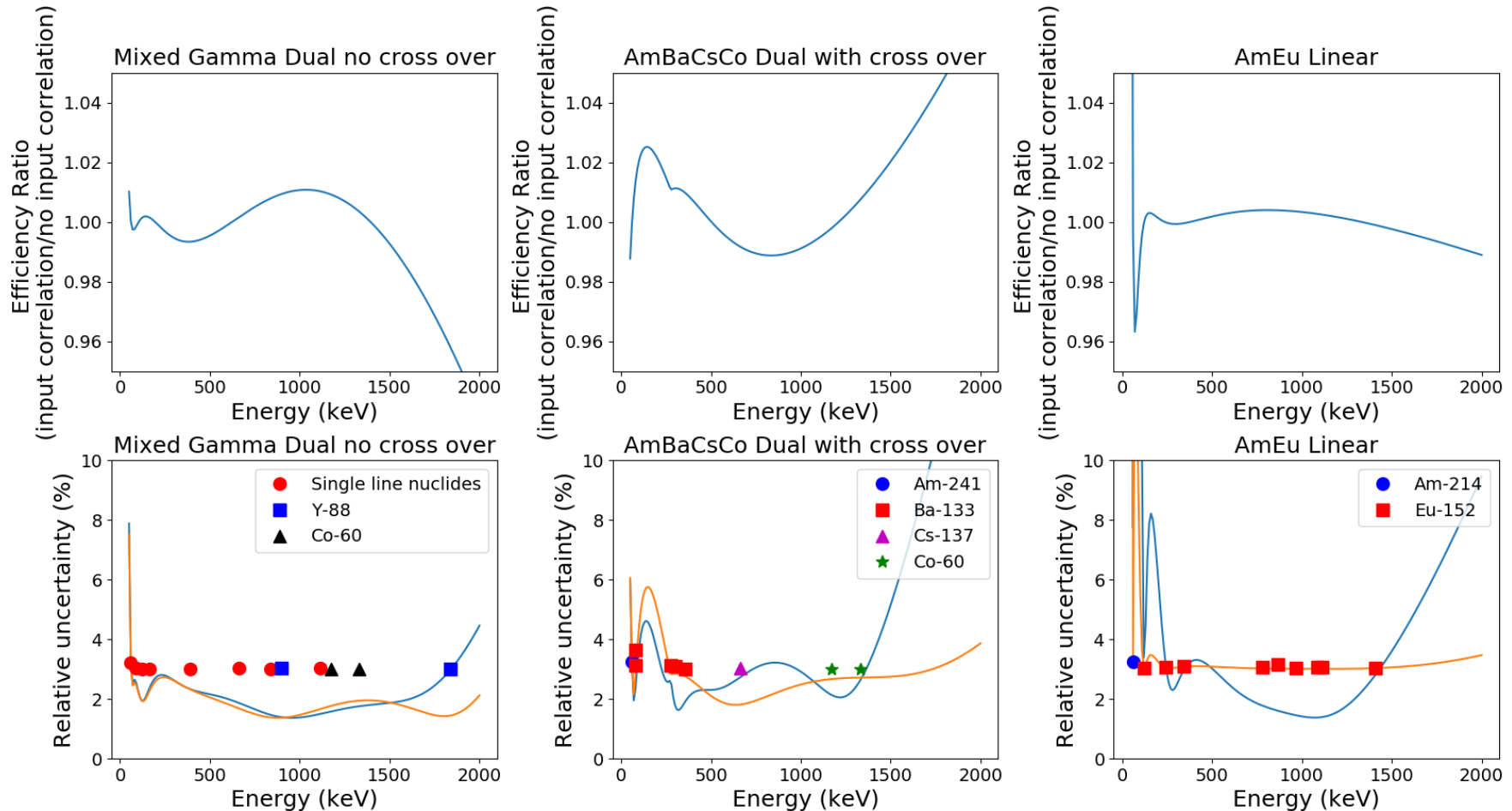
Nuclide	Energy
Am-241	59.6
	79.6
	81.0
Ba-133	276.4
	302.8
	356.0
	383.9
Cs-137	661.7
Co-60	1173.2
	1332.5

