

# A High-Order Low-Order Algorithm with Exponentially-Convergent Monte Carlo for $k$ -Eigenvalue Problems

Simon Bolding and Jim Morel

11 November 2014

ANS Winter Meeting



# Outline

- 1 Overview
- 2 Low-Order Solver
- 3 High-Order Solver
- 4 High-Order Low-Order Algorithm
- 5 Results
- 6 Conclusions

# A High-Order Low-Order Solution

## Basic Idea

Build a low-order (LO) system that can be efficiently solved, with high-order (HO) correction from pure absorber MC simulations

- Useful when global solution of particle density needed, with low statistical noise
- The LO solution preserves the HO solution, upon convergence (consistent)
- The LO solver handles non-linearities and in-scattering source
  - Produces a linear-discontinuous (LD) representation of sources
  - Lower **dimensional** problem
- The HO system is a fixed-source transport problem
  - ECMC allows for the statistical noise to be reduced globally, giving accurate consistency terms

# The Thermal Radiative Transfer Equations

- 1D, gray, isotropic scattering, backward Euler
- $\sigma$  can be a function of  $T$
- Transport equation

$$\mu \frac{\partial I^{n+1}}{\partial x} + \left( \sigma_t + \frac{1}{c \Delta t} \right) I^{n+1} = \frac{\sigma_s}{2} \phi^{n+1} + \frac{1}{2} (\sigma_a a c T^4)^{n+1} + \frac{I^n}{\Delta t c}$$

- Material energy equation

$$\rho c_v \frac{T^{n+1} - T^n}{\Delta t} = \int_{-1}^1 \sigma_a I^{n+1}(x, \mu) d\mu - (\sigma_a a c T^4)^{n+1}$$

$$\phi(x) = \int_{-1}^1 I^{n+1}(x, \mu) d\mu$$

# Eigenvalue Problem

- 1D transport equation

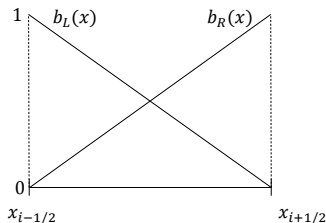
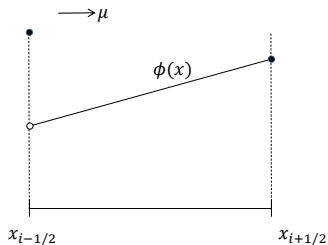
$$\mu \frac{\partial \psi}{\partial x} + \Sigma_t \psi(x, \mu) = \frac{1}{4\pi} \left( \Sigma_s + \frac{\nu \Sigma_f}{k_{\text{eff}}} \right) \phi(x)$$
$$\phi(x) = 2\pi \int_{-1}^1 \psi(x, \mu) d\mu$$

- MC solutions typically use power iteration
  - Accelerate source convergence with LO solution (e.g., CMFD)

We will use a low-order solution that handles the scattering and fission operators to determine  $k_{\text{eff}}$  and  $\phi(x)$ . The LO solution preserves MC transport solution

# LO Discretization & Space-Angle Moments

- Linear Discontinuous (LD) finite elements in space and half range angular integrals



Spatial moments

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

Angular half-ranges

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

# LO System

- Taking moments of TE yields **4 equations**, per cell  $i$ , e.g.

$$-2\mu_{i-1/2}^+ \phi_{i-1/2}^+ + \langle \mu \rangle_{L,i}^+ \langle \phi \rangle_{L,i}^+ + \langle \mu \rangle_{R,i}^+ \langle \phi \rangle_{R,i}^+ + \Sigma_t h_i \langle \phi \rangle_{L,i}^+ - \frac{\Sigma_s h_i}{4\pi} \left( \langle \phi \rangle_{L,i}^+ + \langle \phi \rangle_{L,i}^- \right) = \frac{1}{k_{\text{eff}}} \frac{\nu \Sigma_f h_i}{4\pi} \langle \phi \rangle_{L,i}^+$$

