

# Methods for Sensitivity and Uncertainty Analysis in Nuclear Engineering Applications

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**University of New Mexico**

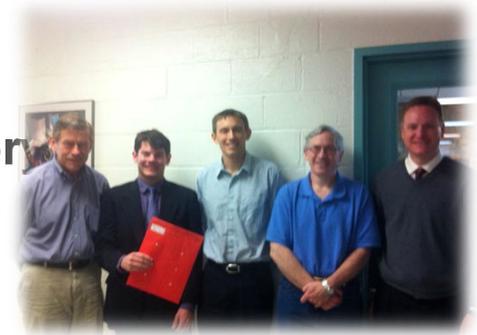


# Presentation Outline

- **Part 1 – Introduction to Sensitivity Coefficients**
- **Part 2 – Applications of Sensitivity Analysis**
  - **Design Optimization**
  - **Uncertainty Quantification**
  - **Benchmark Experiment Selection**
  - **Experimental Data Assimilation**
- **Part 3 – Current Research**

# About Me

- ❑ Born and raised in Florida
- ❑ Education:
  - ❑ 2007: B.S. in Nuclear Engineering, University of Florida
  - ❑ 2008: M.S. in Nuclear Engineering, University of Florida
  - ❑ 2012: PhD. in Nuclear Engineering, University of Michigan
- ❑ 2012-2018: R&D Scientist, Oak Ridge National Laboratory
- ❑ 2018-Present: Assistant Professor, University of New Mexico



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# Sensitivity Analysis and Uncertainty Propagation

Sensitivity coefficients provide insight on the sources and impact of uncertainty in nuclear engineering models.



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**Input Information:**  
Nuclear Data ( $\Sigma$ ), Atom  
Densities ( $N$ ), Material  
Densities ( $\rho$ )

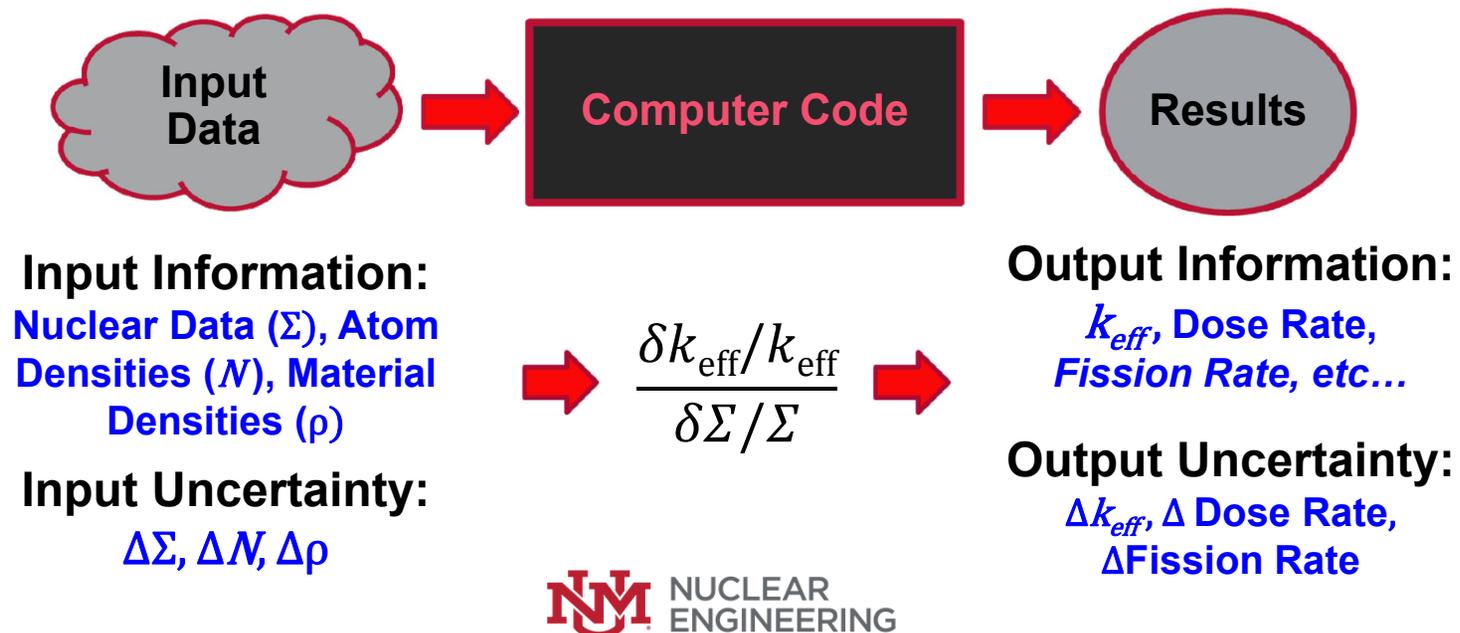
**Input Uncertainty:**  
 $\Delta\Sigma$ ,  $\Delta N$ ,  $\Delta\rho$

**Output Information:**  
 $k_{eff}$ , Dose Rate,  
Fission Rate, etc...

**Output Uncertainty:**  
 $\Delta k_{eff}$ ,  $\Delta$  Dose Rate,  
 $\Delta$  Fission Rate

# Sensitivity Analysis and Uncertainty Propagation

Sensitivity coefficients provide insight on the sources and impact of uncertainty in nuclear engineering models.



# Introduction to Sensitivity Coefficients

- Sensitivity coefficients describe the fractional change in a response that is due to perturbations, or uncertainties, in system parameters.

$$S_{R,\Sigma_x} = \frac{\delta R/R}{\delta \Sigma_x/\Sigma_x}$$

- The SCALE TSUNAMI code calculates sensitivity coefficients for critical eigenvalue or reaction rate ratio responses:

$$R = k_{eff} \qquad R = \frac{\langle \Sigma_1 \phi \rangle}{\langle \Sigma_2 \phi \rangle}$$

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# Introduction to Sensitivity Coefficients

- Eigenvalue sensitivity coefficient calculations in TSUNAMI are calculated using the First-Order Perturbation Equation.

$$S_{k,\Sigma_x} = \Sigma_x \frac{\langle \Phi^* \left( \lambda \frac{\delta F}{\delta \Sigma_x} - \frac{\delta B}{\delta \Sigma_x} \right) \Phi \rangle}{\lambda \langle \Phi^* F \Phi \rangle}$$