“AmBaCsCo”

Nuclide	Energy
Am-241	59.6
	121.8
	244.7
	344.3
	778.9
Eu-152	867.3
	964.0
	1085.8
	1112.0
	1408.0

“AmEu”

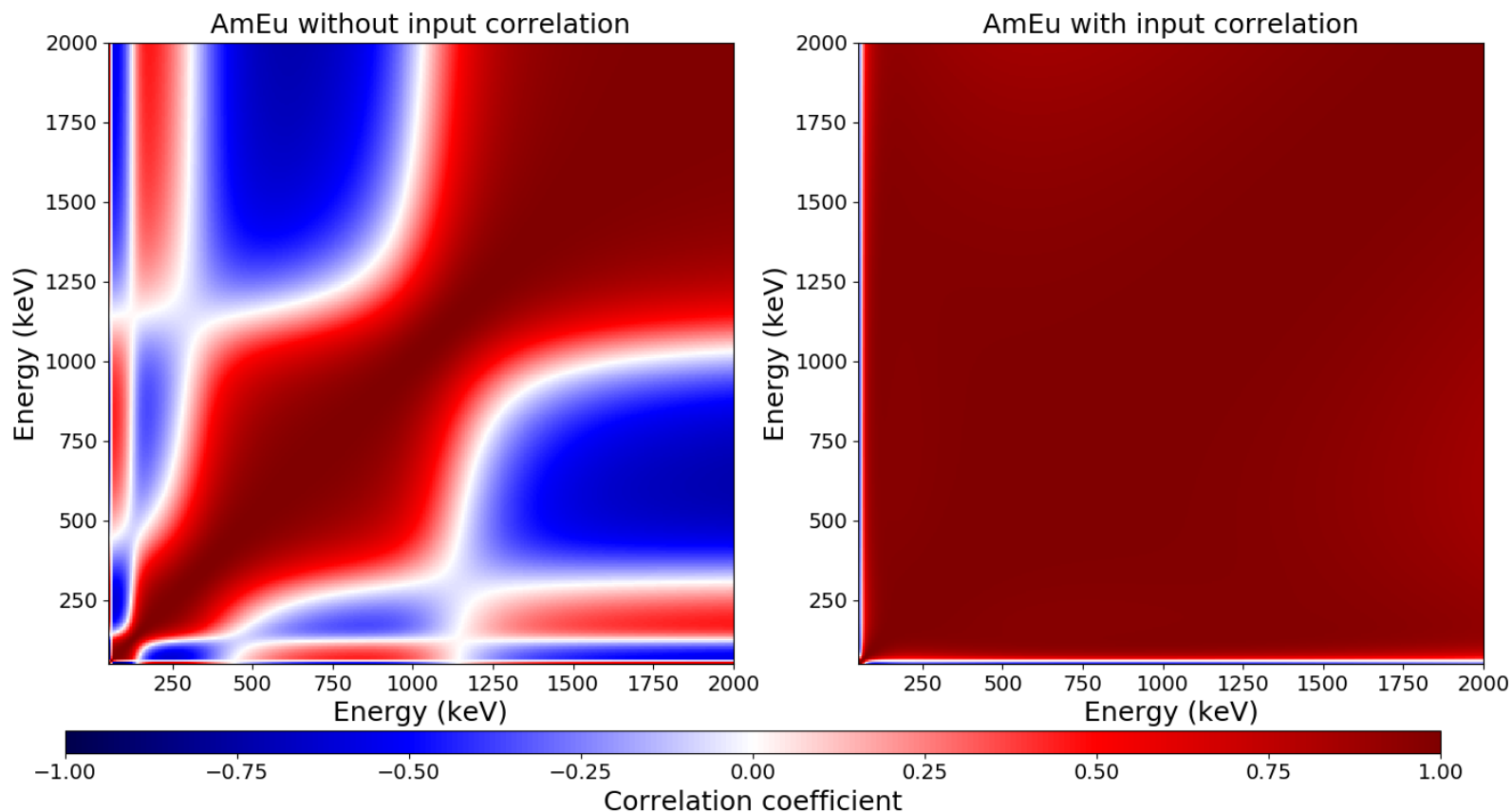
# Changes in Efficiency and Uncertainty with Correlated Data