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

# Computing LO Consistency terms from HO Solution

- For  $\mu > 0$ ,  $L$  moment

$$\langle \mu \rangle_{L,i}^+ \simeq \frac{\frac{2}{h_i} \int_0^1 \int_{x_{i-1/2}}^{x_{i+1/2}} \mu b_{L,i}(x) \tilde{\psi}^{HO}(x, \mu) d\mu dx}{\frac{2}{h_i} \int_0^1 \int_{x_{i-1/2}}^{x_{i+1/2}} b_{L,i}(x) \tilde{\psi}^{HO}(x, \mu) d\mu dx} \quad (1)$$

- ECMC gives LDfE  $\tilde{\psi}^{HO}(x, \mu)$  for computing terms directly
- To get  $S_2$ , set  $\langle \mu \rangle^\pm = \pm \frac{1}{\sqrt{3}}$



# Solving LO System with Power Iteration

- Global System: 
$$\mathbf{D}\Phi = \frac{1}{k_{\text{eff}}} \mathbf{F}\Phi$$

## Algorithm

- 1 Guess  $\Phi^{(0)}$  and  $k_{\text{eff}}^{(0)}$

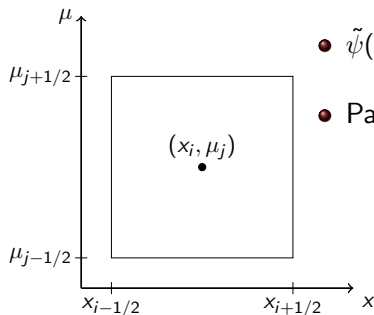
$$\Phi^{(l+1)} = \frac{1}{k_{\text{eff}}^{(l)}} \mathbf{D}^{-1} \mathbf{F} \Phi^{(l)}$$
$$k_{\text{eff}}^{(l+1)} = k_{\text{eff}}^{(l)} \frac{\int \nu \Sigma_f \phi^{(l+1)} dx}{\int \nu \Sigma_f \phi^{(l)} dx}.$$

- 2 Converge  $\phi(x)$  and  $k_{\text{eff}}$
- 3 **Accelerate**  $\Phi^{(l)}$  and  $k_{\text{eff}}^{(l+1)}$  after each power iteration with Nonlinear Krylov Acceleration (NKA)

# Exponentially Convergent Monte Carlo

- An iterative form of residual Monte Carlo
- Requires a functional form of the angular flux in space and angle
  - Use finite element representation
- Can reduce statistical error globally  $\propto e^{-\alpha N}$ 
  - Does not make difficult problems easier
- Adaptive mesh refinement allows the error to continue to be reduced
  - Increase  $N$  per batch by factor of new cells added

# Space-Angle Mesh and MC Implementation Details



- $\tilde{\psi}(x, \mu) = \psi_{a,i} + \psi_{x,i} \frac{2}{h_x}(x - x_i) + \psi_{\mu,i} \frac{2}{h_\mu}(\mu - \mu_i)$

- Path-length estimators of moments, e.g.

$$\psi_{x,i} = \frac{6}{h_x^2 h_\mu} \iint_{\mathcal{D}} (x_{c,j} - x_i) \psi(x, \mu) dx d\mu$$

- Particles are allowed to stream,  $w(s) = w_0 e^{-\Sigma_t s}$
- LDFE and upwinding **eliminates** surface tallies
- Cell-wise, global representation allows for **stratified** sampling
  - $N_{i,j} \propto |r_i(x, \mu)|$
  - Force  $N_i \geq N_{\min}$  and adjust particle weights

