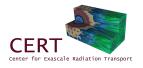
#### A High-Order Low-Order Algorithm with Exponentially-Convergent Monte Carlo for Thermal Radiative Transfer

Simon Bolding, Matt Cleveland, and Jim Morel

4 August 2016







We are interested in modeling thermal radiation transport in the high energy density physics regime

Modeling materials under extreme conditions Temperatures  $\mathcal{O}(10^6)~\mathrm{K}$  or more

Photon radiation transports through a material Significant energy may be exchanged

We want to improve efficiency of Monte Carlo calculations e.g., inertial confinement fusion, supernovae, et. al.

## Our method has been applied to a simplified model: the 1D frequency-integrated radiative transfer equations

#### Energy balance equations for radiation and material.

Radiation intensity  $I(x, \mu, t)$ , material temperature T(x, t)

$$\frac{1}{c}\frac{\partial I}{\partial t} + \mu \frac{\partial I}{\partial x} + \sigma_t I(x, \mu, t) = \frac{1}{4\pi} \sigma_a a c T^4,$$

$$C_v \frac{\partial T(x, t)}{\partial t} = \sigma_a \phi(x, t) - \sigma_a a c T^4$$

Equations are nonlinear and may be tightly coupled Absorption opacity  $(\sigma_a)$  can be a strong function of T

## Implicit Monte Carlo (IMC) is the standard Monte Carlo transport method for TRT problems

The system is *linearized* over a time step  $t \in [t^n, t^{n+1}]$  Opacities are evaluated with  $T(t^n)$ 

- Produces a linear transport equation with effective emission and scattering terms
- MC particle histories are simulated tallying radiation energy deposition
- Emission source is not fully time-implicit.
   Uses MC integration over Δt for intensity

## We developed a high-order low-order (HOLO) method that improves on several drawbacks of IMC

Standard IMC	HOLO Method
Large statistical noise possible	ECMC is very statistically efficient for TRT problems
Effective scattering can make MC very expensive	MC solution has no scattering
Linearization can cause non-physical results (maximum principle violations)	Fully implicit time-discretization and LO solution resolves nonlinearities
Reconstruction of linear emission shape limits artificial energy propagation	Linear-discontinuous FE for $T(x)$ preserving equilibrium diffusion limit



Overview of algorithm

Derivation of the LO equations

Exponentially Convergent MC High-Order Solver

Computational Results

Extensions and Ongoing Research



#### Overview of algorithm

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## Basic idea is a nonlinear low-order system with high-order angular correction from Monte Carlo transport solves

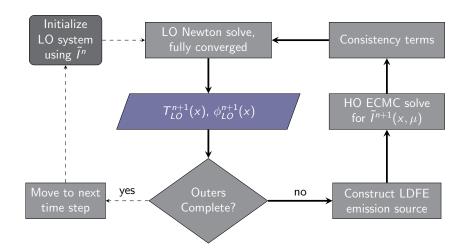
#### The **LO** system is space-angle moment equations, on a fixed finite-element (FE) spatial mesh

- Reduced dimensionality in angle allows for solution with Newton's method
- ▶ Output: linear-discontinuous  $\phi(x)$  and T(x)Construct LDFE scattering and emission source

#### The **HO** system is a pure-absorber transport problem

- Solved with exponentially-convergent MC (ECMC) for efficient reduction of statistical noise
- Output: consistency terms

## Iterations between the HO and LO systems can be performed each time step





Overview of algorithm

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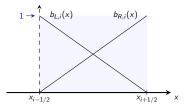
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## The LO equations are formed as *consistently* as possible with spatial and angular moments of TRT equations

The time discretization is backward Euler for both the HO and LO equations

Spatial moments are weighted with FE basis functions:



$$\langle \cdot \rangle_{L,i} = \frac{2}{h_i} \int_{x_{i-1/2}}^{x_{i+1/2}} b_{L,i}(x)(\cdot) dx$$

