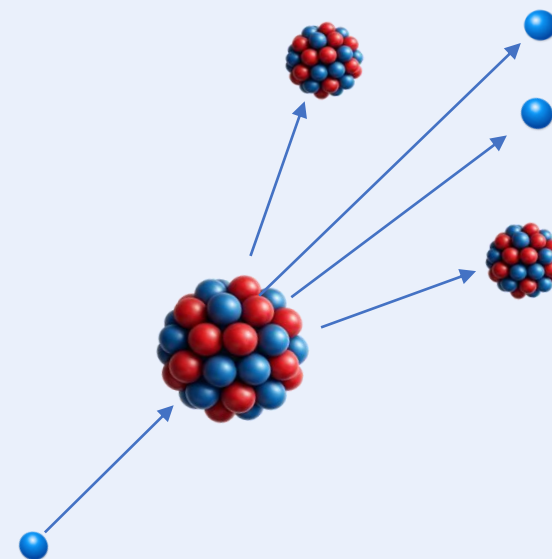


# Adjustment of $^{239}\text{Pu}$ evaluations using Bayesian Inference

Cross Section Evaluation Working Group – Brookhaven National Lab, NY

November 4, 2025



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Design Physics Division

# Calibrating to Criticality benchmarks with Bayesian Inference

- Use Importance sampling to estimate posterior means and covariances

**Godiva (HEU-MET-FAST 001):**

$$k_{\text{eff}} = 1.0 \pm 0.00213$$

**Jezebel (PU-MET-FAST 001):**

$$k_{\text{eff}} = 1.0 \pm 0.002$$

- Samples are drawn from the prior using new LLNL code EMU (Evaluated Means and Uncertainties)
- EMU can sample the ENDF covariance and generate stochastic libraries

$$P(x|D) \propto \mathcal{L}(D|x)P(x)$$

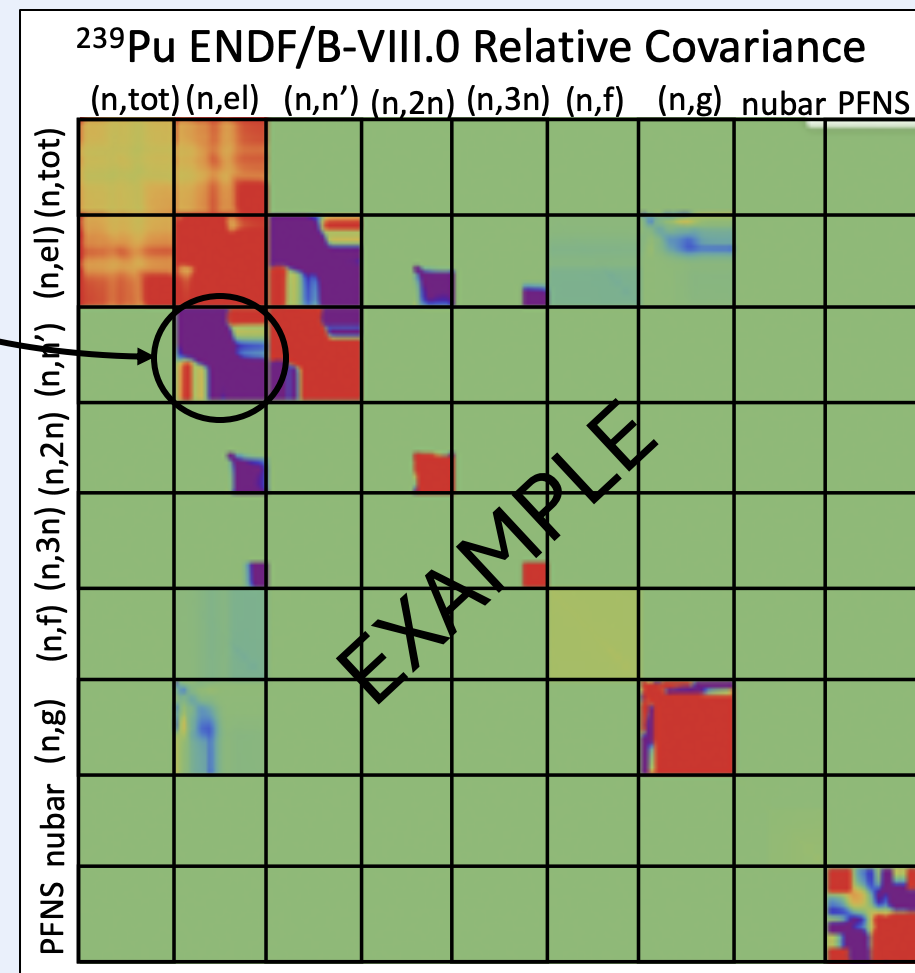
$$p(x|D) \propto \underbrace{\prod_{i=1}^N \exp \left[ -\frac{1}{2} \frac{\left( y_i^{(\text{th})}(\vec{x}) - y_i^{(\text{exp})} \right)^2}{\sigma_{e,i}^2 + \sigma_{m,i}^2} \right]}_{\text{Integral/Differential Data Likelihood}} \times \underbrace{\exp \left[ -\frac{1}{2} (\vec{x} - \vec{\mu})^T \Sigma^{-1} (\vec{x} - \vec{\mu}) \right]}_{\text{Nuclear Data Prior}}$$

## Importance Sampling

$$\hat{\sigma}_y^2 = \sum_{i=1}^N \tilde{w}(x^{(i)}) \left[ y(x^{(i)}) - \hat{\mu}_y \right]^2 \quad \hat{\mu}_y = \sum_{i=1}^N \tilde{w}(x^{(i)}) y(x^{(i)})$$