# High Order System and ECMC Algorithm

- **Pure absorber** transport problem

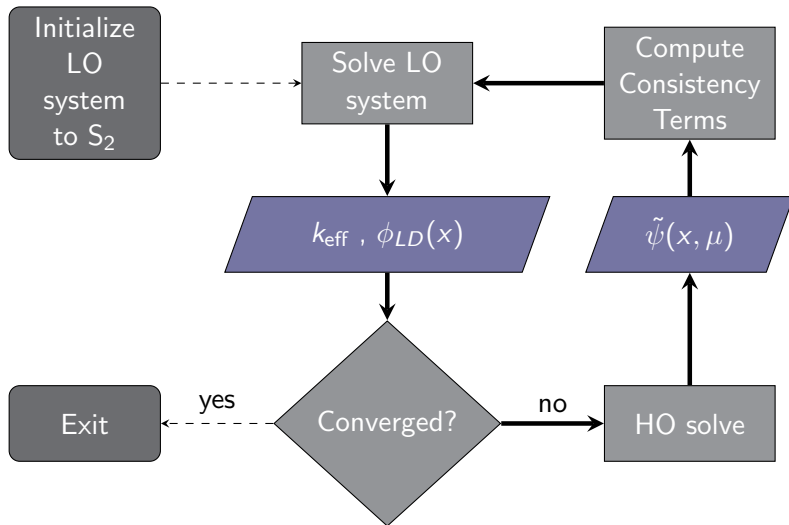
$$\left[ \mu \frac{\partial}{\partial x} + \Sigma_t \right] \psi(x, \mu) = \frac{1}{4\pi} \left( \Sigma_s + \frac{1}{k_{\text{eff}}^{LO}} \right) \phi^{LO}(x)$$

$$\mathbf{L}\psi = q^{LO}$$

## ECMC Algorithm

- Residual Equation:  $\mathbf{L}\tilde{\epsilon}^{(m)} = \tilde{r}^{(m)} = q^{LO} - \mathbf{L}\tilde{\psi}^{(m)}$
- Compute  $\tilde{\epsilon}^{(m)} = \mathbf{L}^{-1}\tilde{r}^{(m)}$  with MC, **projecting** the solution
- Update:  $\tilde{\psi}^{(m)} = \tilde{\psi}^{(m-1)} + \tilde{\epsilon}^{(m)}$ 
  - If  $\tilde{\epsilon}$  is reduced each batch, **exponential convergence achieved**
  - $h$ -refine when  $\epsilon(x, \mu)$  not represented sufficiently
  - Must sufficiently reduce noise in  $\epsilon$  tallies, each batch
- Repeat until  $\|\tilde{\epsilon}\|_2 < \text{tol} \times \|\psi\|_2$

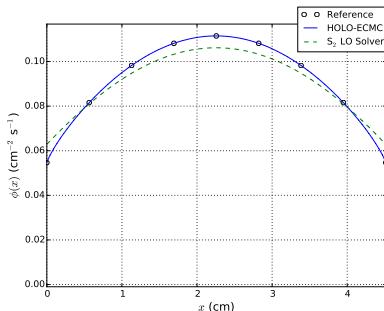
# High-Order Low-Order Algorithm



# Critical slab benchmark

## Problem Parameters

- $k_{\infty} = 2.29$ ,  $\Sigma_t = 0.326 \text{ cm}^{-1}$
- $2.4 \times 10^5$  histories per batch,  $100 \times$  &  $20 \mu$  cells
- Adaptive HO convergence

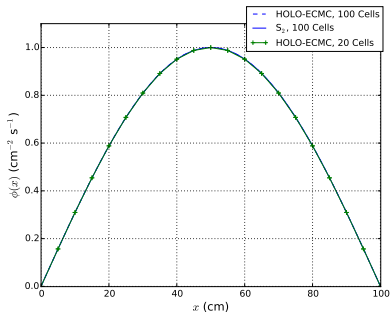


- $|\Delta\Phi| < 10^{-4}$  in 4 outer iterations, using  $\sim 2.4 \times 10^7$  histories
- For 10 independent simulations:
  - $\overline{k_{\text{eff}}} = 0.999998$ ,  
 $\sigma(k_{\text{eff}}) = 4.1 \times 10^{-6}$ ,  
 $\text{max dev.} = 1.1 \times 10^{-5}$
  - $\overline{\sigma_{\text{rel}}(\phi_i)} = 1.4 \times 10^{-5}$