Half-range integrals reduce angular dimensionality

$$I^+(x) = 2\pi \int_0^1 I(x,\mu) \mathrm{d}\mu$$

### Apply moments to the TRT equations and manipulate to form angular consistency terms

For example, apply  $\langle \cdot \rangle_{L,i}^+$  to a streaming term:

$$\begin{split} \left\langle \mu \frac{\partial I}{\partial x} \right\rangle_{L}^{+} &= -\frac{2}{h_{i}} \{ \mu I \}_{i-1/2}^{+} + \frac{1}{h_{i}} \left[ \langle \mu I \rangle_{L,i}^{+} + \langle \mu I \rangle_{R,i}^{+} \right] \\ &= -\frac{2}{h_{i}} \frac{\{ \mu I \}_{i-1/2}^{+}}{I_{i-1/2}^{+}} I_{i-1/2}^{+} + \frac{1}{h_{i}} \frac{\langle \mu I \rangle_{L,i}^{+}}{\langle I \rangle_{L,i}^{+}} \langle I \rangle_{L,i}^{+} + \frac{1}{h_{i}} \frac{\langle \mu I \rangle_{R,i}^{+}}{\langle I \rangle_{R,i}^{+}} \langle I \rangle_{R,i}^{+} \end{split}$$

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#### Now, close the final moment equations:

- ► Approximate consistency terms with  $\tilde{l}_{HO}^{n+1}$  from previous HO solve
- ▶ Eliminate remaining face unknowns with LD spatial closure  $T^4(x)$  and T(x) also assumed LD



Overview of algorithm

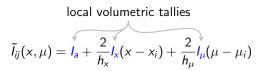
Derivation of the LO equations

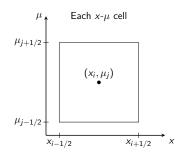
Exponentially Convergent MC High-Order Solver

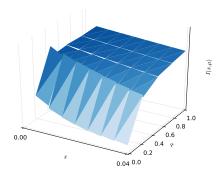
Computational Results

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# We use a projection $\tilde{I}(x,\mu)$ onto a space-angle LDFE mesh to represent the solution







We apply the ECMC algorithm to the pure-absorber HO transport equation

$$\left[\mu \frac{\partial}{\partial x} + \left(\sigma_t + \frac{1}{c\Delta t}\right)\right] I^{n+1} = \frac{1}{4\pi} \left[\sigma_a a c \left(T_{LO}^{n+1}\right)^4 + \sigma_s \phi_{LO}^{n+1}\right] + \frac{\tilde{I}^n}{c\Delta t}$$

$$\mathbf{L} I^{n+1} = q$$

For each batch m:

- ► Evaluate residual source:  $r^{(m)} = q \mathbf{L}\tilde{I}^{n+1,(m)}$
- ► Estimate  $\epsilon^{(m)} = \mathbf{L}^{-1} r^{(m)}$  via MC simulation
- Update solution:  $\tilde{I}^{n+1,(m+1)} = \tilde{I}^{n+1,(m)} + \tilde{\epsilon}^{(m)}$

## Our HO system allows for straight-forward variance reduction and source biasing

 $I^n(x,\mu)$  is often an **excellent** estimate of  $I^{n+1}(x,\mu)$ No MC sampling from thermal equilibrium regions

Histories stream without collision along path s, weight reduces as  $w(s) = w_0 e^{-\sigma_t s}$ 

Use cell-wise systematic sampling for  $|r^{(m)}|$  source Particularly effective in thick cells

- ▶ Particles in each x- $\mu$  cell  $\propto |r^{(m)}|$  in cell
- Set minimum n for cells except for cells in thermal equilibrium



Overview of algorithm

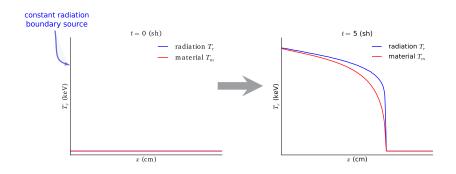
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## We will test our method with several standard **Marshak Wave** problems