Ratios of efficiency values calculated with and without source standard correlations (top);  
Efficiency uncertainties with (orange) and without (blue) source standard correlations



## Correlations Between Calculated Efficiencies



The **Correlation Coefficient** is a *relative* measure of covariance, between -1 and +1

$$\phi_{1,2} = \frac{\text{COV}(\varepsilon_1, \varepsilon_2)}{\sigma(\varepsilon_1)\sigma(\varepsilon_2)}$$



## NID with Interference Correction Engine

- ◆ A New Version of the Genie 2000 NID with Interference Correction Algorithm was also prototyped in Python

- ▶ Calculates correlations between measured nuclide line activities using measured data and correlations between calculated efficiency values

$$V_{ij} = \frac{N_i N_j \text{cov}(\varepsilon_i, \varepsilon_j)}{\varepsilon_i \varepsilon_j \varepsilon_i \varepsilon_j}$$

- ▶ Modified Least-Squares fitting routine correctly incorporates data correlation matrix for Weighted Mean Activity and Interference Correction fits.

$$\chi^2 = \sum_{ij} \left( \frac{N_i}{\varepsilon_i} - \sum_{\alpha} A_{\alpha} I_{\alpha i} \right) V_{ij}^{-1} \left( \frac{N_j}{\varepsilon_j} - \sum_{\beta} A_{\beta} I_{\beta j} \right)$$

- ▶ Outputs: Weighted Mean Activities and Weighted Mean Activity Uncertainties, which are stored in spectral data file
- ◆ No need for User to calculate or enter correlations
- ◆ Algorithm tested using simulated data (shown next)



## Weighted Mean NID Examples: Testing Approach

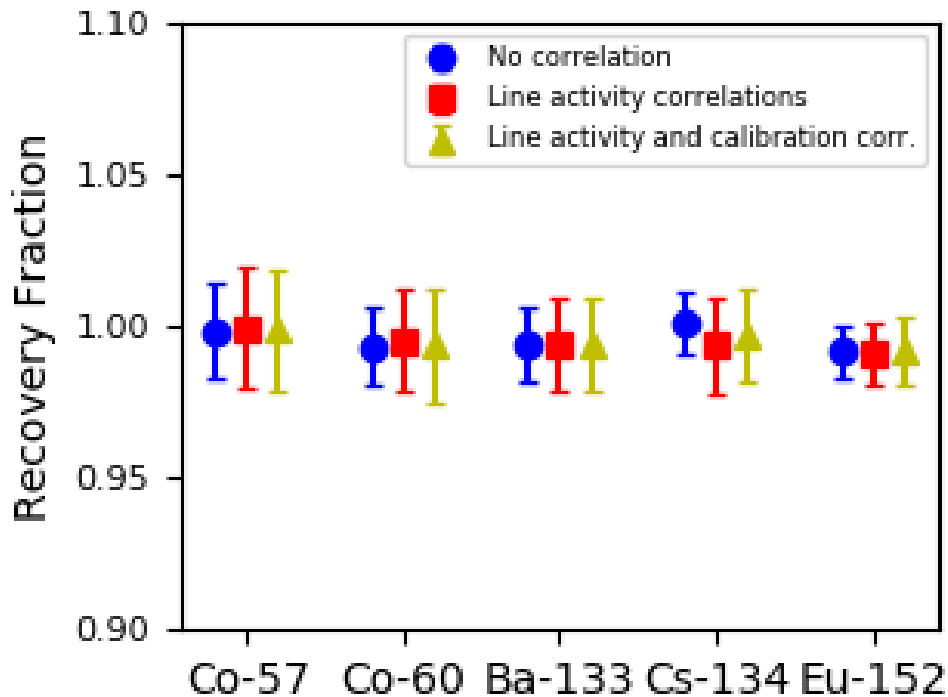
- ◆ Five simulated sample spectra were created for multi-line nuclides to demonstrate the effects of correlations on NID activity uncertainties
- ◆ Each Spectra was analyzed with each of the three calibrations:
  - I. Mixed Gamma (Low Correlation),
  - II. AmBaCsCo (medium Correlation),
  - III. AmEu (High Correlation)
- ◆ Counting uncertainties small compared to efficiency uncertainties
- ◆ NID Analysis was performed with:
  1. Genie Today (no correlations),
  2. New Fitting method with ability to take into account correlations (but no input efficiency correlations identified),
  3. New fitting method & efficiency input correlations identified.