# Introduction to Sensitivity Coefficients

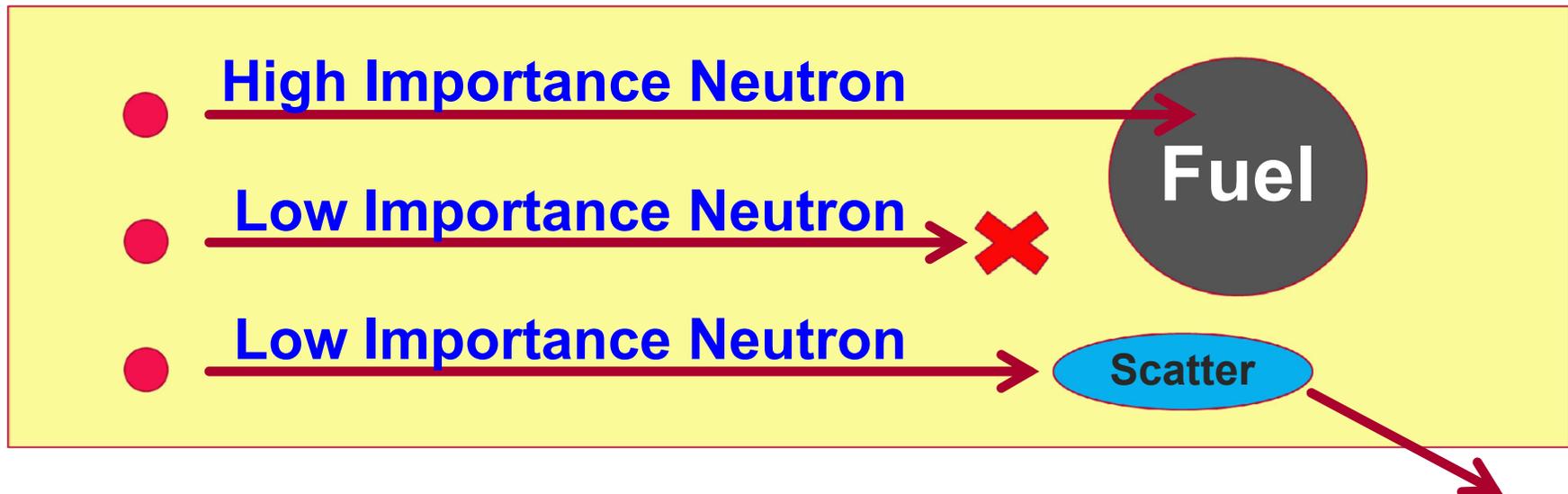
- For a sample capture reaction (*cap.*), the First-Order Perturbation Equation reduces to:

$$S_{k, \Sigma_{cap.}} = \frac{\langle \Phi^* \Sigma_{cap.} \Phi \rangle}{\frac{1}{k} \langle \Phi^* \Sigma_f \Phi \rangle}$$

- Tallying reaction rates is relatively straightforward for a Monte Carlo code.
- Tallying the adjoint flux ( $\Phi^*$ ) is more challenging.

# What is the Adjoint Flux?

- The adjoint flux, or “importance,” describes how much a neutron will contribute to a response.



# Introduction to Sensitivity Coefficients

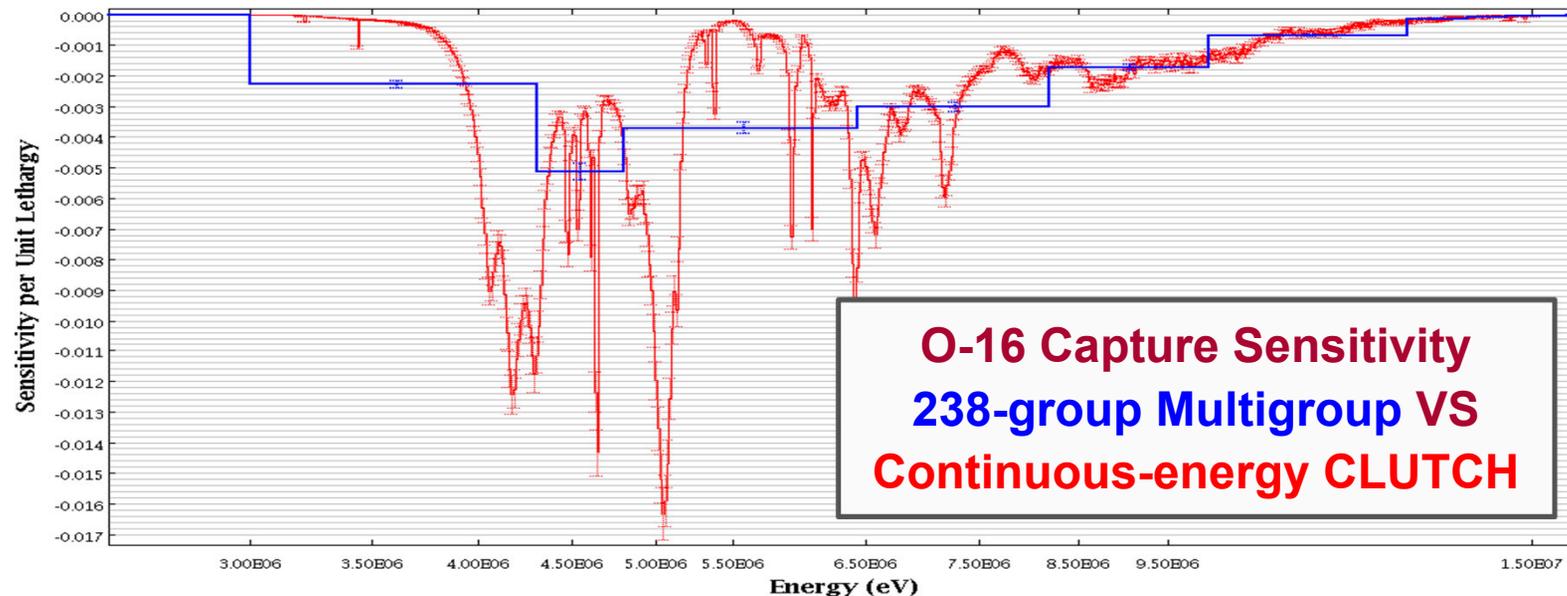
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# Continuous-Energy Resolution

- Continuous-energy capabilities allow for a better understanding what phenomena contribute a system's uncertainty.