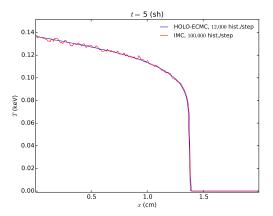
Figures depict radiation temperature  $T_r = \sqrt[4]{\phi/ac}$ 

#### Implementation specifics for results are given below:

- ► HOLO method implemented as stand-alone C++ code IMC results from Jayenne (LANL code)
- $\blacktriangleright$   $\Delta t$  increases from 0.01 ns to 0.1 ns
- ▶ One HO solve per time step, with two LO solves
  - each HO solve has 3 ECMC batches no adaptive mesh refinement

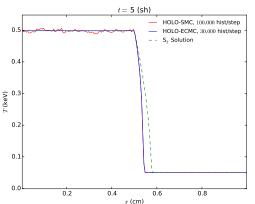
The HOLO method produces significantly less noise than IMC for a typical Marshak Wave: **FOM=145** 

- $ightharpoonup \sigma_a \propto T^{-3}$
- ► Transient solution after 5 shakes ( $\sim$  520 steps) 200  $\times$  cells (and 4  $\mu$  cells for ECMC)



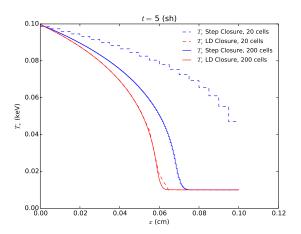
#### ECMC is more efficient than standard MC as a HO solver

- ▶ Left half is optically thin ( $\sigma$ =0.2 cm<sup>-1</sup>), right half is thick ( $\sigma_a$ =2000 cm<sup>-1</sup>). 8  $\mu$  cells
- Results for HOLO with different HO solvers: ECMC (FOM=10,000), standard MC (FOM=0.46), and an S<sub>2</sub> solution



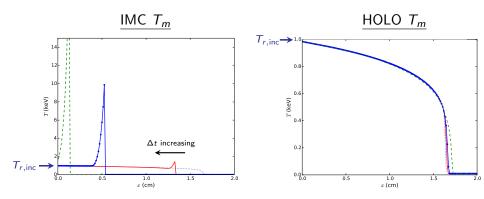
## The LDFE discretization for the LO equations preserves the equilibrium diffusion limit

▶ Large, constant  $\sigma_a$  and small  $c_v$ 



# Our HOLO method preserves the maximum principle with sufficient nonlinear convergence

- ▶ Material temperatures plotted; all simulations end at t = 0.1 sh  $\sigma_a \propto T^{-3}$ ,  $c_v$  small,  $\Delta t \in [10^{-4}, 10^{-2}]$  sh
- ► LO Newton iterations required damping





Overview of algorithm

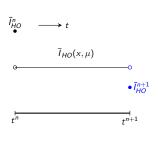
Derivation of the LO equations

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### The time variable can be included in the ECMC trial space with a consistent LO time closure



Include continuous  $\frac{1}{c}\frac{\partial}{\partial t}(\cdot)$  in **L** for residual source leaving T(x) still implicit

Sample and track particle histories in time.

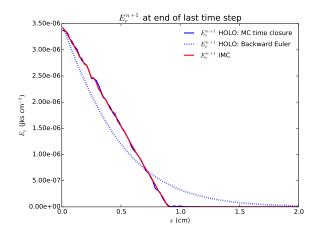
Tally the time-averaged and  $t^{n+1}$  error

In LO equations, parameterize  $\phi_{LO}^{n+1}$  in terms of **time-averaged** unknowns, e.g.,

$$\langle \phi \rangle_{L,i}^{n+1} = 2 \; \gamma_{L,i}^{HO} \; \overline{\langle \phi \rangle}_{L,i} - \langle \phi \rangle_{L,i}^{n}$$

