

Covariance Analysis of Experimental Shape Data

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Keegan J. Kelly,
J.M. O'Donnell, D. Neudecker, M. Devlin, J.A. Gomez

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The analysis of shape data including normalization and the impact on prompt fission neutron spectrum measurements



K.J. Kelly *, J.M. O'Donnell, D. Neudecker, M. Devlin, J.A. Gomez

Why Shape Data are Different

The basic problem:

- Shape data (PFNS shape, probability distributions, relative measurements, etc.) are inherently different than absolute data (absolute cross sections, etc.)
- Primary difference is in the covariance treatment

Typical Experimentalist Treatment:

- I found a fully-correlated uncertainty source (beam flux, scaling factor)
 - I'm trying to report a shape result, so I can ignore it
- I found a partially correlated uncertainty source (most sources)
 - Keep it, or ignore part of it, or state that it's correlated, or something else...
 - How much do I keep? What is the covariance of the reduced uncertainty??
 - Give to evaluator and move on

A Normalization Procedure with Covariance Propagation Solves This Problem

The Basic Idea for the Covariance Math

Un-norm. Covariance \rightarrow

$$\text{cov}[p]_{ij} = \sigma_{ij}$$

Normalized Data \rightarrow

$$n_i = S \frac{p_i}{\sum_j p_j w_j} = S \frac{p_i}{A} \propto \frac{p_i}{A}$$

Sensitivity Matrix \rightarrow

$$\begin{aligned}\Delta_{ik} &= \frac{\partial n_i}{\partial p_k} \\ &= \frac{\delta_{ik}}{A} - \frac{p_i}{A^2} w_k\end{aligned}$$

Norm. Covariance \rightarrow

$$\begin{aligned}\text{cov}[n]_{ij} &= \sum_k \sum_l \Delta_{ik} \sigma_{kl} \Delta_{jl} \\ &= \sum_k \sum_l \left(\frac{\delta_{ik}}{A} - \frac{p_i}{A^2} w_k \right) \left(\frac{\delta_{jl}}{A} - \frac{p_j}{A^2} w_l \right) \sigma_{kl}\end{aligned}$$

Correlation Matrix \rightarrow

$$\text{cor}[n]_{ij} = \frac{\text{cov}[n]_{ij}}{\sqrt{\text{cov}[n]_{ii} \text{cov}[n]_{jj}}}$$

Expected: Fully-Correlated Uncertainties Drop Out

Consider a covariance source of the form: $\sigma_{ij} = f^2 p_i p_j$

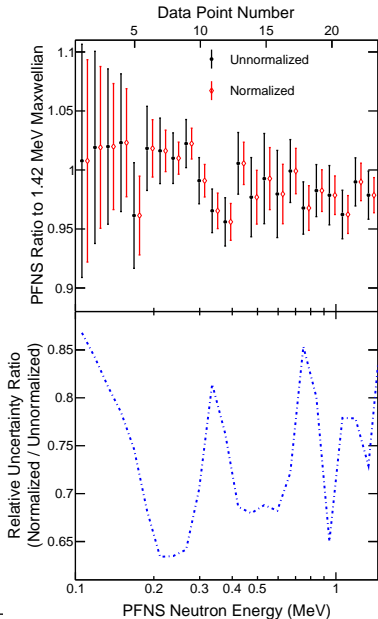
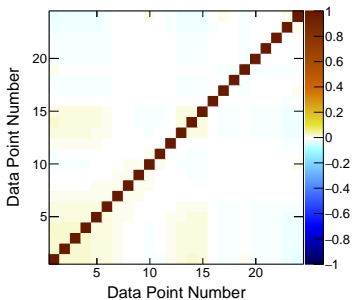
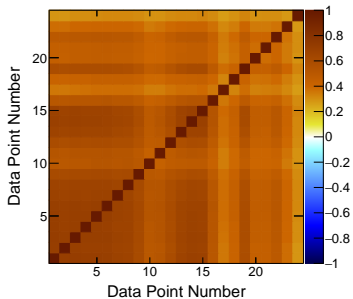
$$\text{COV} [n]_{ij} = \sum_k \sum_l \left(\frac{\delta_{ik}}{A} - \frac{p_i}{A^2} w_k \right) \left(\frac{\delta_{jl}}{A} - \frac{p_j}{A^2} w_l \right) f^2 p_k p_l.$$

