Title: MCNP Continuous-Energy Sensitivity and Uncertainty Progress and Application

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MCNP Continuous-Energy Sensitivity and Uncertainty Progress and Application

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Abstract

The DOE/NNSA Nuclear Criticality Safety Program (NCSP) has funded the development of a sensitivity/uncertainty capability for criticality safety methods development at LANL. Updates are given on the continuous-energy sensitivity capability in MCNP and other related efforts. Efforts on the development of the capability for processing of covariance data by NJOY are summarized. The application of these techniques and the development of new software infrastructure to support validation for Pu operations at LANL is discussed.
Introduction

- Current Status & Recent Developments
- Application at LANL
- Future Prospects
Motivation

- Sensitivity/uncertainty analysis allows us to quantify how well (or poorly) software predicts criticality.
Continuous-energy $k$ sensitivity coefficient capability is available today in MCNP6.1 (released summer 2013).

- Uses adjoint weighted methodology used in TSUNAMI with Iterated Fission Probability method.
  - Benchmarked with analytic solutions, direct perturbations, and comparisons with TSUNAMI.
- Robust method with minimal user involvement.
  - Define the isotopes, reactions, and energy grid, run, and get results.
Next MCNP6 beta release includes minor enhancements to the sensitivity capability:

- Legendre coefficient sensitivities for scattering distributions.
- Output format in TSUNAMI-B SDF format for interoperability with SCALE tools.
- Support for additional reaction MTs and minor bug fixes.

MCNP6.1.1 to be released in the next few months.
Often the scattering distributions and uncertainties are given as Legendre moments.

Can express renormalized sensitivity coefficient $\hat{S}_{k,f}(\mu)$ as Legendre moment sensitivity $\hat{S}_{k,f,\ell}$.

Given a defined cosine grid with $N$ bins with index $i$, the $\ell$th Legendre moment sensitivity is

$$\hat{S}_{k,f,\ell} = \frac{2\ell + 1}{2} f_\ell \sum_{i=0}^{N-1} (\mu_{i+1} - \mu_i) \frac{P_\ell(\mu_{i+1/2})}{F_{i+1/2}} \hat{S}_{k,f,i+1/2}$$

Presented at NCSD Topical (Sep. 2013) and ANS Winter Meeting (Nov. 2013).
The linearly anisotropic ($P_1$) component of elastic scattering may have a significant effect on $k$.

Higher orders of scattering are typically not important, and neither is anisotropy of inelastic scattering.

For fast systems with significant leakage, core and reflector materials are often significant and should be included in nuclear data adjustments.

For thermal systems, scattering distributions matter less.
Jezebel $^{239}\text{Pu}$ Elastic Moment Sensitivity

![Graph showing the sensitivity of k-eff to incident energy for different Pu-239 Pn Elastic cases.](image)

Operated by Los Alamos National Security, LLC for the U.S. Department of Energy’s NNSA
Flattop (HEU) $^{238}$U Elastic Moment Sensitivity

![Graph showing keff sensitivity vs. incident energy for different U-238 elastic channels.](Image)

- U-238 P1 Elastic
- U-238 P2 Elastic
- U-238 P3 Elastic
- U-238 P4 Elastic
- U-238 P5 Elastic
Fixed-Source Sensitivity Prototype

- Apply adjoint weighting calculations to fixed-source calculations for subcritical measurements.

\[ S_{R,x} = \frac{\langle f_{R,x}, \psi \rangle + \langle \psi^\dagger, Q_x \rangle - \langle \psi^\dagger, H_x \psi \rangle}{R}. \]

- Verification performed with direct perturbations; shows good agreement.
- Analysis of subcritical Thor core measurements.
- Presented at ANS Winter Meeting (Nov. 2013).
- Future extension to arbitrary responses in eigenvalue calculations.
Application to Experimental Measurements

- Thor core sensitivity study:
Application to Experimental Measurements

- Thor core pieces:
### Sensitivity to Core Mass

- Predict glory hole mass impact on SNAP response.