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# Advanced Sensitivity Methods: Generalized Perturbation Theory

- Generalized Perturbation Theory (GPT) calculates sensitivity coefficients for responses that can be expressed as the ratio of reaction rates.

$$S_{R,\Sigma_x} = \frac{\delta R/R}{\delta \Sigma_x/\Sigma_x} \quad R = \frac{\langle \Sigma_1 \phi \rangle}{\langle \Sigma_2 \phi \rangle}$$

- Calculating generalized sensitivity coefficients requires solving an inhomogeneous, or generalized, adjoint equation:

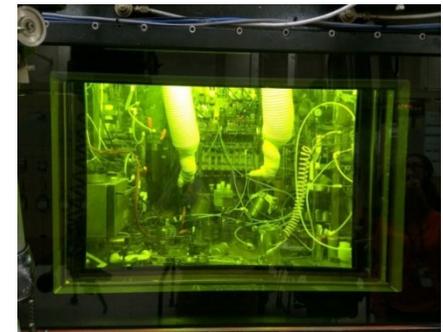
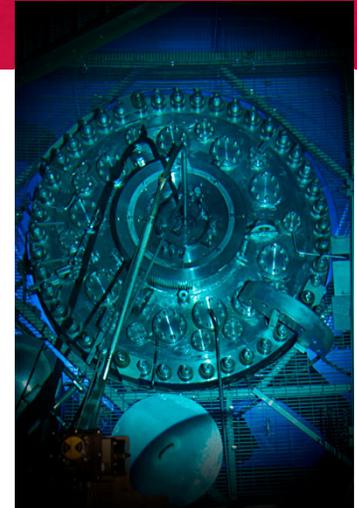
$$(L^* - \lambda P^*)\Gamma^* = S^*$$

- Applications for GPT sensitivity/uncertainty analysis include:
  - Relative Powers
  - Isotope Conversion Ratios
  - Multigroup Cross Sections



# Isotope Production Opportunities

- The High Flux Isotope Reactor (HFIR) at ORNL provides one of the highest intensity neutron fluxes.
- HFIR can provide unique isotopes, some of which have no alternative U.S. production source, including:
  - $^{14}\text{C}$  useful in medical applications such as studying diabetes, gout, anemia, and acromegaly
  - $^{63}\text{Ni}$  explosives detection, airport security
  - $^{229}\text{Th}$  provides  $^{225}\text{Ac}$  for  $\alpha$ -particle cancer therapy
  - $^{238}\text{Pu}$  radioisotope power for space exploration
  - $^{254}\text{Es}$  production of super-heavy elements
  - $^{252}\text{Cf}$  source of neutrons for nuclear reactor startup, study of materials with neutron diffraction, oil well logging, and neutron spectroscopy





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## Sensitivity Applications: Uncertainty Quantification

- Sensitivity coefficients can be combined with cross section uncertainties to quantify the uncertainty in a response.

$$S_{k,\Sigma_x} \cdot \text{Cov}_{\Sigma_x,\Sigma_y} \cdot S_{k,\Sigma_y}^T = \sigma_k^2$$

## The Sandwich Equation

# Sensitivity Applications: Uncertainty Quantification

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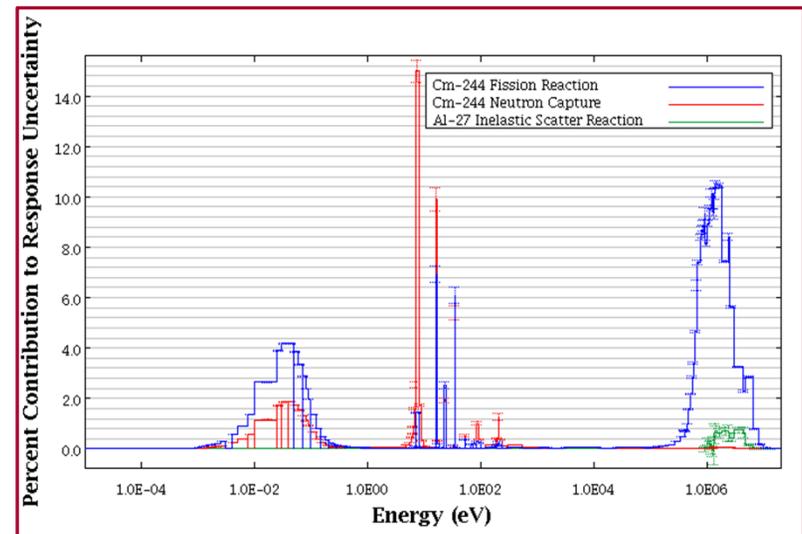
$$\begin{array}{ccccccc}
 S_{k,\Sigma_x} & \cdot & \text{Cov}_{\Sigma_x,\Sigma_y} & \cdot & S_{k,\Sigma_y}^T & = & \sigma_k^2 \\
 \uparrow & & \uparrow & & \uparrow & & \uparrow \\
 \left(\frac{\delta k/k}{\delta \Sigma/\Sigma}\right) & & \left(\frac{\Delta \Sigma}{\Sigma}\right)^2 & & \left(\frac{\delta k/k}{\delta \Sigma/\Sigma}\right) & & \left(\frac{\Delta k}{k}\right)^2
 \end{array}$$

## The Sandwich Equation

# Identifying Nuclear Data Needs

- Sensitivity-based uncertainty analyses offer insight on which reactions and neutron energies contribute the most uncertainty to responses of interest.

| Reaction Contributions to the Uncertainty in the $^{244}\text{Cm}$ Conversion Ratio |               |
|---|---------------|
| $^{244}\text{Cm}$ Fission Reaction  | 17.62%        |
| $^{244}\text{Cm}$ Neutron Capture   | 4.96%         |
| $^{27}\text{Al}$ Inelastic Scatter Reaction   | 0.72%         |
| $^{244}\text{Cm}$ Elastic Scatter Reaction  | 0.59%         |
| $^1\text{H}$ Elastic Scatter Reaction   | 0.56%         |
| <b>Total Data-Induced Uncertainty</b>   | <b>18.33%</b> |



Energy-dependent contributions to the uncertainty in the  $^{244}\text{Cm}$  conversion ratio.