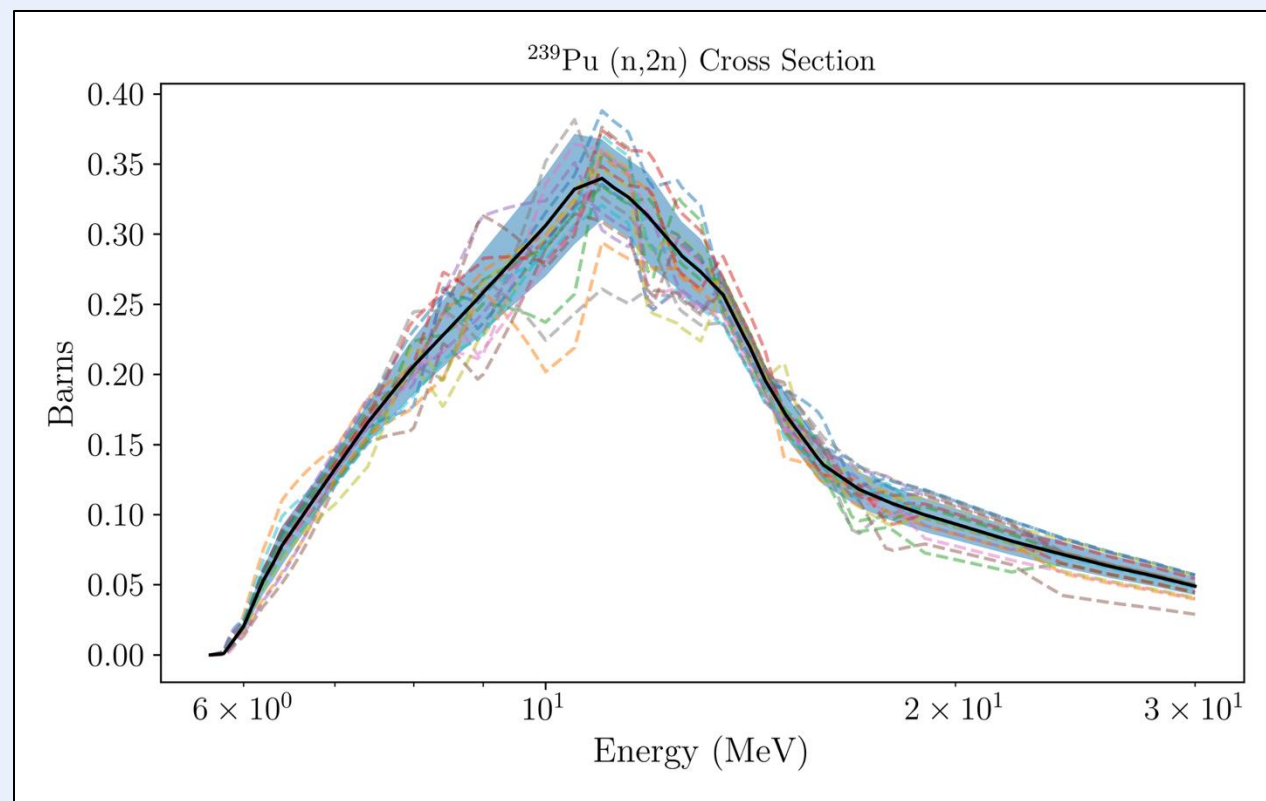
# Using EMU to generate samples

- Prior samples are taken from the ENDF/B-VIII.1 library
- Includes cross-channel correlations and constraints from the total cross section (no cross-material samples yet...)
- 1700 samples are taken to generate the prior distribution and each sample is weighted by the likelihood
- Sampling is expensive! **Importance sampling reduces this cost by weighting each sample appropriately**



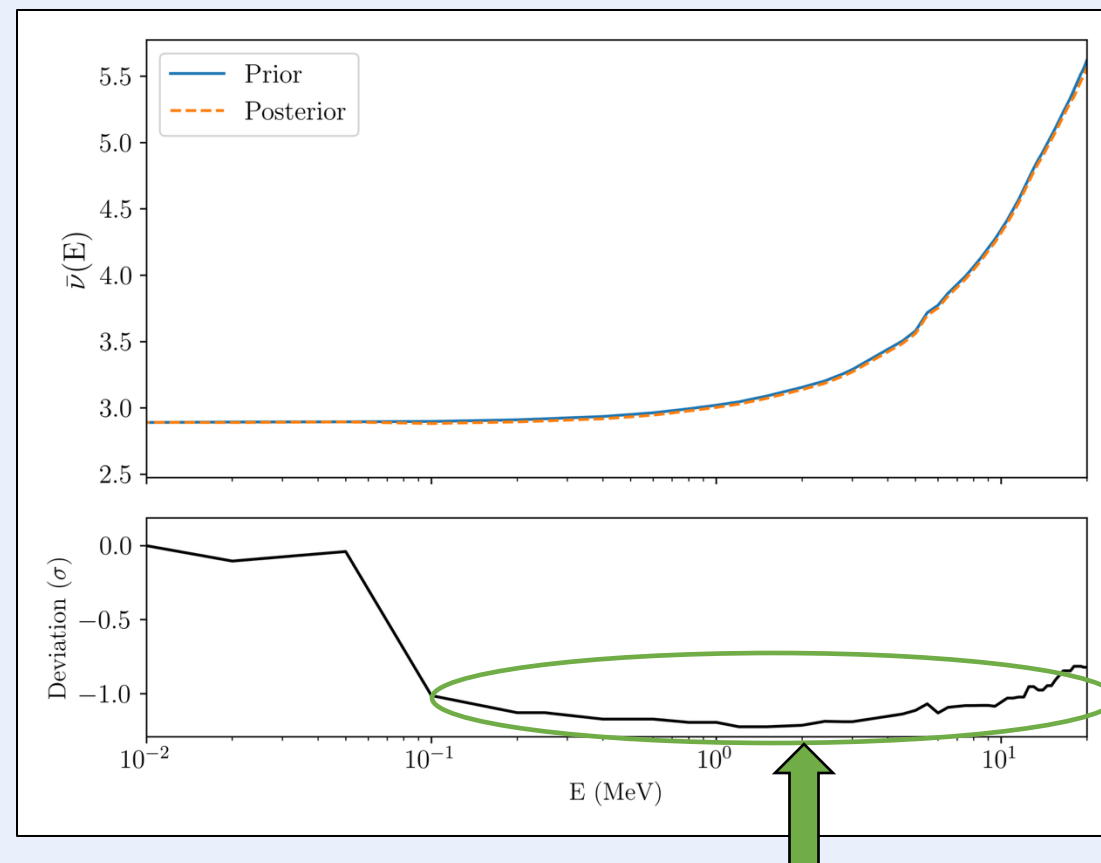
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# Taking fTPC as truth for $^{239}\text{Pu}$ fission cross section