<table>
<thead>
<tr>
<th></th>
<th>Description</th>
<th>Pu Mass</th>
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</thead>
<tbody>
<tr>
<td>A</td>
<td>Center Only + 206.9 g Pu</td>
<td>206.9 g</td>
</tr>
<tr>
<td>B</td>
<td>Bottom &amp; Center + 206.9 g Pu</td>
<td>206.9 g</td>
</tr>
<tr>
<td>C</td>
<td>Top &amp; Center + 206.9 g Pu</td>
<td>206.9 g</td>
</tr>
<tr>
<td>D</td>
<td>Fully Assembled + 206.9 g Pu</td>
<td>206.9 g</td>
</tr>
<tr>
<td>E</td>
<td>Fully Assembled + 109.5 g Pu</td>
<td>109.5 g</td>
</tr>
<tr>
<td>F</td>
<td>Fully Assembled + 49.0 g Pu</td>
<td>49.0 g</td>
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</table>
Sensitivity to Core Mass

- Reference results are for full removal of glory hole mass.

<table>
<thead>
<tr>
<th>Pert.</th>
<th>$S_{R,x}$</th>
<th>$\Delta R_Q$</th>
<th>$\Delta R_H$</th>
<th>$C/E$</th>
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<td>0.942</td>
</tr>
</tbody>
</table>

- $S_{R,Q}$ assumed to be unity.
Doppler Reactivity Coefficients (Gonzales, UNM)

- Use sensitivity methodologies for Doppler reactivity coefficients.
- Use temperature series expansion to compute cross-section derivatives.
- Effective cross section shows reasonable agreement with direct perturbations.
- Current work is on scattering kernel derivative.
Uncertainty/Covariance Project

- Develop a capability in the MCNP framework that automates uncertainty quantification of nuclear data.
  - Beginning of calculation, query covariance data and automatically create sensitivity profiles.
  - Run criticality calculation normally and compute $k$ and $S_{k,x}$ for all data.
  - At end of calculation, read covariance data and compute uncertainty in $k$ using “sandwich rule”:
    \[(\delta k)^2 = \textbf{SCS}^T.\]

- First, however, covariance data is needed!
Uncertainty/Covariance Project

- Studied two formats:
  - Compressed upper triangular or full matrix.
  - Principal eigenvectors.
- Principal eigenvector method requires less storage, but savings may not be large enough in many cases to justify the cost.
- Issues identified in NJOY for processing covariance data. Being worked on.
Uncertainty/Covariance Project

- Prototype version of MCNP6 developed that automatically computes uncertainty in $k$.
- Uses covariance data generated from NJOY that is processed externally.
- Preliminary results presented at NCSD Topical (Sep. 2013).
- Identifies what I believe are inconsistencies in ENDF/B-VII.1 covariance data.
Preliminary Uncertainty Results

- Covariance data generated with ENDF/B-VII.1 for $^1$H, $^{16}$O, $^{235}$U, $^{238}$U, and $^{239}$Pu.
- No thermal scattering law covariances included.
- Benchmarks:
  - Bare-HEU Sphere (Lady Godiva)
  - Reflected-HEU Sphere (Flattop)
  - Uranium-Hydride Experiment
  - Light-Water Moderated LEU Lattice
  - Bare-Pu Sphere (Jezebel)
  - Reflected-Pu Sphere (Flattop)
  - Light-Water Moderated MOX Lattice
  - Pu Solution (Light-Water)
Preliminary Uncertainty Results

- MCNP calculated $k$ with nuclear data uncertainty:
LANL Pu operations currently halted.

One issue identified by reviews is the validation for MCNP criticality safety calculations.
  - My opinion: Previous effort was a very good start, but not sufficient.

Goal is to develop a robust computational tool set to assist with validation that fits within their MCNP-centric workflow.

