

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

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# Outline

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

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# Overview

- We are interested in modeling thermal radiation transport in the high-energy density physics regime
  - Temperatures on order of  $10^6$  K or more
  - Significant energy and momentum may be exchanged with material

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  - Temperatures on order of  $10^6$  K or more
  - Significant energy and momentum may be exchanged with material
- Radiative transfer simulations important in modeling:
  - Material under extreme conditions
  - Inertial confinement fusion
  - Supernovae and other astrophysical phenomena.

# The Grey TRT Equations

- The 1D, grey equations

$$\frac{1}{c} \frac{\partial I}{\partial t} + \mu \frac{\partial I}{\partial x} + \sigma_t I(x, \mu) = \frac{1}{2} \sigma_a a c T^4(x),$$

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

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

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- Fundamental unknowns are the radiation intensity  $I(x, \mu)$  and material temperature  $T(x)$
- Absorption cross section ( $\sigma_a$ ) can be a strong function of  $T$
- Equations are nonlinear and may be tightly coupled

# The Implicit Monte Carlo Method

- Equations often solved with Monte Carlo (MC) via the Implicit Monte Carlo (IMC) method
- The material energy equation is linearized over a time step and eliminated from the system
  - Results in linear transport equation with a discrete emission and scattering terms
  - Continuous integration of time derivative



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  - Results in linear transport equation with a discrete emission and scattering terms
  - Continuous integration of time derivative
- Drawbacks
  - Effective scattering cross section can be very large
    - Local acceleration methods based on discrete diffusion approximations
  - Nonlinearities not converged
  - Reconstruction of linear source shape in cell

# An alternative High-Order Low-Order approach

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  - Non-linear temperature dependence converged with Newton's method
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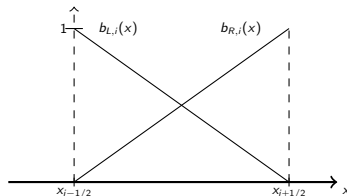
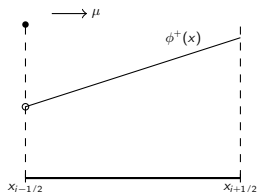
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  - **Output:** a linear-discontinuous (LD) FE representation of scattering and emission source
- The HO system is a fixed-source, pure-absorber transport problem
  - No effective scattering events
  - Solved with ECMC for efficient reduction of statistical noise
  - **Output:** angular consistency terms

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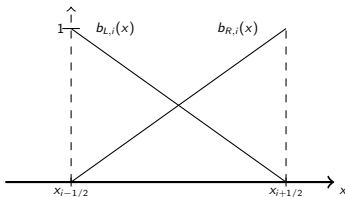
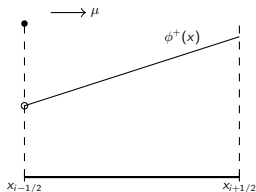
# LO Discretization & Space-Angle Moments

- The LO system is “S<sub>2</sub>-like” equations (similar to [Wolters 2013]), with a backward Euler discretization in time
- Linear discontinuous (LD) FE in space for  $\phi$ ,  $T$ , and  $T^4(x)$



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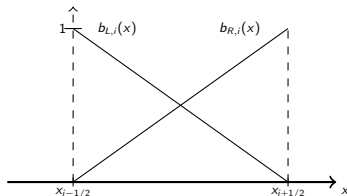
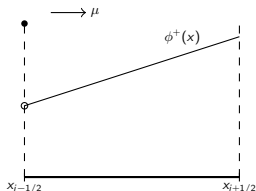
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- Half-range integrals over  $+$  and  $- \mu$
- Examples of moments:

Spatial: left basis

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

Angular: positive flow

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



## Forming the LO System

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$$\{\mu\}_{L,i}^+ := \frac{\int_0^1 \int_{x_{i-1/2}}^{x_{i+1/2}} \mu b_{L,i}(x) I^{n+1}(x, \mu) dx d\mu}{\int_0^1 \int_{x_{i-1/2}}^{x_{i+1/2}} b_{L,i}(x) I^{n+1}(x, \mu) dx d\mu}$$

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- 3 Use  $\tilde{I}_{HO}^{n+1}(x, \mu)$  & LD spatial closure to close the system
  - Cell unknowns:  $\langle \phi \rangle_{L,i}^{n+1,+}$ ,  $\langle \phi \rangle_{R,i}^{n+1,+}$ ,  $\langle \phi \rangle_{L,i}^{n+1,-}$ ,  $\langle \phi \rangle_{R,i}^{n+1,-}$ ,  $T_{L,i}$ ,  $T_{R,i}$
  - Non-linear, fully discrete system, solved with approximate Newton's method

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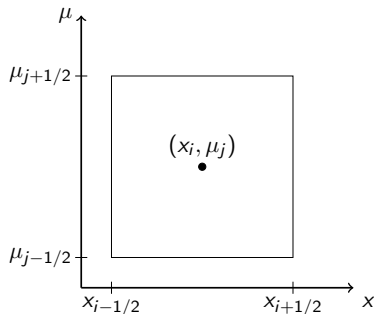
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# Overview of Exponentially Convergent Monte Carlo