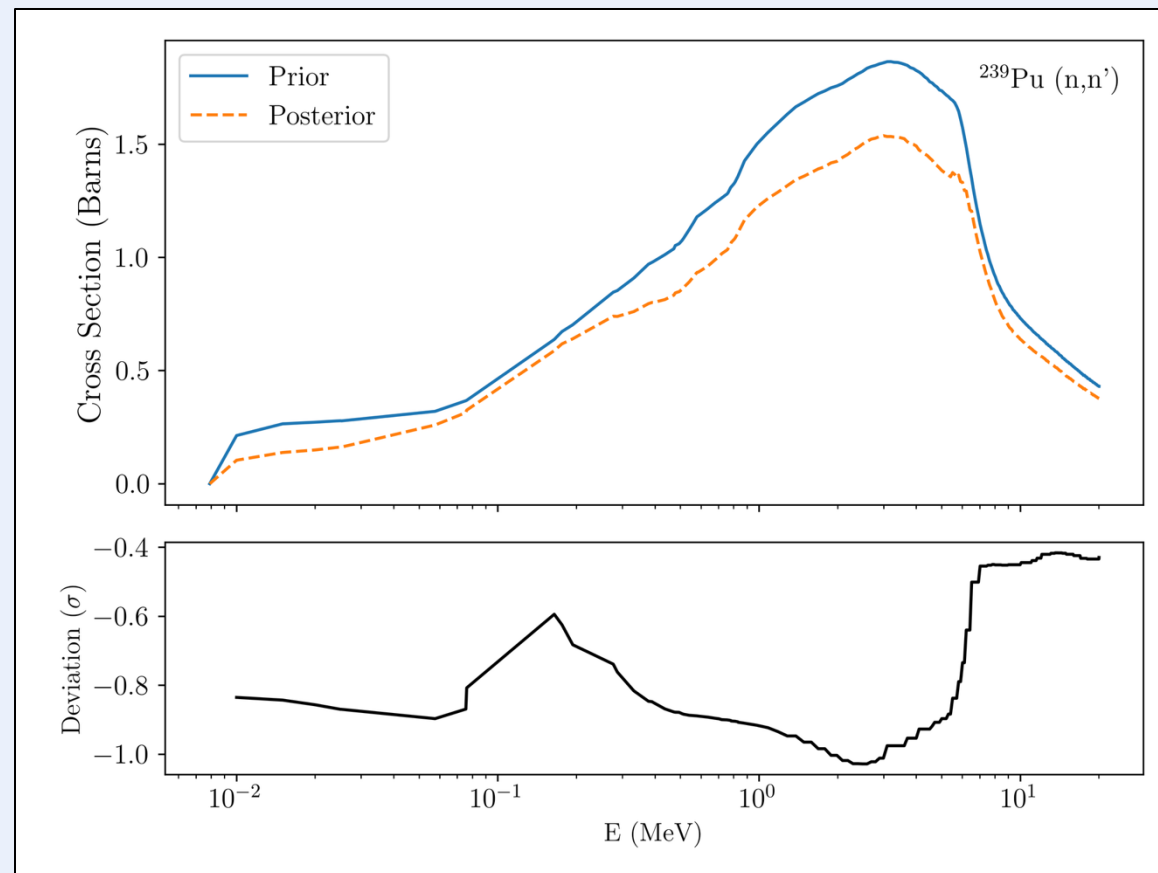
- First calibration was done by adjusting the fission cross section of  $^{239}\text{Pu}$  to fTPC data. Causes large discrepancies with Pu-based critical benchmark data, e.g. Jezebel  $\sim 1.01$
- Bayesian inference is used to infer other  $^{239}\text{Pu}$  degrees of freedom to match Jezebel benchmark again. These include: elastic,  $\bar{\nu}$ , capture, (n,2n), and inelastic
- Results show significant changes to  $\bar{\nu}$  and inelastic cross section
- Some changes to elastic cross section and minor changes to capture, (n,2n)



Over  $1\sigma$  change to  $\bar{\nu}$ !