# Optically thick, near-critical slab

## Problem Parameters

- $k_{\infty} = 1$ ,  $\Sigma_t = 30.0 \text{ cm}^{-1}$ ,  $\Sigma_s = 29.5 \text{ cm}^{-1}$ ,  $\text{DR} \approx 0.999$
- Relative Tolerance of  $1.0\text{E-}05$  for all solvers

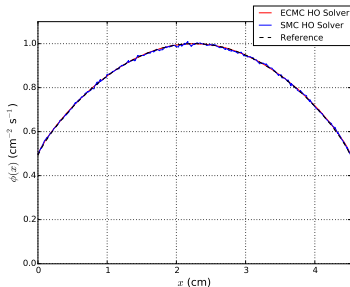


- Fission source convergence:
  - Power iteration not converged after 10,000 iterations
  - NKA converged in 246 iterations
- 1 outer iteration,  $4.8 \times 10^6$  total histories
- $k_{\text{eff}} = 0.99997$

# Comparison of statistical noise for standard and ECMC HO solvers

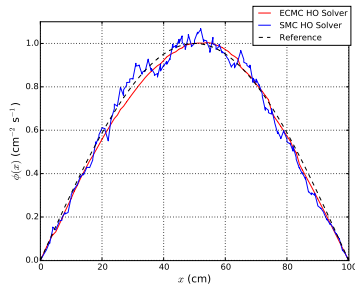
- One HO solve, with a fixed number of histories  $N$ .  
Comparison of **ECMC** with 5 batches or standard MC (**SMC**)

## Critical Slab



$$N = 1.5 \times 10^5$$

## Diffuse Slab, $k_{\infty} = 1$



$$N = 10^6$$



## Current & Future Development

- Can Solve for  $k_{\text{eff}}$  and fission source with HOLO method
  - Pure absorber histories are cheaper than standard MC simulations
  - ECMC efficiently reduces noise globally
  - LO solver handles scattering, fission, and
- Stratified source sampling more efficient at reducing statistical variance than a constant number of particles per cell
- Need to use the estimated statistical error in tallies for  $\tilde{\epsilon}(x, \mu)$

# Questions?

## A High-Order Low-Order Algorithm with Exponentially-Convergent Monte Carlo for $k$ -Eigenvalue Problems

Simon Bolding and Jim Morel

11 November 2014



## Backup Slides



## Algorithm

- ❶  $\tilde{\psi}^{(0)} = \tilde{\psi}$  or from last batch this time step
- ❷ Using Monte Carlo, solve  $\epsilon^{(m)} = \mathbf{L}^{-1} \tilde{r}^{(m)}$ 
  - Use volumetric tallies, weighted with  $x$  and  $\mu$  basis moments time  $\psi$  to construct LD  $\tilde{\epsilon}^{(m)}(x, \mu)$  over the current space-angle mesh
- ❸  $\tilde{\psi}^{(j+1)} = \tilde{\psi}^{(m)} + \tilde{\epsilon}^{(m)}$
- ❹ **IF** error stagnation:
  - Refine mesh based on relative jump error in  $\tilde{\psi}(x, \mu)$
- ❺ Repeat 2-4 until  $\|\tilde{\epsilon}\|_2 < \text{tol} \times \|\psi\|_2$

## Algorithm

- ➊ Initialize  $\langle \mu \rangle^\pm$  parameters to  $S_2$
- ➋ Solve LO system using power iteration
- ➌ Build  $q^{LD}$  for HO solver, and set  $\tilde{\psi}$  to latest HO estimate on coarsest  $x-\mu$  mesh
- ➍ Solve  $\tilde{\psi}(x, \mu) = \mathbf{L}^{-1} q^{LO}$  using ECMC
- ➎ Compute new  $\langle \mu \rangle^\pm$  parameters using  $\tilde{\psi}^{HO}$  over LO mesh
- ➏ Repeat 2-5 until  $\underline{\phi}^{LO}$  is converged
  - Use adaptive convergence criteria

# Exponential Convergence Plot