## The time-closure parameters preserve accuracy of MC time integration in the LO solution

- Material has  $\sigma_a=10^{-6}~{\rm cm}^{-1}$ , so temperature uncouples take 3 large time steps and compare  $E_R^{n+1}=\phi^{n+1}/c$
- ▶ 300,000 histories/step, 100 spatial cells, **FOM=0.53**

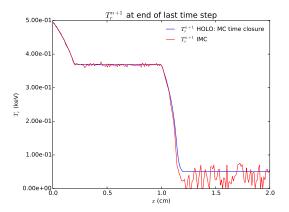


### With sufficient histories in a mix of optical thicknesses HOLO is more efficient than IMC

- ▶ 3 Material regions:  $\sigma_a = 100 \text{ cm}^{-1}$ ,  $\sigma_a = 0.01 \text{ cm}^{-1}$ ,  $\sigma_a = 100 \text{ cm}^{-1}$ .
- ▶  $10^6$  histories per time step, 200 cells, t = 2 sh

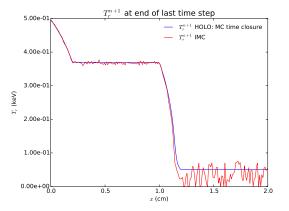
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- ▶  $10^6$  histories per time step, 200 cells, t = 2 sh



 HOLO with Backward Euler has same temporal accuracy, but better statistics for this problem A few other potential improvements were investigated:

Resolving negative intensities in optically thick cells using an artificial source and alternate trial space

Using HO solution to estimate spatial closure by including face tallies in ECMC

Source iteration with diffusion synthetic acceleration to iteratively solve the LO equations

ECMC is very efficient for TRT simulations and fits well in global HOLO method

The LO system can resolve nonlinearities with bounded angular consistency terms

Next step is to extend to higher dimensions main hurdle to overcome is infrastructure

ECMC is very efficient for TRT simulations and fits well in global HOLO method

The LO system can resolve nonlinearities with bounded angular consistency terms

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More details in: S.R. Bolding, M. Cleveland, and J.E. Morel. A HOLO Algorithm with ECMC for Radiative Transfer. NS&E: M&C 2015 Special Issue, 2016. Accepted.

## Simulations of Neutron Multiplicity Experiments with Perturbations to Nuclear Data

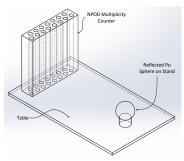
Simon Bolding and C.J. Solomon

4 August 2016





# Multiplicity experiments were performed for validating subcritical simulations



\*Not to scale

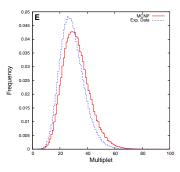
#### **Experimental Parameters**

- ▶ 94% <sup>239</sup>Pu sphere
- ▶ 5 reflector configurations None to 3.0 cm HDPE shells

Experiments repeated w/ <sup>252</sup>Cf

C.J. Solomon modeled experiments with modified MCNP5

# MCNP5 simulated multiplicity distributions showed discrepancy with experiments for Pu, but not for <sup>252</sup>Cf



Pu with 3.0-cm HDPE reflector

#### Previous work by Mattingly [2010]

- ► Caused by <sup>239</sup>**Pu** nuclear data
- ▶ Decreased energy-integrated  $\overline{\nu}$  ,  $\overline{\nu}$  : mean # of neutrons/fission

#### **ENDF-VII** raised $\overline{\nu}$ for <sup>239</sup>Pu to match $k_{\rm eff}$ benchmarks

ightharpoonup is  $\sim 2\,\sigma$  above measured data for  $E < 1.5~{
m MeV}$ 

Can we reduce discrepancy in multiplicity distributions without significantly altering  $k_{\text{eff}}$ ?