$$\begin{aligned} \text{COV} [n]_{ij} &= \sum_k \sum_l \left(\frac{\delta_{ik} \delta_{jl}}{A^2} - \delta_{ik} \frac{p_j w_l}{A^3} - \delta_{jl} \frac{p_i w_k}{A^3} + \frac{p_i p_j w_k w_l}{A^4} \right) f^2 p_k p_l \\ &= \frac{f^2 p_i p_j}{A^2} \sum_k \sum_l \left(1 - \frac{p_l w_l}{A} - \frac{p_k w_k}{A} + \frac{p_k p_l w_k w_l}{A^2} \right), \\ &= \frac{f^2 p_i p_j}{A^2} \left(1 - \sum_l \frac{p_l w_l}{A} - \sum_k \frac{p_k w_k}{A} + \sum_k \frac{p_k w_k}{A} \sum_l \frac{p_l w_l}{A} \right), \end{aligned}$$

Each summation adds to unity, and so

$$\text{COV} [n]_{ij} = 0.$$

Expected: Fully-Correlated Uncertainties Drop Out



Maybe Unexpected: Redistribution of Covariances

Consider data set in which one data point, z , is known perfectly, so

$$\sigma_{zz} = 0 \text{ and } \sigma_{iz} = 0 \text{ (no uncertainty, no correlations for point } z)$$

but, $\sigma_{ij} \neq 0$ for all $i, j \neq z$

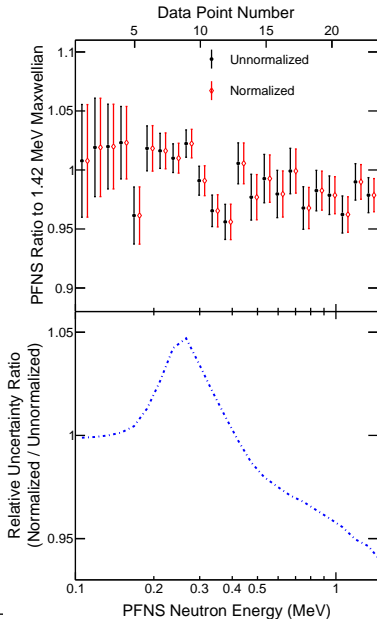
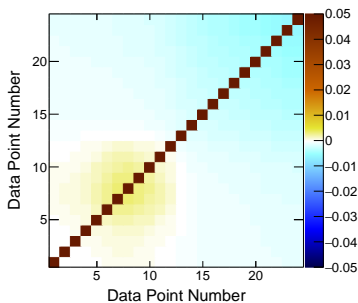
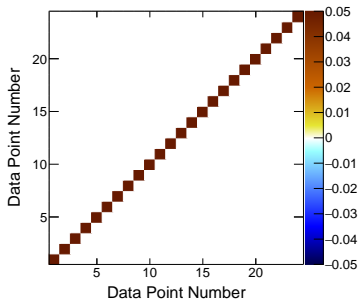
$$\text{cov} [n]_{iz} = \sum_{k \neq z} \sum_{l \neq z} \left(\frac{\delta_{ik}}{A} - \frac{p_i}{A^2} w_k \right) \left(-\frac{p_z}{A^2} w_l \right) \sigma_{kl},$$

- This is generally non-zero, and therefore point z has nonzero variance and covariance with other points, i, j

A well-measured data point gets a larger 1-D uncertainty assigned to it

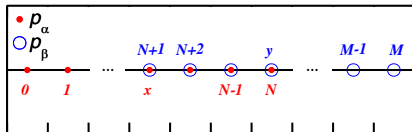
- Odd to most experimentalists, but the *shape* is properly represented

Maybe Unexpected: Redistribution of Covariances



Necessary: Consistency When Combining Shapes

- There are choices to make when combining shape data sets
(relevant to reporting data from Chi-Nu PFNS results)
- Only one single, correct shape exists for a set of data points



Independent Axis (Energy, time, etc.)