Sample nuclides and energies for NID spectra

Nuclide	Energy	Nuclide	Energy	Nuclide	Energy
CO-60	1173.2	CO-57	122	Eu-152	121.8
	1332.5		137		244.7
			344.3		
			778.9		
			867.3		
			964.0		
			1085.8		
			1112.0		
			1408.0		
BA-133	79.6	CS-134	475.4		
	81.0		563.2		
	276.4		569.3		
	302.8		604.7		
	356.0		795.9		
383.9	802.0				
			1168.0		

## Impact on Weighted Mean Activity Uncertainties - I

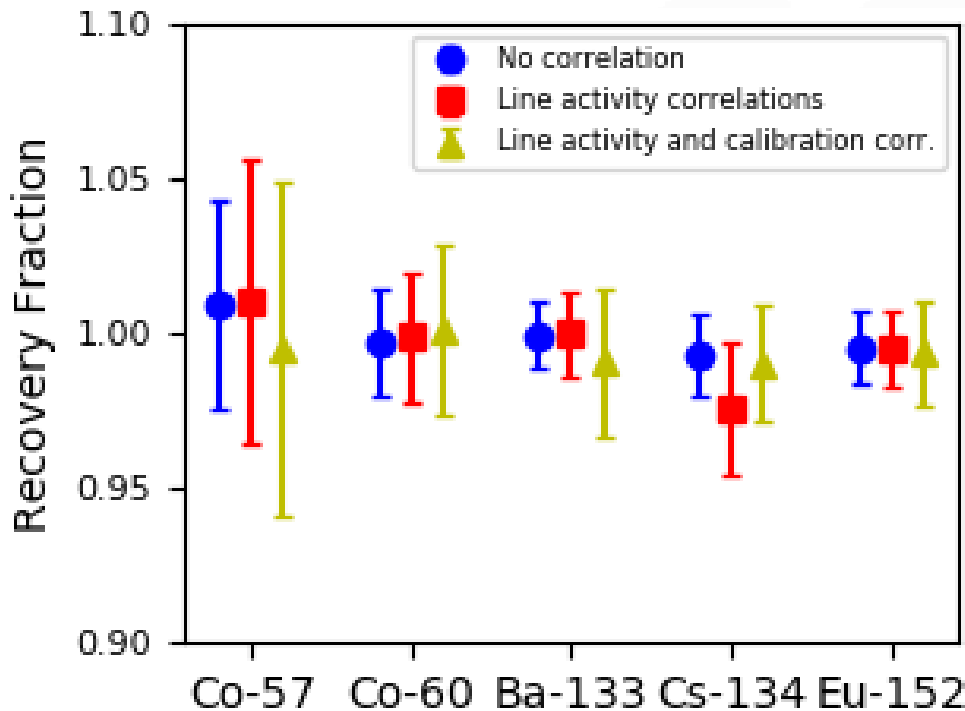
- Mixed Gamma Calibration Standard (minimal correlation)
- “Dual” Efficiency Model
- Changes in reported activities were minimal
- Changes in reported uncertainties were more significant:



Sample Nuclide	No Correlation	Line Activity Correlations Only	Line Activity & Cal. Standard Correlations
CO-57	1.574%	1.991%	1.986%
CO-60	1.275%	1.699%	1.890%
BA-133	1.248%	1.522%	1.501%
CS-134	1.024%	1.603%	1.538%
EU-152	0.685%	1.036%	1.104%

## Impact on Weighted Mean Activity Uncertainties - II

- AmBaCsCo Calibration Standard (moderate correlations)
- Dual Curve (crossover at 276 keV)
- Similar results to previous (activities show minimal change)
- Line activities generally show greater correlation

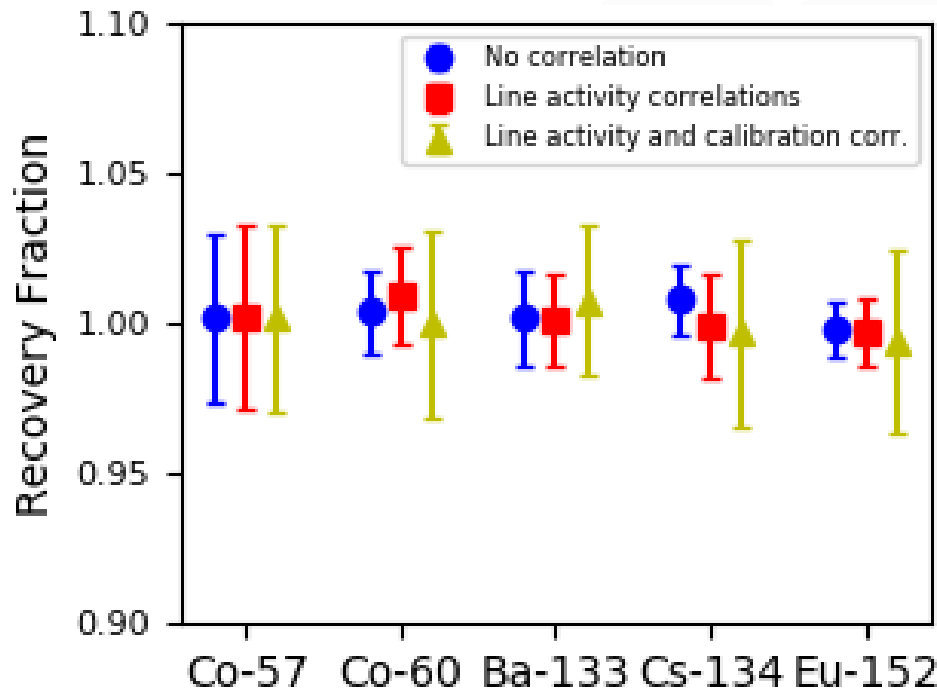


Sample Nuclide	No correlation	Line Activity Correlations Only	Line Activity & Cal. Standard Correlations
CO-57	3.319%	4.506%	5.425%
CO-60	1.733%	2.105%	2.729%
BA-133	1.081%	1.360%	2.390%
CS-134	1.347%	2.192%	1.869%
EU-152	0.888%	1.232%	1.707%



## Impact on Weighted Mean Activity Uncertainties - III

- AmEu Calibration Standard (greatest correlation)
- Linear Efficiency Model
- Similar results to previous (activities show minimal change)
- When Calibration Standard correlations are included, final activity uncertainties > Cal Source Uncertainty (3%) (except Ba-133)



Sample Nuclide	No correlation	Line Activity Correlations Only	Line Activity & Cal. Standard Correlations
CO-57	2.757%	3.039%	3.092%
CO-60	1.349%	1.646%	3.058%
BA-133	1.586%	1.554%	2.450%
CS-134	1.144%	1.719%	3.087%
EU-152	0.756%	1.107%	3.073%



## Interference Correction Examples – Testing Approach

- Simulated sample spectra with Se-75, Co-57, and Eu-152 were used to demonstrate the effects of correlations on Interference Correction results
- Both Co-57 lines are shared by other nuclides, so activity quantification/identification relies on interference correction to identify excess counts at 122 & 136 keV