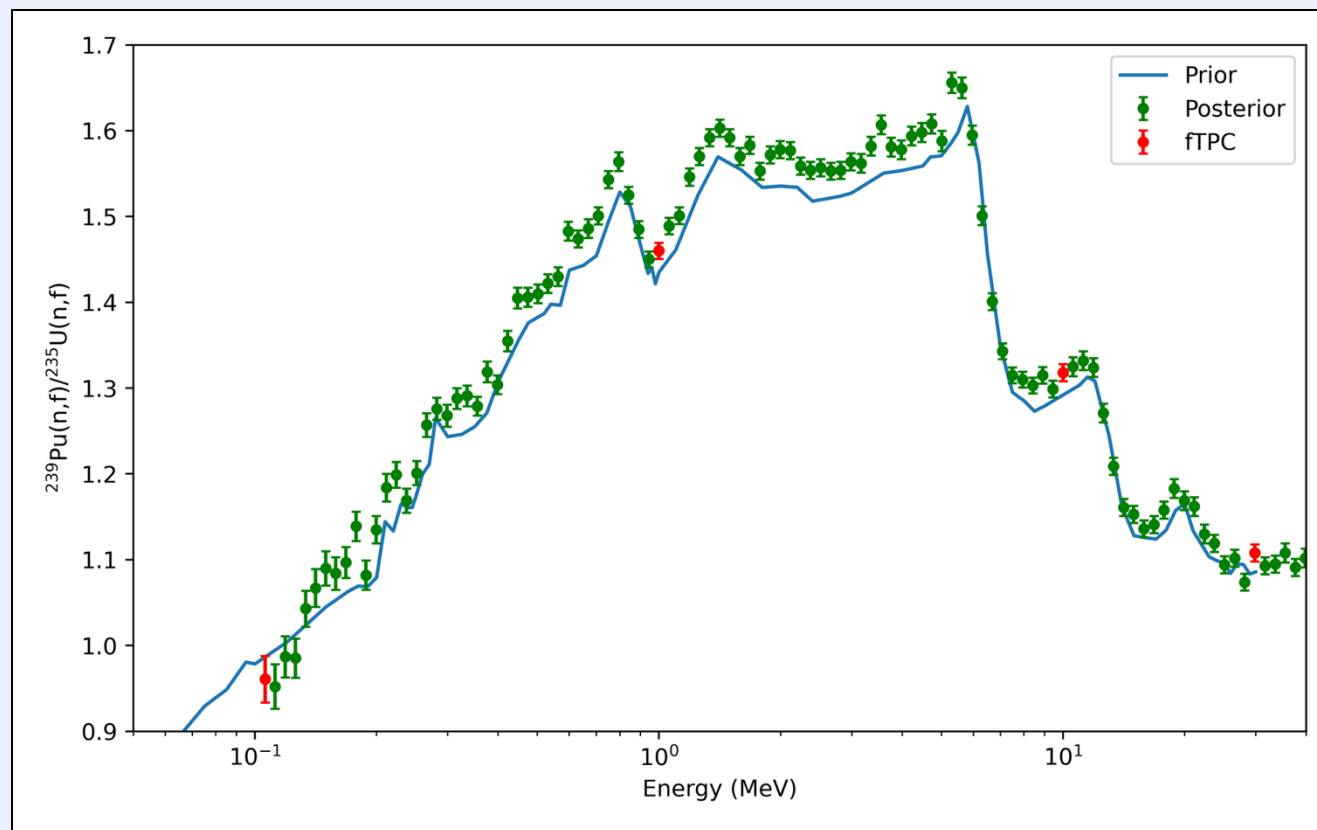
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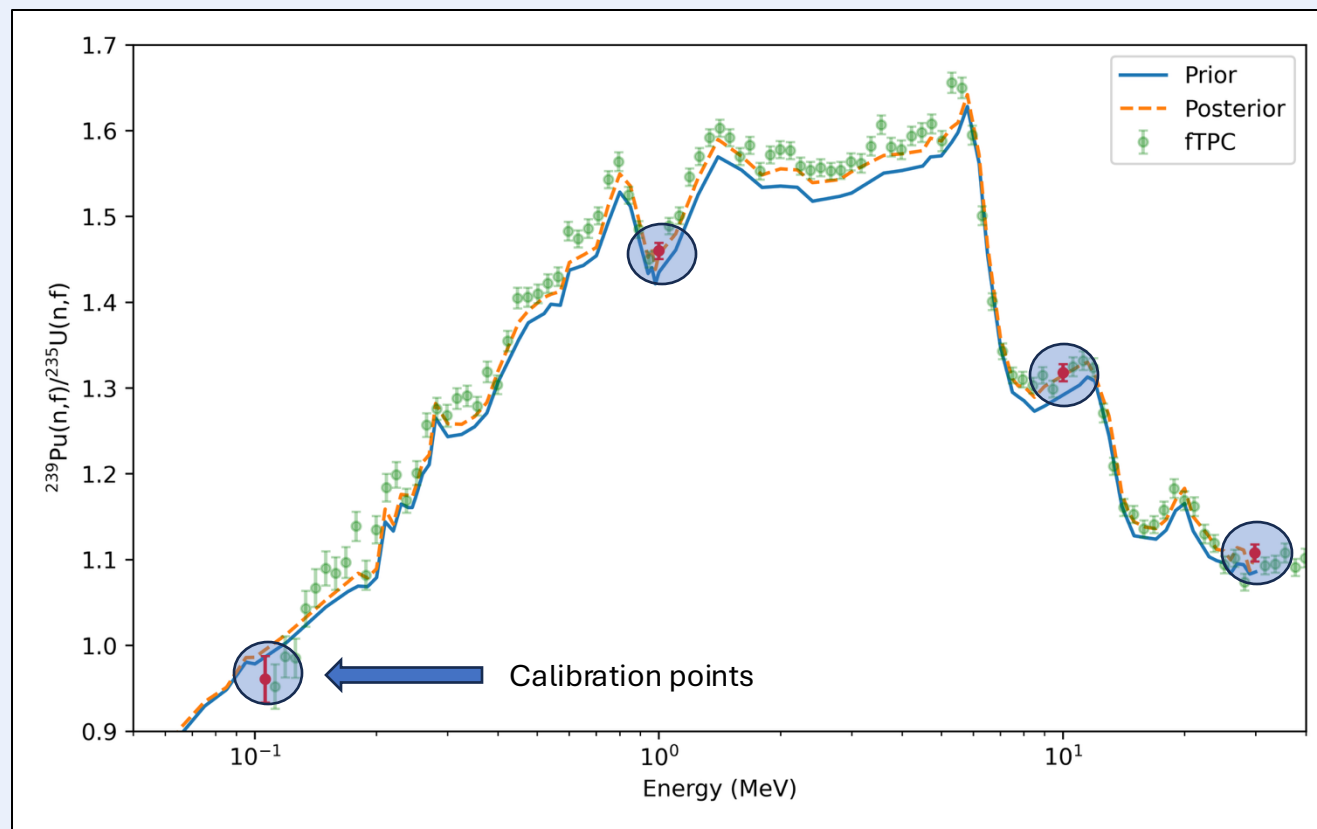
# Explicit calibration to fTPC data