Based on the ORNL sensitivity/uncertainty methodologies funded by NCSP.
Sensitivity/Uncertainty Methodology

- Use sensitivity coefficients and covariance data as similarity parameter $c_k$ to identify benchmarks relevant to a set of computational models.
- Uses MCNP6.1 sensitivity coefficient capability, ENDF/VII.1 nuclear data, and ORNL 44-group covariance libraries that have been processed into LANL format.
- Search benchmark suite to develop weighting factors for calculational margin computation.
  - NCS validation suite has been expanded to over 1000 benchmarks covering a broad set of areas.
- Hope is to distribute these capabilities outside LANL and possibly as a library for MCNP that integrates validation as part of the $k$ calculation.
Expanded Validation Suite

- Current NCS benchmark set (1095 cases) from ICSBEP:

<table>
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<th></th>
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<th>Inter</th>
<th>Therm</th>
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<td>1</td>
<td>0</td>
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<tr>
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<td>0</td>
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<td>0</td>
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<tr>
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<tr>
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<td>10</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>10</td>
</tr>
</tbody>
</table>
Selection of Benchmarks

- Compute the $c_k$ of process and benchmarks.
- Find maximum $c_k$, use to determine total “weight” required for validation:
  \[ w_{req} = A + B(1 - c_{k,max}). \]

- $A = 25$, $B = 75$. Need more benchmarks for lower $c_{k,max}$.
- Reduce acceptance $c_{k,acc}$ until total weight of included benchmarks reaches $w_{req}$.
- Weight for a benchmark is
  \[ w = \frac{c_k - c_{k,acc}}{c_{k,max} - c_{k,acc}}. \]
Calculational Margin Methodology

- Uses extreme value theory to find probability that worst case bias in $k$ is $< 0.99$.
- Including more benchmarks only increases the calculational margin; fail safe.
- Cumulative density function for bias in $k$ for a benchmark $j$:

$$F_j(x) = (1 - w_j) + \frac{w_j}{2} \left[ 1 + \text{erf}\left(\frac{x - \beta_j}{\sqrt{2\sigma_j^2}}\right)\right].$$

- Cumulative density function for worst case bias:

$$F(x) = \prod_{j=1}^{N} F_j(x).$$
Pu Metal-Water Mix Results

Pu Concentration (g/L) vs. Calculational Margin

- Pu-239 Metal-Water Mix

Graph showing the relationship between Pu concentration (g/L) and calculational margin.
Margin of Subcriticality

- Make recommendation for starting point for NCS analyst.
- Margin for unknown and undetected software errors in MCNP6.1: 0.005.
- Margin for variability in cross section data:
  - Currently substitute in JEFF-3.1 nuclear data.
  - Planned approach is to use residual cross-section uncertainty from GLLS methodology in TSURFER from ORNL.
Future Validation Capability Development Work

- Apply validation tools to Pu systems (metals, oxides, solutions) as demonstration.
- Train NCS division on use of new validation tools for U systems for NCERC.
- Investigate integration with MCNP and release of 1095 problem validation suite.
Summary

- MCNP currently supports generation of continuous-energy sensitivity coefficients.
- New developments in Legendre moment sensitivities (to be released), fixed-source sensitivities (prototype), and Doppler coefficients (being researched).
- Future thrusts in sensitivity to generalized responses in eigenvalue problems and $\alpha$ eigenvalue problem for prompt period measurements (e.g., Godiva-IV).
- Development of uncertainty quantification capabilities in or with MCNP proceeding.
- Immediate LANL needs in Pu operations has spurred the development of external capabilities for validation using S/U methodologies funded by NCSP and developed by ORNL.
Acknowledgments

- NCS Validation Team: Jeremy Conlin, Jeff Favorite, Skip Kahler, Alyssa Kersting, Kent Parsons, Jessie Walker.
- Thank you for the funding from the DOE/NNSA Nuclear Criticality Safety Program.
Questions?