- Each batch tallies the **error** in current estimate of solution
  - Can reduce statistical error **globally**  $\propto e^{-\alpha N}$
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  - In TRT problems, old angular intensity provides a *very good* guess of new solution, significantly reducing the required number of histories
- Requires a **discretized** form of the angular intensity  $\tilde{I}(x, \mu)$ 
  - Use **projection** of the solution onto a space-angle FE mesh
  - LD FE  $\tilde{I}^{HO}(x, \mu)$  allows for direct computation of consistency terms



# High Order System and ECMC Algorithm

- **Pure absorber** transport problem because we have LD representation of  $T^4$  from LO solution

$$\left[ \mu \frac{\partial}{\partial x} + \left( \sigma_t + \frac{1}{c \Delta t} \right) \right] I^{n+1}(x, \mu) = \frac{I^n}{c \Delta t} + \boxed{\frac{1}{2} \sigma_a a c T_{LO}^{n+1,4}}$$
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  - Particles are allowed to stream along path  $s$ ,  $w(s) = w_0 e^{-\sigma_t s}$
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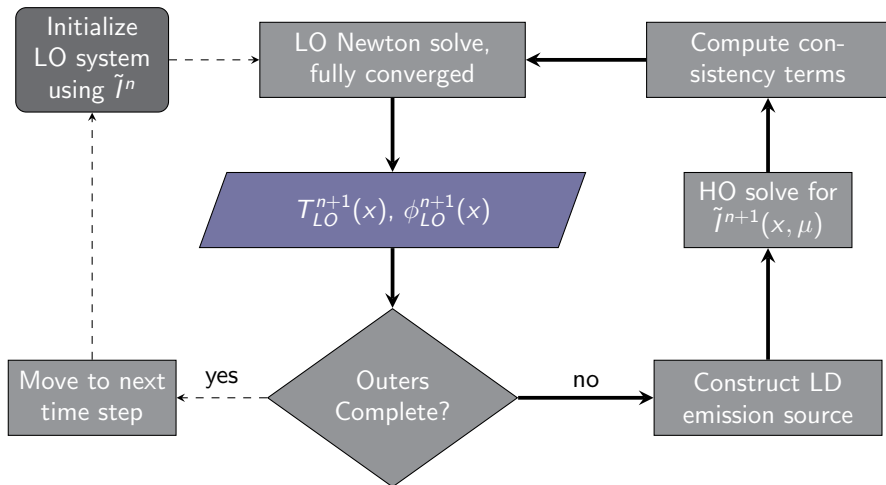
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  - Use cell-wise **systematic** sampling for  $r^{(m)}$  source
- Update:  $\tilde{I}^{n+1,(m)} = \tilde{I}^{n+1,(m-1)} + \tilde{\epsilon}^{(m)}$ 
  - If  $\tilde{\epsilon}$  is reduced each batch, **exponential convergence achieved**

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# Relevant Implementation Specifics

- Difficulties in resolving the solution near the wave-front
  - The LD representation for  $I$  results in negative values within a cell
  - In these results, no correction is applied to the HO solution, and the LO solution uses lumped LD and  $S_2$  equivalent terms in bad cells

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- Difficulties in resolving the solution near the wave-front
  - The LD representation for  $I$  results in negative values within a cell
  - In these results, no correction is applied to the HO solution, and the LO solution uses lumped LD and  $S_2$  equivalent terms in bad cells
- For all results
  - LO Newton iterations are fully converged each solve
  - Fixed  $\Delta t$  of 0.001 sh
  - One HO solve per time step

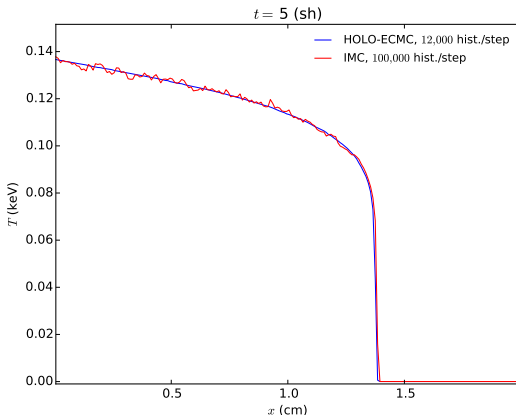


# Marshak Wave Test Problem

- From equilibrium, a radiation source is applied at the left boundary at  $t = 0$ . With  $\sigma_a \propto T^{-3}$ , energy slowly moves across the system.