Perform energy-dependent perturbations of  $\overline{\nu}(E)$  in <sup>239</sup>Pu Random samples drawn from ENDF-VII.1 covariance data

Compare experimental and simulated multiplicity dist. and a  $k_{\rm eff}$  benchmark (Jezebel)

Compare  $\overline{\nu}(E)$  results to uniform shifts of microscopic cross sections

# LANL Python nuclear data library was modified to generate energy-dependent $\overline{\nu}$ samples

- 1. Generate a correlated sample of  $\overline{\nu}(E)$ 
  - Assumed multivariate Gaussian with group-averaged covariances

2. Modify continuous  $\overline{\nu}(E)$  data in **ACE** file

3. Perform all MCNP simulations with modified ACE data

## A cost function provides a measure of inaccuracy for each data realization

Reduced  $\chi^2$  values for the 5 multiplicity experiments and criticality benchmark

$$\chi^{2}_{\text{red,mult},m} = \frac{1}{N_{bins} - 1} \sum_{i=1}^{N_{bins}} \frac{(P_{i}^{\text{exp}} - P_{i}^{\text{mcnp}})^{2}}{\sigma^{2}(P_{i}^{\text{exp}}) + \sigma^{2}(P_{i}^{\text{mcnp}})}$$

Equally weight  $\chi^2$  values in a cost function

A lower score indicates higher accuracy

$$\mathsf{Cost} = \sum_{m=1}^{5} \chi^2_{\mathsf{red},\mathsf{mult},m} + \chi^2_{\mathsf{red},k_{\mathsf{eff}}}$$

# Multiplicity and $k_{\rm eff}$ simulations were performed for 500 unique realizations of $\overline{\nu}$ data

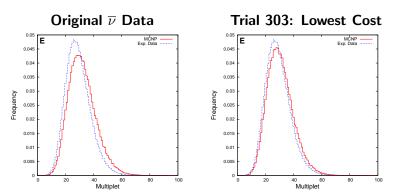
Trial	Cost	$\chi^2_{k_{\mathrm{eff}}}$
$\overline{ u}$ -1.14%	164.24	33.66
303	197.07	4.18
55	267.9	0.01
Original	426.86	0.27

MCNP criticality test suite performed for best data which includes 39 criticality benchmarks  $w/^{239}Pu$ :

Trial	RMSD
$\overline{ u}$ -1.14%	1.23%
303	0.51%
Original	0.49%

# Energy-dependent $\overline{\nu}$ perturbations improved all 5 multiplicity distributions

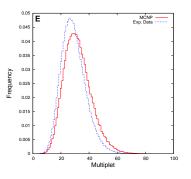
▶ Plots for best data realization and 3.0 cm HDPE case



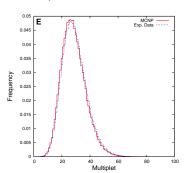
▶ Best data set reduced bias in  $1^{st}$  and  $2^{nd}$  moments, averaged over all 5 simulations, by  $\sim 35\%$ 

### Adjusting the fission cross section **uniformly** showed good correction to multiplicity simulations





 $\sigma_f$  decreased 1.5%



- ▶ High accuracy for all simulations:  $\sum \chi^2_{red,mult,m} = 14.6$
- $k_{\text{eff}}$  is **not** preserved:  $\chi^2_{k_{\text{eff}}} = 22.6$

### Simulations of Multiplicity Experiments with Nuclear Data Perturbations

Energy-dependent  $\overline{\nu}$  perturbations reduced inaccuracies in multiplicity while preserving  $k_{\rm eff}$ 

- ▶ Majority of cross-correlation terms  $\mathcal{O}(10^{-4})$  or less
- σ<sub>f</sub> may need more investigation not sensitive to capture cross section

Subcritical simulations should be considered in validation of nuclear data

Next step is to extend methodology for evaluating nuclear data

More details for nuclear data uncertainty work in

S.R. Bolding and C.J. Solomon. Simulations of Multiplicity Distributions with Perturbations to Nuclear Data. Proceedings, ANS Winter Meeting, 2013.