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# Sensitivity Applications:

## Benchmark Similarity Assessment for Next-Generation Reactor Systems

- The similarity coefficient,  $c(k)$  or  $c_k$ , describes the amount of nuclear data-induced uncertainty that is shared by two systems.

$$\begin{array}{ccccccc} S_{R_1, \Sigma_x} & \cdot & \text{COV}_{\Sigma_x, \Sigma_y} & \cdot & S_{R_2, \Sigma_y}^T & = & \sigma_{R_1, R_2}^2 \\ \uparrow & & \uparrow & & \uparrow & & \uparrow \\ \left( \frac{\delta R/R}{\delta \Sigma/\Sigma} \right) & & (\Delta \Sigma/\Sigma)^2 & & \left( \frac{\delta R/R}{\delta \Sigma/\Sigma} \right) & & (\Delta R/R)^2 \end{array} \quad \rightarrow \quad c_k = \frac{\sigma_{R_1, R_2}^2}{\sigma_{R_1} \sigma_{R_2}}$$

# Sensitivity Applications:

## Benchmark Similarity Assessment for Next-Generation Reactor Systems

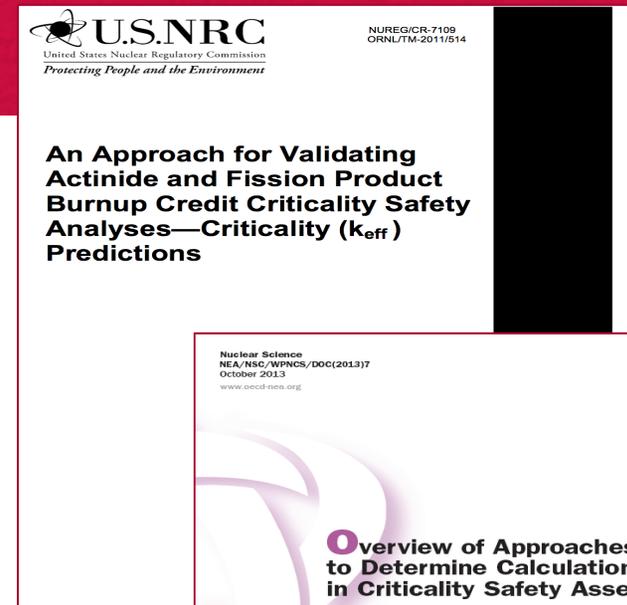
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- The TSUNAMI-IP code calculates  $c_k$  values between a target application and reference benchmark experiments.

# TSUNAMI in Practice

- ▶ U.S. Nuclear Regulatory Commission
  - ▶ Nuclear Materials Safety and Safeguards, Nuclear Reactor Regulation, Office of New Reactors
- ▶ U.S. DOE / Areva / Duke Energy
  - ▶ Mixed Oxide Fuel Fabrication Facility
- ▶ Candu Energy
  - ▶ ACR-1000 Design Validation
- ▶ Atomic Energy of Canada, Ltd.
  - ▶ ACR-700 NRC Review/PIRT
- ▶ U.S. DOE
  - ▶ Yucca Mountain post-closure criticality safety
- ▶ Global Nuclear Fuels
  - ▶ Transportation package licensing
- ▶ Svensk Kärnbränslehantering AB
  - ▶ Swedish used fuel repository
- ▶ Organization for Economic Cooperation and Development, Nuclear Energy Agency
  - ▶ International Expert Groups



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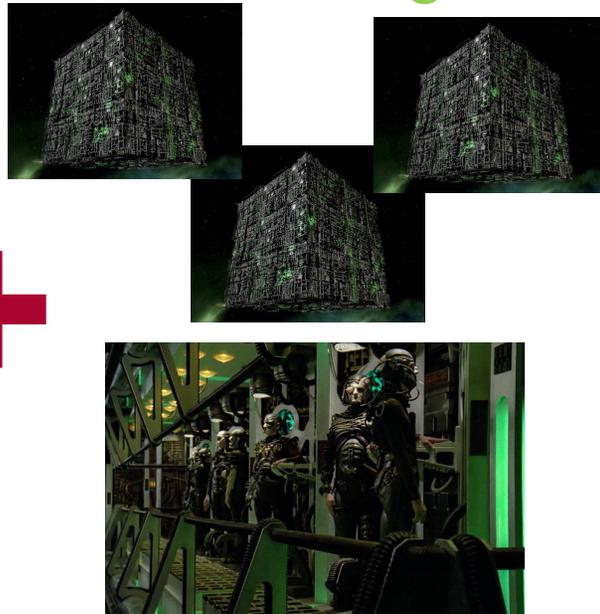
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# Data Assimilation

Data



The Borg



Data Assimilation



# Experimental Data Assimilation

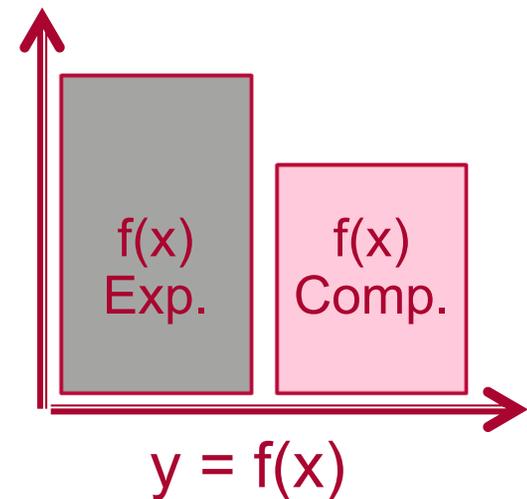
Monte Carlo radiation transport M&S tools are extremely high fidelity, relying mostly on first principle assumptions.

- **Premise:**

Disagreement between experimental results and high-fidelity M&S tools is caused primarily by errors in nuclear data.

- **Corollary:**

We can calibrate nuclear data evaluations by comparing experimental and computed results.



# Experimental Data Assimilation

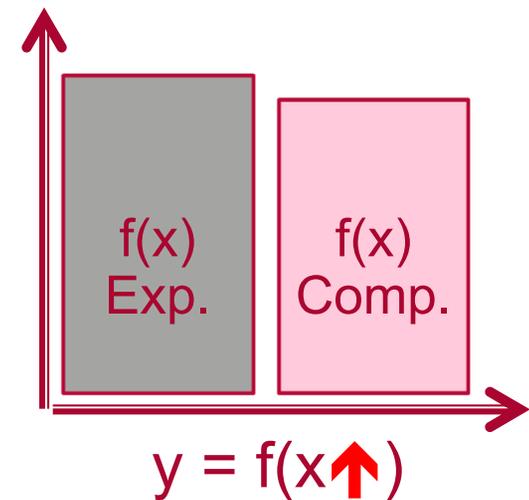
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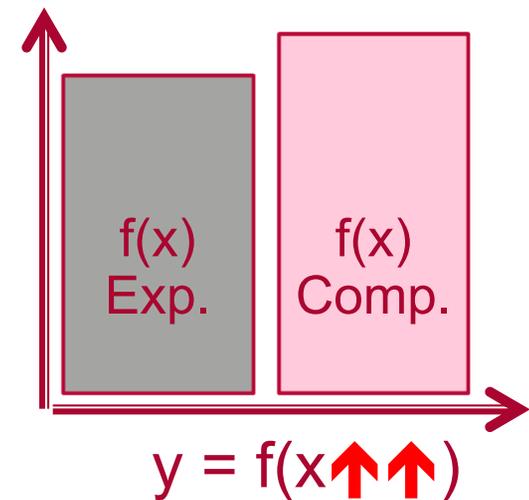
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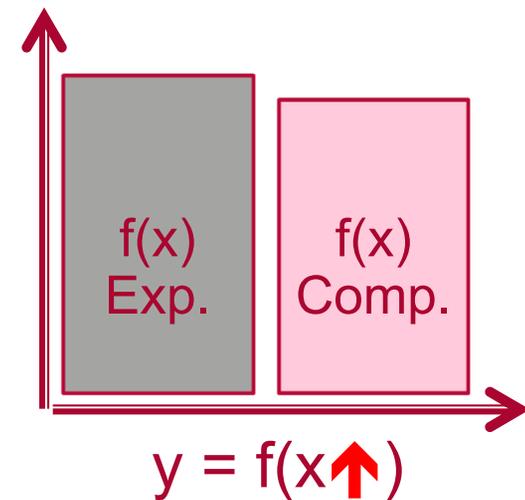
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# TSURFER Tools for Data Adjustment and Experimental Data Assimilation