- **Allow both  $^{239}\text{Pu}$  and  $^{235}\text{U}$  to fully vary according to the ENDF covariance**
- fTPC data and Godiva/Jezebel benchmarks will determine what nuclear data needs to change to meet experimental constraints
- Changes to both the  $^{239}\text{Pu}$  and  $^{235}\text{U}$  fission cross sections
- Less extreme changes to other reaction channels ( $<1\sigma$ )
- New covariance data for posterior distributions informed from fTPC and criticality constraints



# Explicit calibration to fTPC data

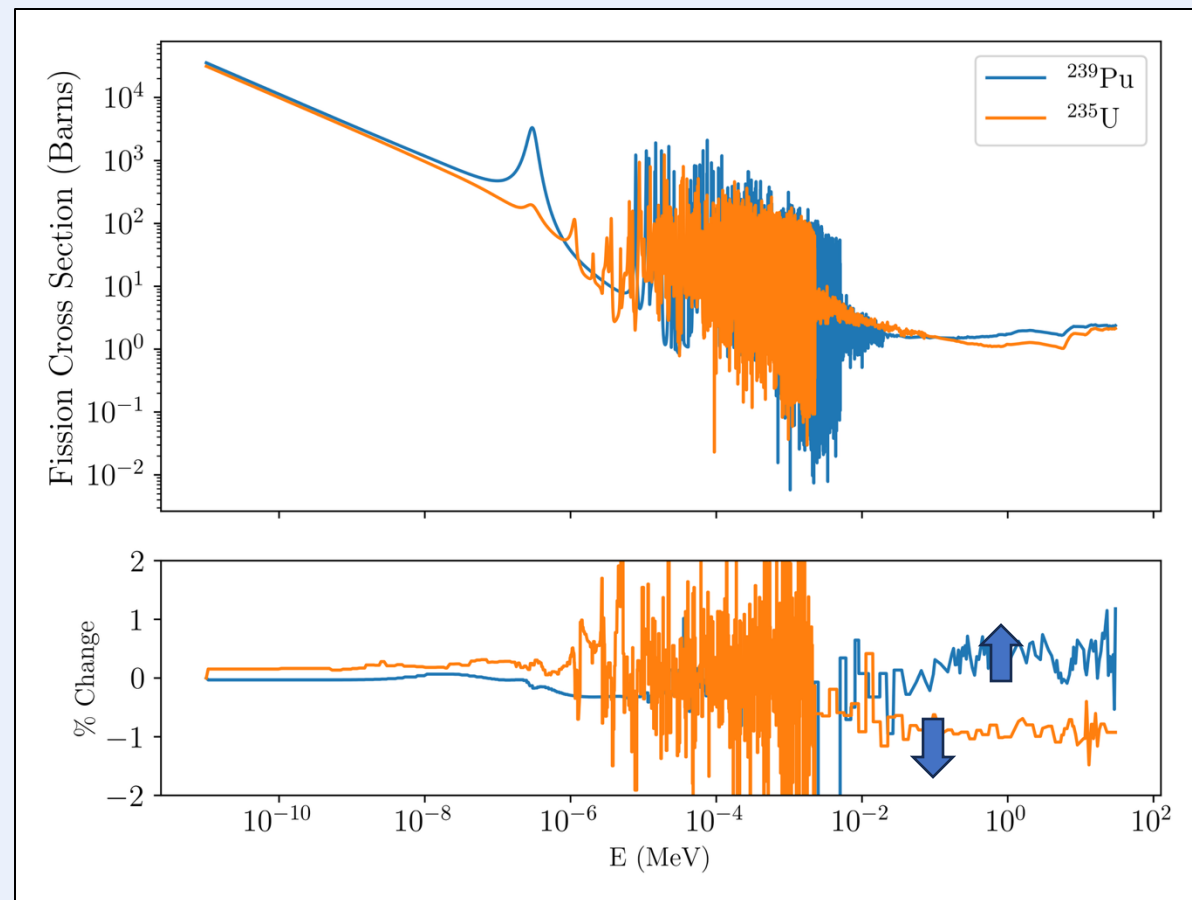
- Allow  $^{235}\text{U}$  to vary and make no assumptions about the underlying fission cross sections.
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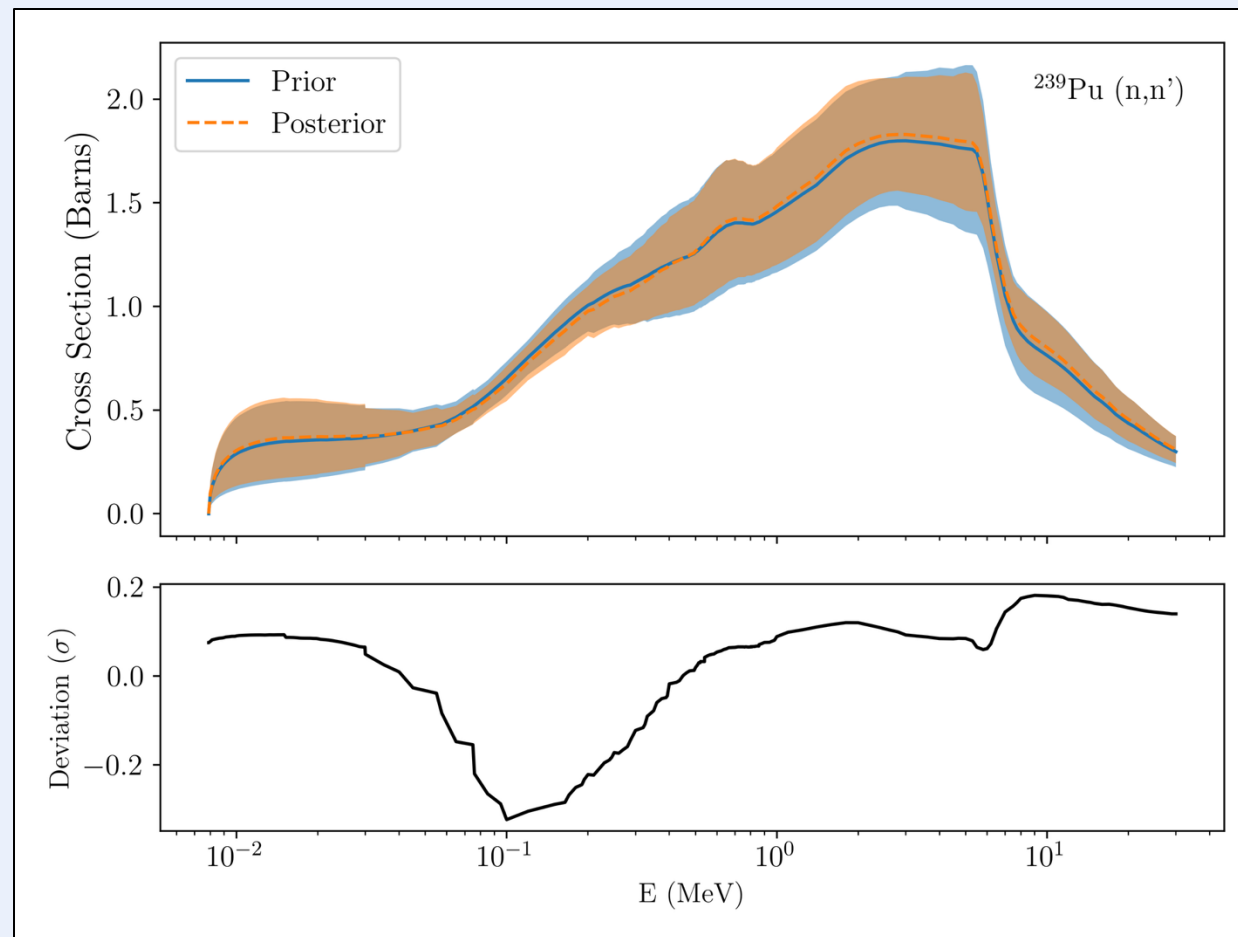