A radiation-hydrodynamics project I did not discuss today was recently submitted to JCP:

S.R. Bolding, J. Hansel, R.B. Lowrie, J.D. Edwards, and J.E. Morel. Second-Order Discretization in Space and Time for Radiation-Hydrodynamics. Journal of Computational Physics, 2016. *Submitted*.

# Simulations of Neutron Multiplicity Experiments with Nuclear Data Perturbations

S.R. Bolding<sup>1</sup>, C.J. Solomon<sup>2</sup>

<sup>1</sup>Texas A&M University, College station, TX <sup>2</sup>Los Alamos National Laboratory, Los Alamos, NM

4 August 2016





#### Backup Slides

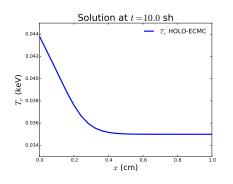
S.R. Bolding<sup>1</sup>, C.J. Solomon<sup>2</sup>

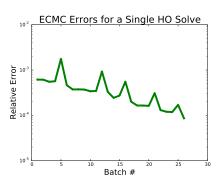
<sup>1</sup>Texas A&M University, College station, TX <sup>2</sup>Los Alamos National Laboratory, Los Alamos, NM



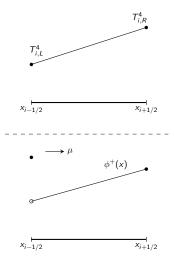


# Exponential convergence can be maintained if the LDFE mesh resolves the solution reasonably





# We can close equations with HO angular information and a linear-discontinuous (LD) spatial discretization



- 1. Assume T(x) and  $T^4(x)$  are LD
- 2. A lagged  $\tilde{l}_{\text{HO}}^{n+1}$  is used to evaluate consistency terms
- 3. Eliminate  $\phi_{i+1/2}^{\pm}$  with LD closure preserving equi. diff. limit

$$\phi_{i+1/2}^+ = 2\langle \phi \rangle_{R,i}^+ - \langle \phi \rangle_{L,i}^+$$

4. Global system solved with Newton's method and lagged **implicit opacities** Energy is always conserved

# We need a way to resolve issues when the LDFE representation of the intensity is negative

Negative intensities can occur in optically thick cells Mesh refinement is of minimal use

 $\tilde{l}_{HO}(x,\mu)$  must be positive for consistency terms to produce a physical, stable LO solution

Independent fix up for LO solution E.g., lumping or preserving balance with floored  $\phi(x)$ 

Can add source  $\delta$  to produce a positive projection  $\tilde{I}_{pos}$  such that  $\tilde{I}_{pos}$  satisfies the latest residual equation

Produce  $\tilde{l}_{pos}$  by scaling  $x - \mu$  moments equally, to estimate source for the next iteration

$$\mathbf{L}\tilde{I}^{(m)} = q - r^{(m)}$$

$$\mathbf{L}\tilde{I}^{(m)}_{\text{pos}} = q - r^{(m)} + \delta^{(m+1)}$$

$$q \to q + \delta^{(m+1)}$$

We can delay error stagnation Investigating alternative positive projection of /

#### Solving LO System with Newton's Method

► Linearization:

$$\underline{B}(T^{n+1}) = \underline{B}(T^*) + (T^{n+1} - T^*) \frac{\partial \underline{B}}{\partial t} \Big|_{t=0}$$

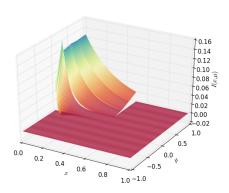
► Modified system

$$\left[\mathbf{D}(\mu^{\pm}) - \sigma_{a}^{*}(1 - f^{*})\right] \underline{\Phi}^{n+1} = f^{*}\underline{B}(T^{*}) + \frac{\underline{\Phi}^{n}}{c\Delta t}$$
$$\hat{\mathbf{D}}\Phi^{n+1} = \underline{Q}$$

$$f = \left(1 + \sigma_a^* c \Delta t \frac{4aT^{*3}}{\rho c_V}\right)^{-1}$$
  $T_i^* = \frac{T_{L,i}^* + T_{R,i}^*}{2}$ 