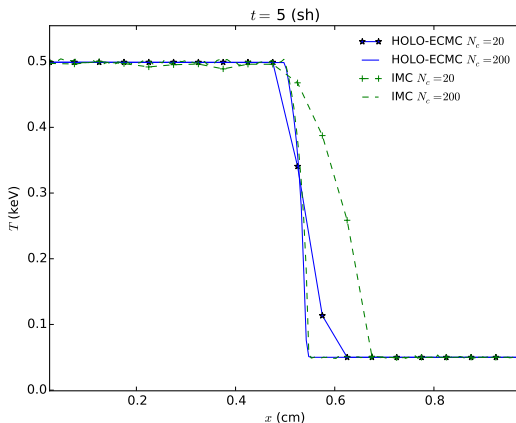
# Marshak Wave Test Problem

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- Transient solution after 5 shakes plotted as  $T_r = \sqrt[4]{\phi/ac}$ , with 200  $x$  cells



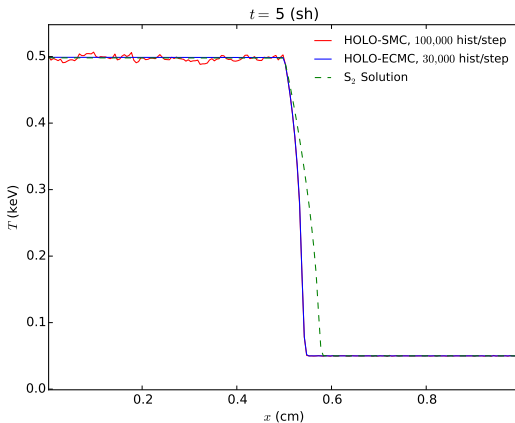
## Two Material Problem, Comparison of Spatial Convergence

- Problem the same as Marshak Wave, but with constant opacities and an optically thin (left) and optically thick (right) region



# Comparison of statistical noise for standard and ECMC HO solvers

- One HO solve, with a *fixed number of histories* per time step, for two different HO solvers: a comparison of **ECMC** with 3 batches and standard MC (**SMC**), as well as  $S_2$  solution



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## Current & Future Development

- Can accurately reproduce IMC results with HOLO method
  - ECMC requires significantly less particles
  - LO solver determines non-linear Material temperature distribution
  - Linear shape within a cell mitigates teleportation error
  - Very efficient for diffusive problems

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  - Linear shape within a cell mitigates teleportation error
  - Very efficient for diffusive problems
- Currently developing strategies for dealing with unresolvable solutions
- Next step will be to implement in 2 spatial dimensions

# Questions?

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

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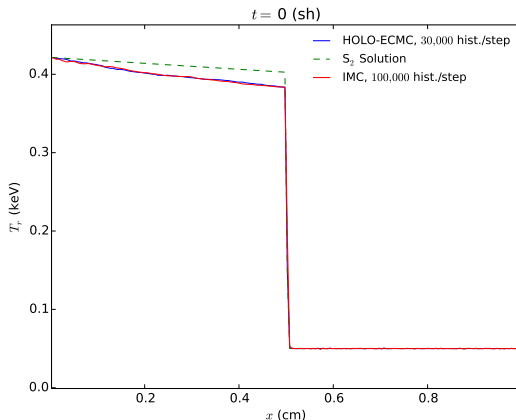
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## Two Material Problem, comparison in optically thin region

- Plot of radiation temperature after 10 time steps



## Algorithm

- 1 Initialize  $\tilde{I}^{n+1,(0)} := \tilde{I}^n$
- 2 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
- 3  $\tilde{\psi}^{(j+1)} = \tilde{\psi}^{(m)} + \tilde{\epsilon}^{(m)}$
- 4 **IF** error stagnation:
  - Refine mesh based on relative jump error in  $\tilde{\psi}(x, \mu)$
- 5 Repeat 2-4 until  $\|\tilde{\epsilon}\|_2 < \text{tol} \times \|\psi\|_2$

## Algorithm

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

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

$$\begin{aligned}
 & -2\mu_{i-1/2}^{n+1,+} \phi_{i-1/2}^{n+1,+} + \{\mu\}_{L,i}^{n+1,+} \langle \phi \rangle_{L,i}^{n+1,+} + \{\mu\}_{R,i}^{n+1,+} \langle \phi \rangle_{R,i}^{n+1,+} + \\
 & \left( \sigma_t^{n+1} + \frac{1}{c\Delta t} \right) h_i \langle \phi \rangle_{L,i}^{n+1,+} - \frac{\sigma_s h_i}{2} \left( \langle \phi \rangle_{L,i}^{n+1,+} + \langle \phi \rangle_{L,i}^{n+1,-} \right) \\
 & = \frac{h_i}{2} \langle \sigma_a^{n+1} a c T^{n+1,4} \rangle_{L,i} + \frac{h_i}{c\Delta t} \langle \phi \rangle_{L,i}^{n,+}, \quad (1)
 \end{aligned}$$

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

## Solving LO System with Newton's Method

- Linearization:  $\underline{B}(T^{n+1}) = \underline{B}(T^*) + (T^{n+1} - T^*) \left. \frac{\partial \underline{B}}{\partial T} \right|_{T^*}$
- Modified system

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

$$\hat{\mathbf{D}} \underline{\Phi}^{n+1} = \underline{Q}$$

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

- Equation for  $T^{n+1}$  based on linearization that is conservative
- Converge  $T^{n+1}$  and  $\langle \phi \rangle$  with Newton Iterations
  - Spatial representation can result in negative temperatures