

Uncertainty Propagation

with Fast Monte Carlo Techniques

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(Monte Carlo vs. Deterministic S/U...)





For each random ENDF file, the benchmark calculation is performed with MCNP. At the end of the *n* calculations, *n* different k_{eff} values are obtained.

 $\sigma_{\text{total}}^2 = \sigma_{\text{statistics}}^2 + \sigma_{\text{nuclear data}}^2$



Advantages of the TMC method

∎ many presentations at ND,

- IS computer time (not human time),
- Successfully applied (criticality, shielding, reactor, burn-up...)
- Most simple path (no additional processing, no covariance required),
- Other variants: AREVA (NUDUNA), GRS (XSUSA), CIEMAT (ACAB), PSI (NUSS), CNRS Grenoble..., based on covariance files,
- Many spin-offs (TENDL covariances, sensitivity, adjustment...)
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But can TMC overtake (at least compete with) S/U methods ?





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There is a solution with Monte Carlo codes: (in fact 2 solutions)

✗ fast GRS method,✗ and fast TMC.



2012: fast GRS method

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First presented in PHYSOR-2012 by W. Zwermann *et al*.. It takes advantage of conditional expectations:

If two output variables $k^{(1)}$ and $k^{(2)}$ are identically distributed and conditionally independent given the vector of nuclear data input then

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In practice:

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- 1. perform i = 1..500 MCNP short calculations with random nuclear data and a fixed seed $s_1 \Longrightarrow k_{eff}^{(1)}(i)$
- 2. repeat for j = 1..500, same random nuclear data but fixed seed $s_2 \implies k_{eff}^{(2)}(j)$

There is no necessity to have small $\sigma_{statistics}$!! each run can be (very) short

fast GRS method

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 2×500 "short" runs ~ $2 \times$ "long" run in time

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If a single calculation takes *m* histories (σ_{stat} small enough), then repeat it *n* times with *m/n* histories, random nuclear data and random seeds.

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- © Example: the Martin-Hoogenboom benchmark

MCNP model: 241 fuel assemblies, with 264 fuel pins each



 \implies 357 × 357 × 100 regions (1.26 × 1.26 × 3.66 cm³): 12.7 million cells

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MCNP model: 241 fuel assemblies, with 264 fuel pins each



 $\implies 357 \times 357 \times 100 \text{ regions } (1.26 \times 1.26 \times 3.66 \text{ cm}^3): 12.7 \text{ million cells}$ Uncertainty on generated local pin power (tally f7) due to ²³⁵U, ²³⁸U, ²³⁹Pu and H in H₂O thermal scattering in each cell ?

Fast TMC method

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1 normal calculation without nuclear data uncertainty takes $n = 2 \times 10^{11}$ histories ($\sigma_{\text{statistics}} = 0.25$ % at the center, 500 weeks on 1 cpu)

 \implies TMC: 500 random runs of $n = 2 \times 10^{11}$ histories (500 weeks for each)

 \implies fast TMC: 500 random runs of $n/500 = 4 \times 10^8$ histories (1 week for each)

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fast TMC and GRS methods: If we can do a calculation once, we can also get nuclear data uncertainties in twice the time (or less).