$$\rho = \left[\sum_{i=x}^N p_i w_i \right] \times \left[\sum_{j=N+1}^y p_j w_j \right]^{-1}$$

$$u_i^{\beta \rightarrow \alpha} = \begin{cases} p_{\alpha i} & \text{if } 0 \leq i \leq N \\ \rho p_{\beta i} & \text{if } N < i \leq M \end{cases}$$

$$u_i^{\alpha \rightarrow \beta} = \begin{cases} \rho^{-1} p_{\alpha i} & \text{if } 0 \leq i \leq N \\ p_{\beta i} & \text{if } N < i \leq M \end{cases}$$

Necessary: Consistency When Combining Shapes

$$\text{cov} [u]_{ij}^{\beta \rightarrow \alpha} = \sum_k \sum_l \left(\delta_{ik} \rho^{\theta_\beta(i)} + p_i \frac{\partial \rho^{\theta_\beta(i)}}{\partial p_k} \right) \left(\delta_{jl} \rho^{\theta_\beta(j)} + p_j \frac{\partial \rho^{\theta_\beta(j)}}{\partial p_l} \right) \sigma_{kl}$$

$$\text{cov} [u]_{ij}^{\alpha \rightarrow \beta} = \sum_k \sum_l \left(\delta_{ik} \rho^{-\theta_\alpha(i)} + p_i \frac{\partial \rho^{-\theta_\alpha(i)}}{\partial p_k} \right) \left(\delta_{jl} \rho^{-\theta_\alpha(j)} + p_j \frac{\partial \rho^{-\theta_\alpha(j)}}{\partial p_l} \right) \sigma_{kl}$$

These covariances are not equal in general.

Normalize the combined data set to recover the single correct combined shape

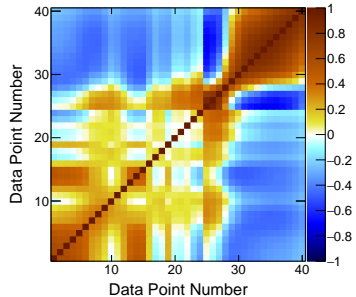
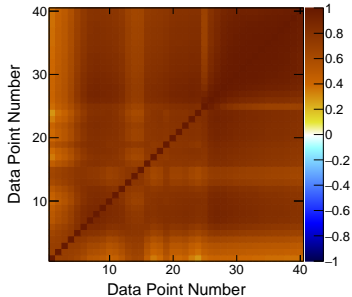
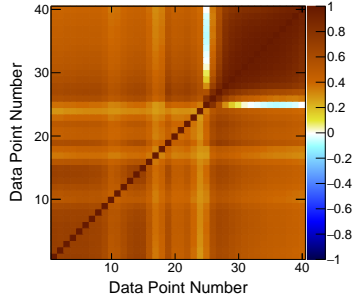
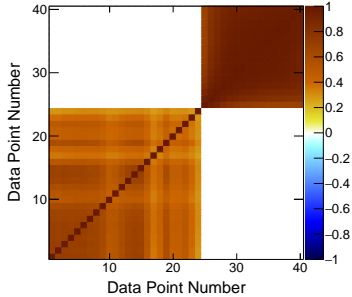
$$n_i^{\beta \rightarrow \alpha} = \begin{cases} p_i / A_T & \text{if } 0 \leq i \leq N \\ \rho p_i / A_T & \text{if } N < i \leq M \end{cases} = \frac{\rho^{\theta_\beta(i)} p_i}{A_T}$$

$$\text{cov} [n]_{ij}^{\beta \rightarrow \alpha} = \sum_k \sum_l \Delta_{ik}^{\beta \rightarrow \alpha} \sigma_{kl} \Delta_{jl}^{\beta \rightarrow \alpha}$$

After some work, $\Delta_{ik}^{\alpha \rightarrow \beta} = \Delta_{ik}^{\beta \rightarrow \alpha}$

$$\Rightarrow \text{cov} [n]_{ij}^{\alpha \rightarrow \beta} = \text{cov} [n]_{ij}^{\beta \rightarrow \alpha}$$

Necessary: Consistency When Combining Shapes



Conclusions, Acknowledgements, References

- Normalization \neq Scaling, even if scaling to have $A = 1$
- Covariance propagation of normalization of an unnormalized experimental data set yields the covariance matrix of the shape
- Normalization is a necessary part of experimental shape data analysis
 - Experimental shapes could be misrepresented in evaluations otherwise...

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