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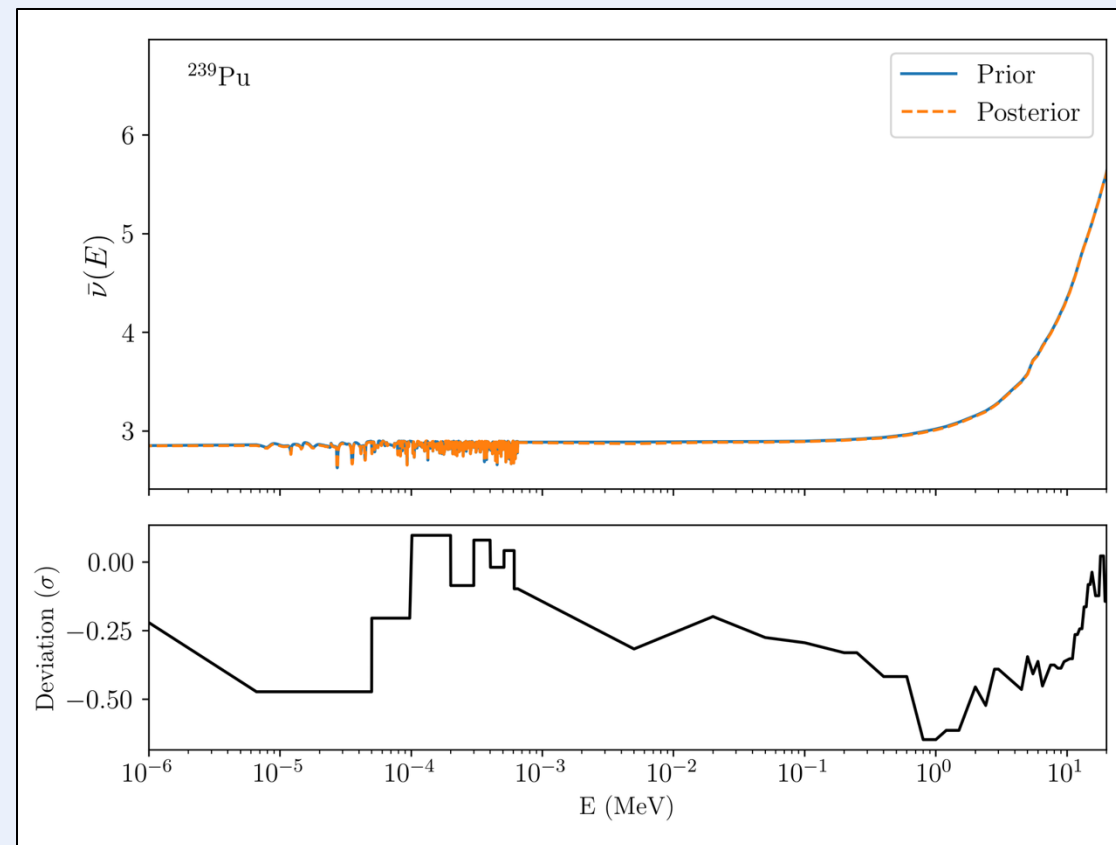
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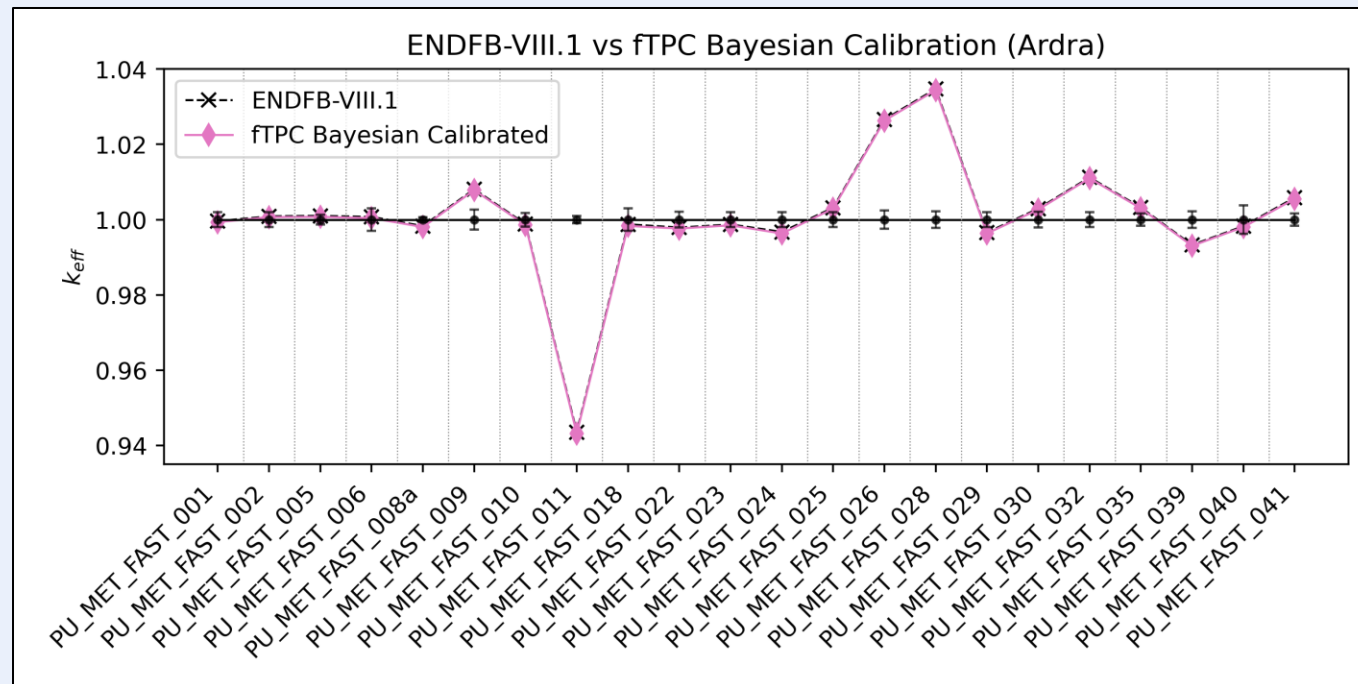
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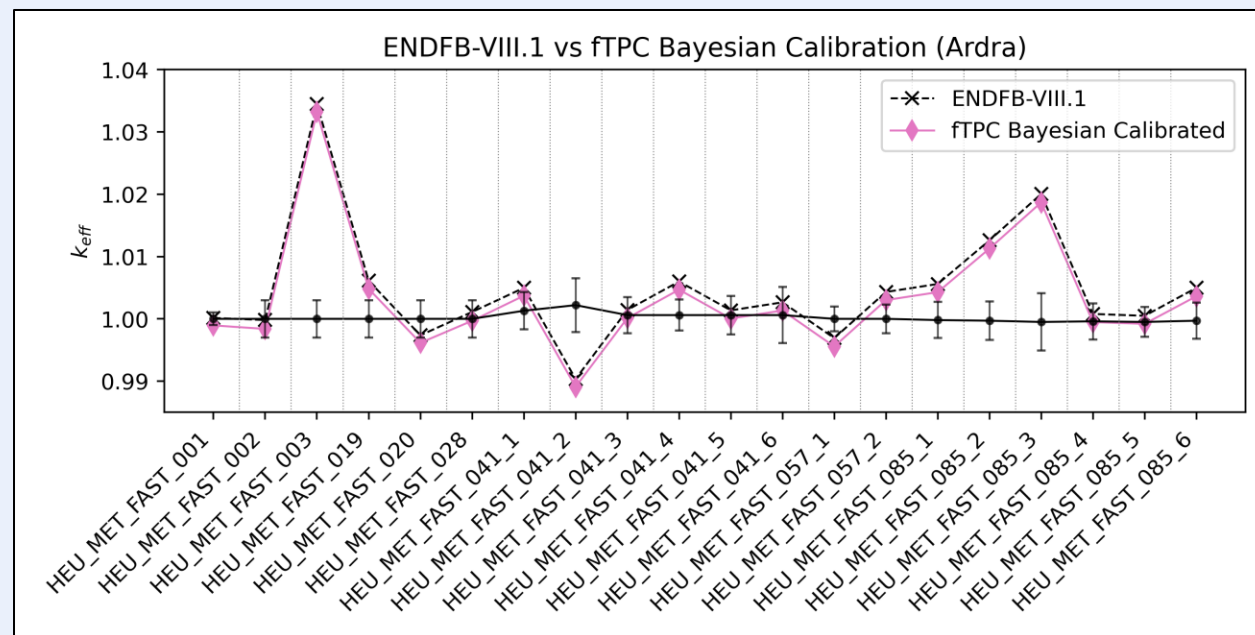
# Takeaways and Comparison with ENDF for Critical Assemblies