TSURFER: Tool for S/U analysis of Response Functionals  
using Experimental Results

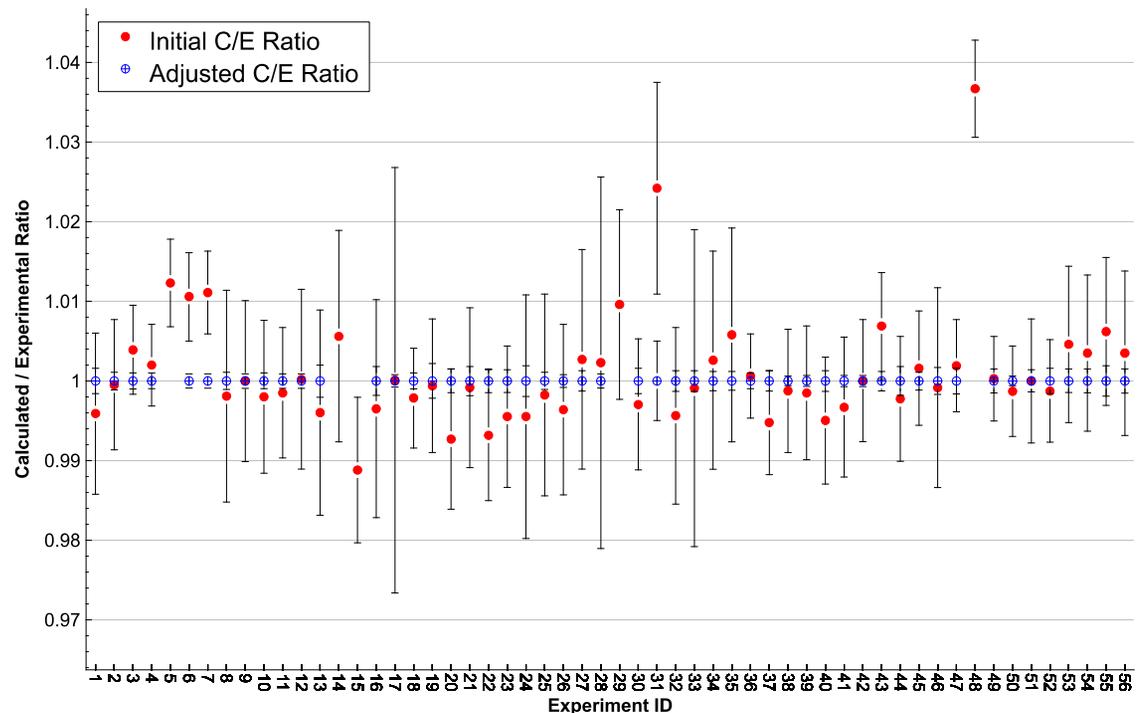
Uses sensitivity information to **consistently** adjust nuclear cross section data and reconcile biases between integral experiment results and computational predictions.

$$S_{f(x), x} = \frac{\partial f(x) / f(x)}{\partial x / x}$$

Modified cross section and cross section uncertainty data is used to anticipate computational biases in criticality safety applications.

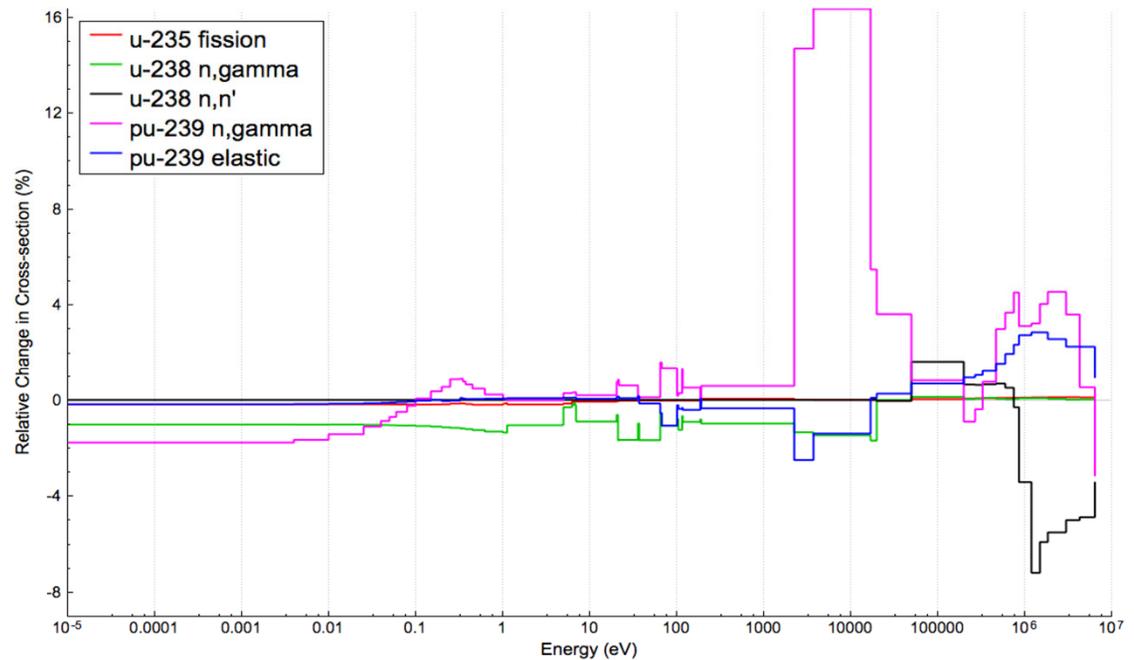
# Data Assimilation and Calibration

- Experimental benchmark data (E) is used to **improve the accuracy** of the **initial computed responses** (C).
- This assimilation consistently adjusts the underlying nuclear data.