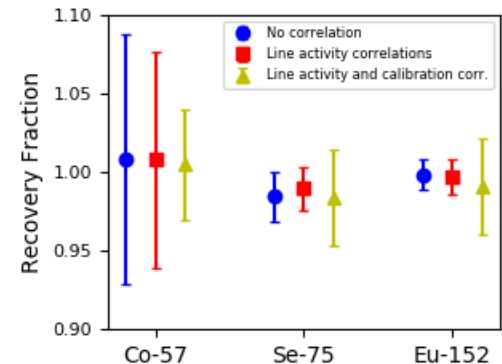
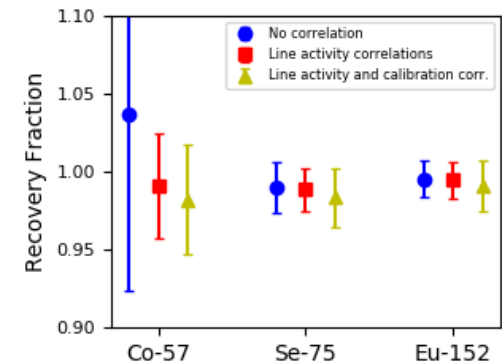
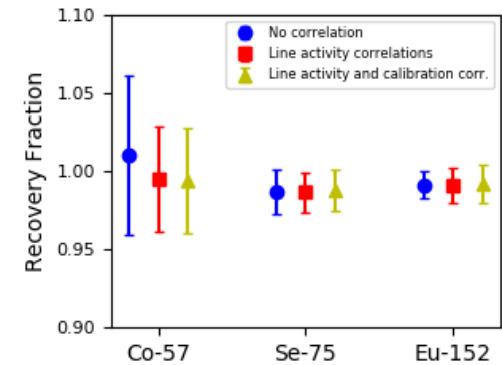
Se-75	Co-57	Eu-152
97		
122	122	122
136	136	
		245
265		
280		
		344
400		
		778
		867
		964
		1086
		1112
		1408

Gamma Energies of Interference Nuclides (keV)

# Impact on Interference Correction Uncertainties - Results

## Activity Uncertainties for Interference Correction Nuclides

Calibration Standard	NID Nuclide	No Correlation	NID Correlation Effects	NID & Cal. Std. Correlation effects.
Mixed Gamma	CO-57	4.992%	3.378%	3.36%
	SE-75	1.447%	1.257%	1.30%
	EU-152	0.719%	1.103%	1.19%
AmBaCsCo	CO-57	11.586%	3.359%	3.49%
	SE-75	1.663%	1.382%	1.86%
	EU-152	0.893%	1.211%	1.65%
AmEu	CO-57	7.819%	6.956%	3.55%
	SE-75	1.607%	1.368%	3.13%
	EU-152	0.770%	1.133%	3.09%





## Conclusions and where we are today

- ◆ The conclusion: approach works well for ideal data sets
- ◆ For more challenging “realistic” data sets
  - ▶ For example, when the efficiency calibration does not quite match the efficiency response or peak fitting is incorrect
  - ▶ Preliminary testing shows the algorithm is extremely sensitive to inconsistencies in the data
  - ▶ A peak selection algorithm is being explored to input only the most accurate data into the algorithm
  - ▶ Fall back to traditional (non-correlated) algorithm if data is too inconsistent



## Conclusions and where we are today

- ◆ Enhanced analysis engines in development for Genie 2000 to correctly account for correlations
  - ▶ Between measured efficiency calibration data
  - ▶ Between nuclide line activities
- ◆ Correlations generally have minimal effect on activity values, but could change some single line nuclide results or interference corrected activities by a percent or more in some cases
- ◆ Correlations have more impact on correctly defining the activity uncertainty, with changes of several percent in uncertainty for common scenarios
- ◆ These effects could be particularly important for high-precision applications
- ◆ Contact: [Karaphillips@mirion.com](mailto:Karaphillips@mirion.com)