- Bayesian Inference can be a valuable tool for informing which reaction channels/nuclear data needs to change due to new experimental/theoretical constraints
- Can provide new covariance data to relatively unconstrained reaction channels, given the experiment is sensitive to the underlying channel
- Also provides cross channel correlations that might be helpful to the evaluation/physics community

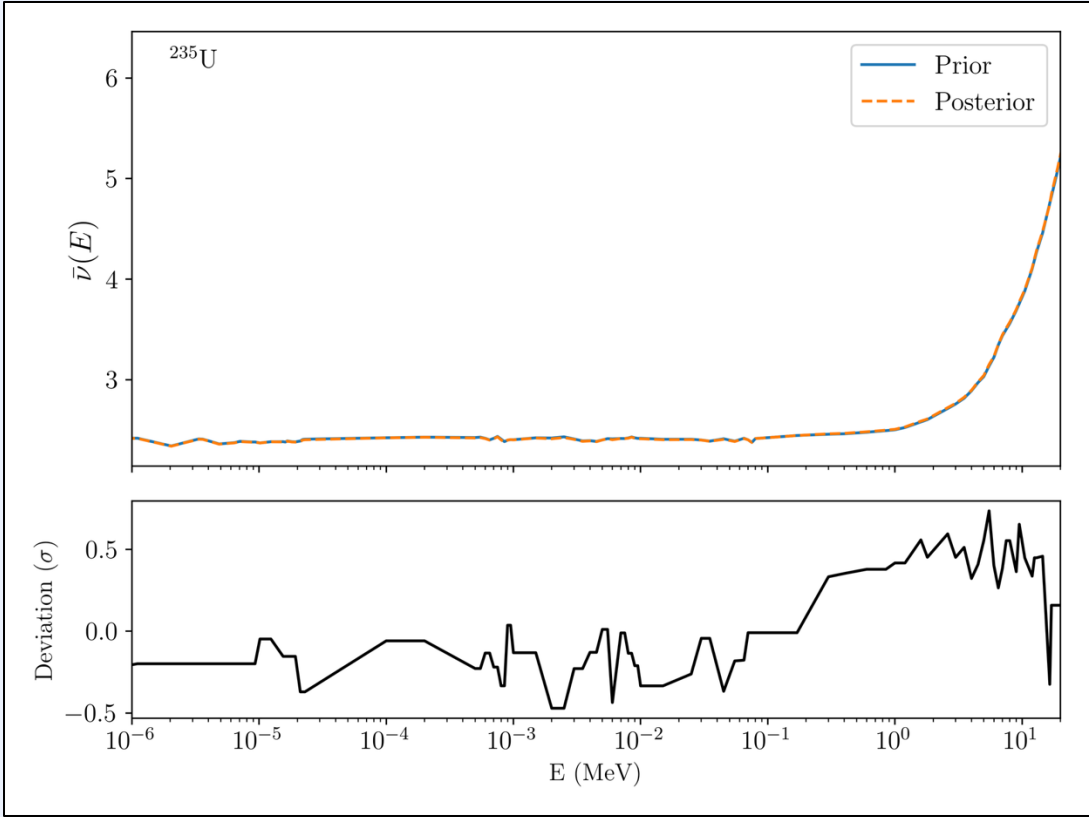
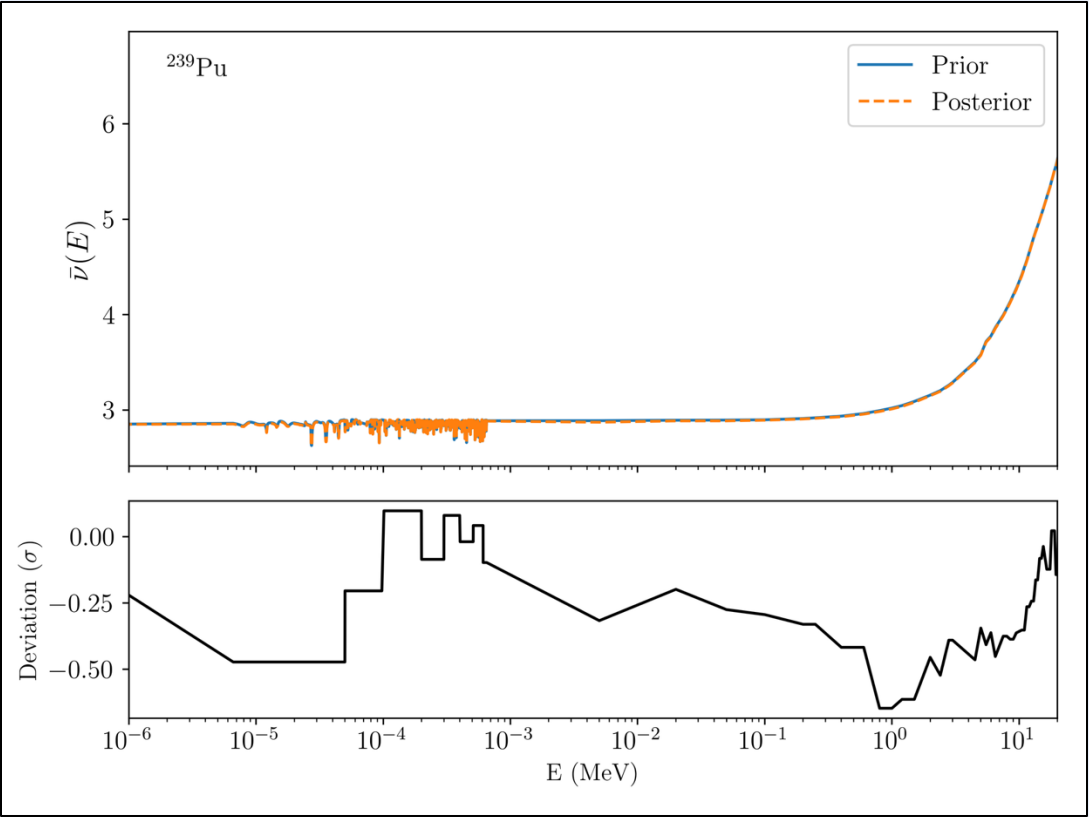


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# Neutron Multiplicities



# Some backup sources

- Updated uncertainties for Godiva taken from Bess, J. D., Briggs, J. B., & Marshall, M. A. (2013)