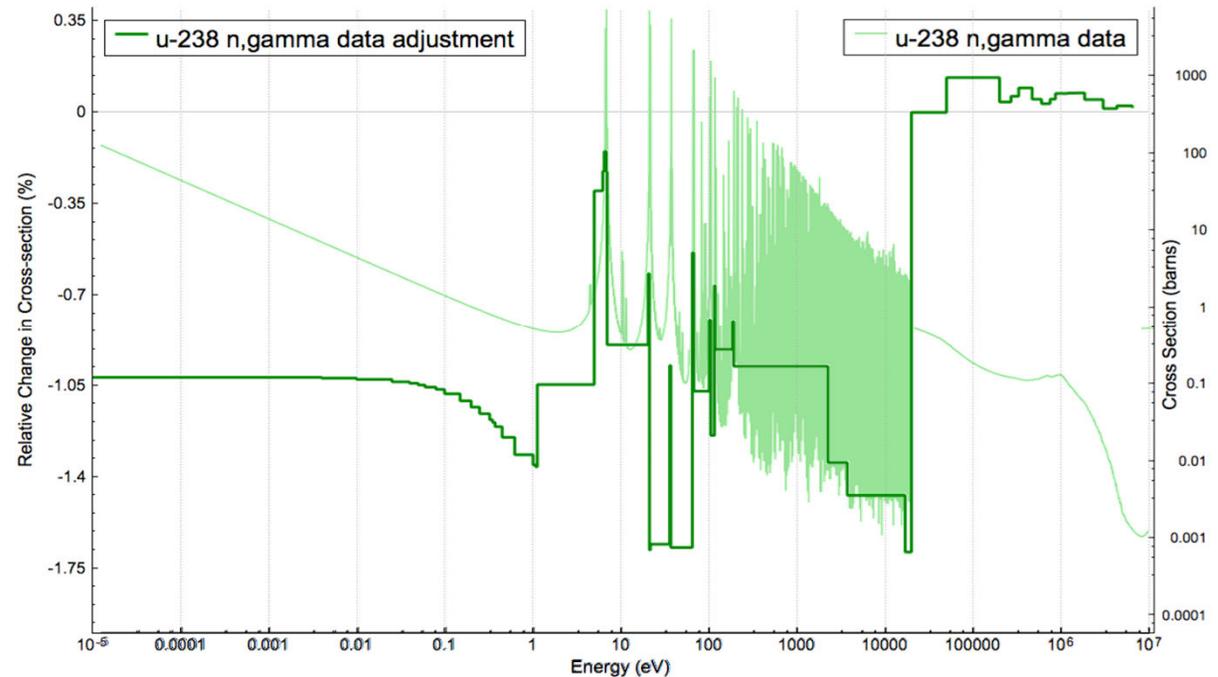
# TSURFER Cross Section Adjustments

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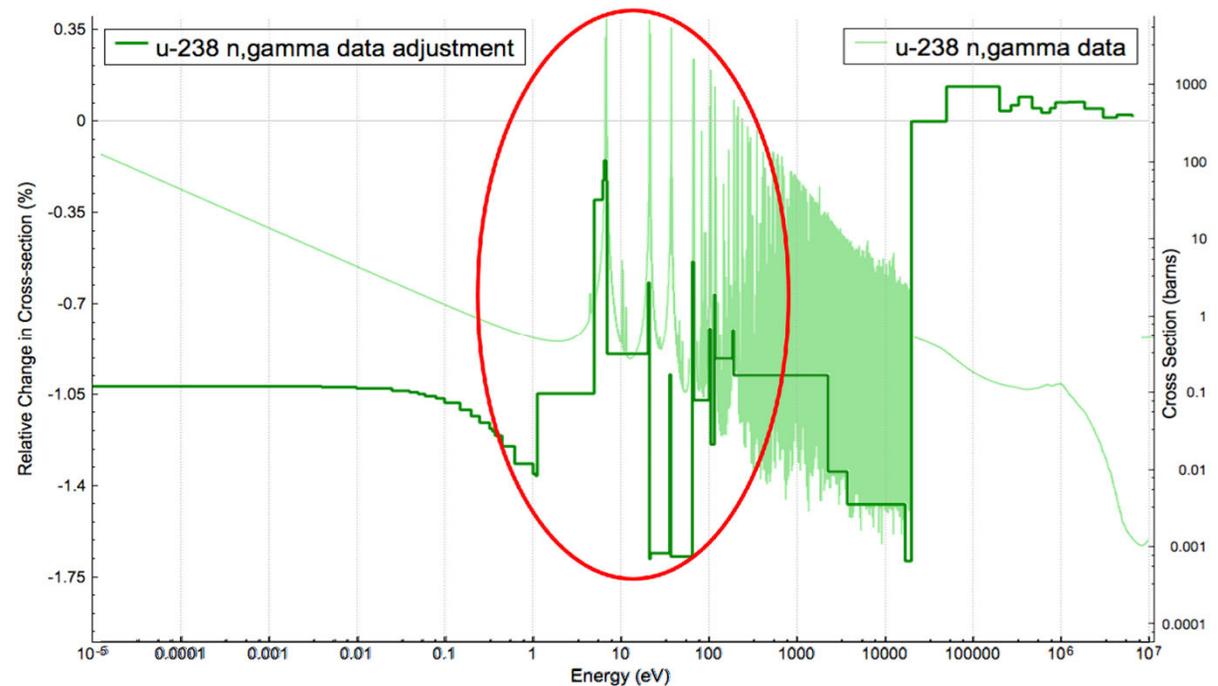
# Realistic Cross Section Adjustments?

- TSURFER adjusts multigroup (i.e. energy-averaged) cross sections.
- This approach cannot generate usable nuclear data evaluations.