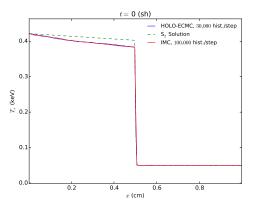
- ▶ Equation for  $T^{n+1}$  based on linearization that is conservative
- ▶ Converge  $T^{n+1}$  and  $\langle \phi \rangle$  with Newton Iterations

The angular flux for the two material problem is difficult to resolve near  $\mu=\mathbf{0}$ 



### Two Material Problem, comparison in optically thin region

▶ Plot of radiation temperature after 10 time steps



Backup slide with timing results

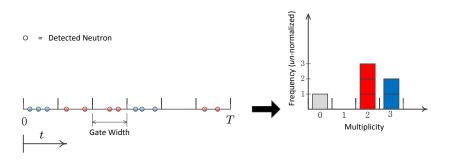
#### Forming the LO System

► Taking moments of TE yields 4 equations, per cell i, e.g.

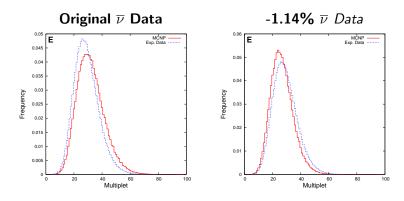
$$\begin{split} -2\mu_{i-1/2}^{n+1,+}\phi_{i-1/2}^{n+1,+} + \{\mu\}_{L,i}^{n+1,+} &\langle\phi\rangle_{L,i}^{n+1,+} + \{\mu\}_{R,i}^{n+1,+} &\langle\phi\rangle_{R,i}^{n+1,+} + \\ \left(\sigma_{t}^{n+1} + \frac{1}{c\Delta t}\right) h_{i}\langle\phi\rangle_{L,i}^{n+1,+} - \frac{\sigma_{s}h_{i}}{2} \left(\langle\phi\rangle_{L,i}^{n+1,+} + \langle\phi\rangle_{L,i}^{n+1,-}\right) \\ &= \frac{h_{i}}{2} \langle\sigma_{a}^{n+1} acT^{n+1,4}\rangle_{L,i} + \frac{h_{i}}{c\Delta t} \langle\phi\rangle_{L,i}^{n,+}, \quad (1) \end{split}$$

- ► Cell unknowns:  $\langle \phi \rangle_{L,i}^+$ ,  $\langle \phi \rangle_{R,i}^+$ ,  $\langle \phi \rangle_{L,i}^-$ ,  $\langle \phi \rangle_{R,i}^-$ ,  $\langle \phi \rangle_{R,i}^-$ ,  $\langle \phi \rangle_{R,i}^-$
- Need angular consistency terms and spatial closure (LD)

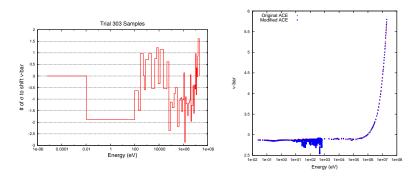
# Neutron multiplicity distributions provide passive multiplication information about a fissionable system



### Energy-Integrated $\overline{\nu}$ Shift – 3.0 cm HDPE reflector



### The best $\overline{\nu}$ data (trial 303)



## Fractional shifts to cross sections were made for comparison

Adjusted cross section uniformly at all energies compensated with  $\sigma_{tot}$  or  $\sigma_{el}$ 

Increasing capture cross section was not effective relative to variance in data

Scaling fission cross section 1.5% ( $< 1\sigma$ ) improved multiplicity distributions

- Adjust elastic scattering  $(\sigma_{el})$  to compensate change in  $\sigma_f$ , for E>1 keV
- lacktriangle Better improvement than *uniform* scaling of  $\overline{
  u}$