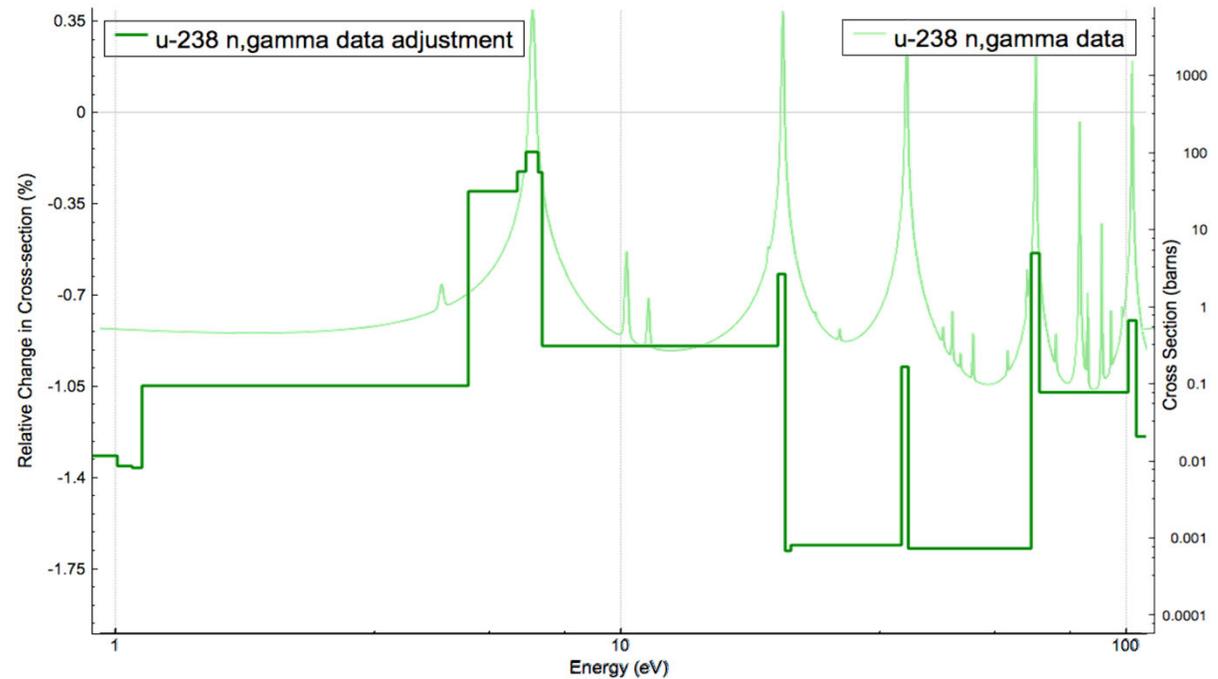
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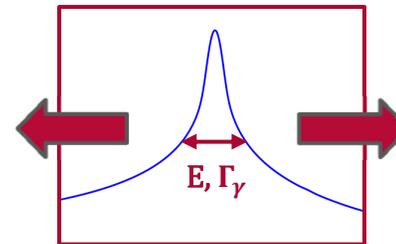
# NNSA Consortia on Monitoring, Technology, and Verification: “Improved Calibration of Evaluated Nuclear Resonance Parameters”

- **Task 1:** Develop sensitivity capabilities for evaluated nuclear data.
- **Task 2:** Modify TSURFER to assimilate experimental data by adjusting fundamental nuclear data.
- **Task 3:** Evaluate the accuracy of nuclear data and nuclear covariance adjustments.

**Task 1:**

$$S_{f(x), FWHM} = \frac{\partial f(x) / f(x)}{\partial FWHM / FWHM}$$

**Task 2:**



**Task 3:**



# LANL Monte Carlo Methods Subcontract:

## 1. Fission Multiplicity Models in Monte Carlo Simulations

- Goal is to use high fidelity fission multiplicity models in Monte Carlo critical and subcritical simulations.
  - Work is exploring the accuracy of MCNP simulations for subcritical ICSBEP benchmark responses using higher fidelity fission physics models.
  - Work will explore using fission matrices to accelerate near-critical simulations.

# LANL Monte Carlo Methods Subcontract:

## 2. S/U-based Validation for Nuclear Criticality

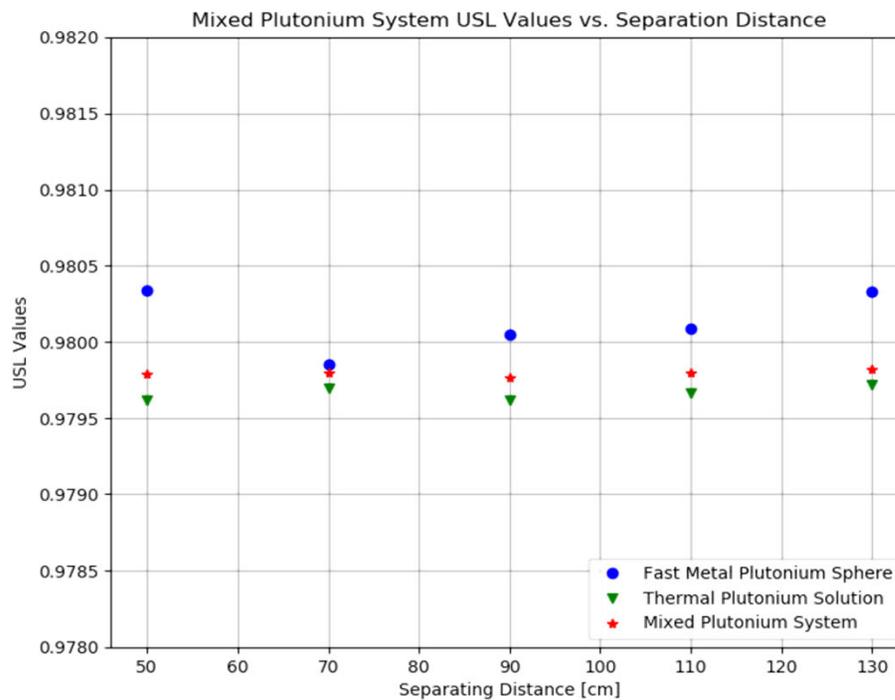
Safety

- Goal is to better understand the use of S/U-based validation in nuclear criticality safety applications.
  - Upper Subcriticality Limit (USL) calculations vary significantly when using sensitivity coefficients for different sub-sections of a system.
  - It is not clear which USL is the most appropriate.

# LANL Monte Carlo Methods Subcontract:

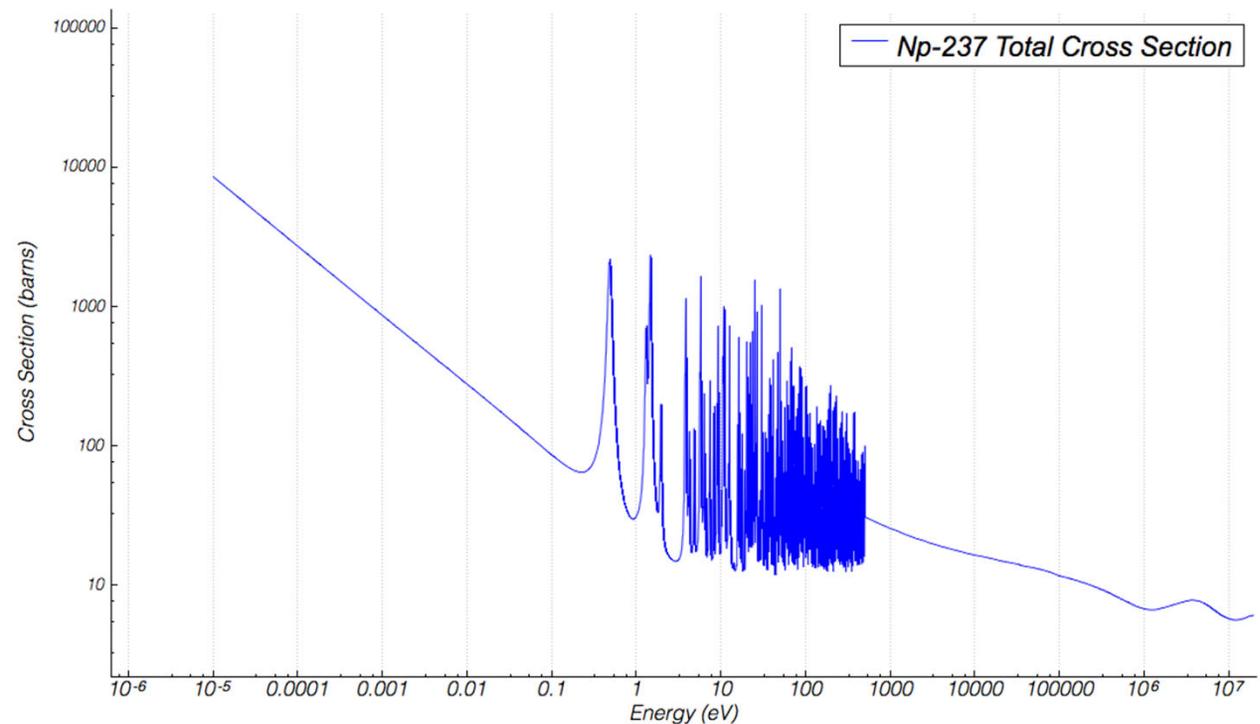
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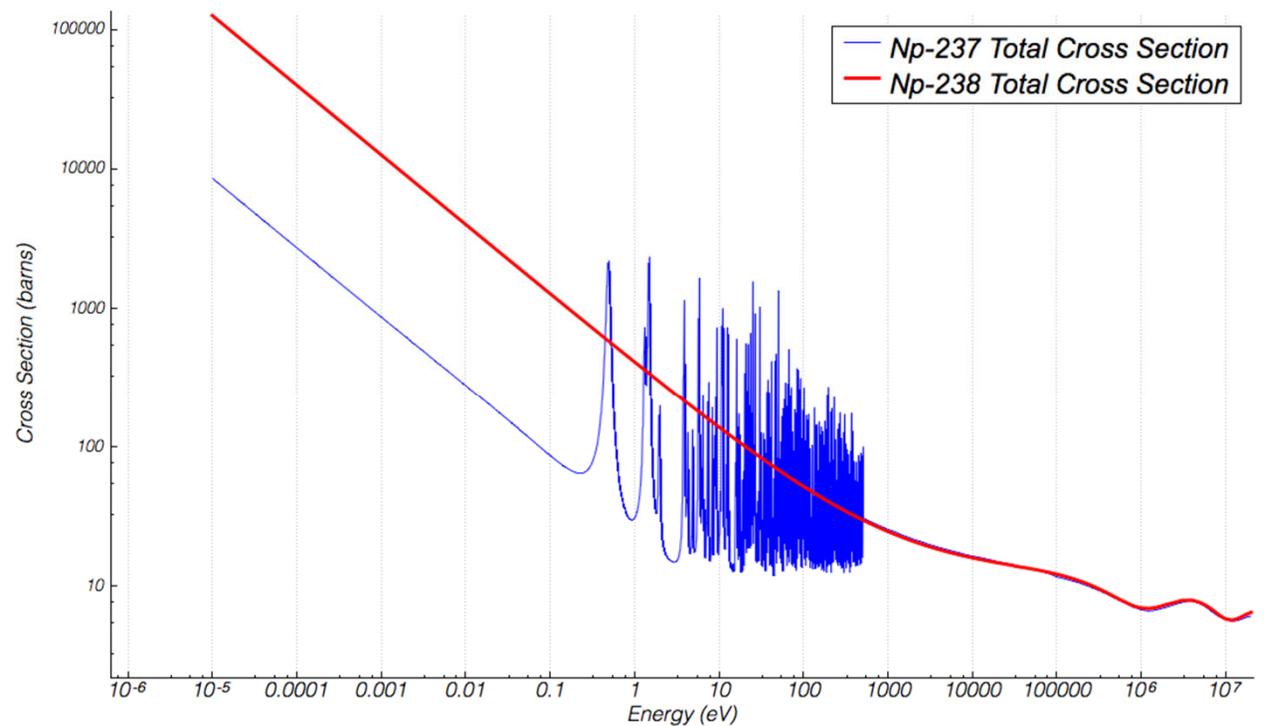
# NEUP Grant: Inferring Cross Sections for Short-lived Radioisotopes

- $^{237}\text{Np}$  is commonly used as the seed material to produce  $^{238}\text{Pu}$ .
- Low quality nuclear data for isotopes in the  $^{238}\text{Pu}$  activation chain introduces computational bias in models for  $^{238}\text{Pu}$  production.



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## NEUP Grant: Inferring Cross Sections for Short-lived Radioisotopes

- Extensive experimental data exists for transcurium irradiation experiments, but current methods cannot assimilate this data.
- Successful assimilation could help infer cross sections for rare or short-lived radioisotopes.

$$S = \frac{d(\text{Isotope Number Density})/\text{Isotope Num. Den.}}{d\Sigma_x/\Sigma_x}$$

- A tool performing **sensitivity analysis for the number density of isotopes in depletion and transmutation calculations** is needed for this cross section data calibration.

# Questions?

Please contact:

Chris Perfetti

[cperfetti@unm.edu](mailto:cperfetti@unm.edu)

*“All models are wrong, but some are useful.”*

– Professor George E. P. Box