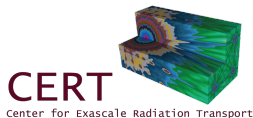


# A High-Order Low-Order Algorithm with Exponentially-Convergent Monte Carlo for Thermal Radiative Transfer

Simon Bolding

Advisor: Jim Morel

22 November 2016



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

Modeling materials under extreme conditions

Temperatures  $\mathcal{O}(10^6)$  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  $T(x, t)$

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

TRT equations are **nonlinear** and may be tightly coupled

Absorption cross section ( $\sigma_a$ ) can be a strong function of  $T$

Typically solved with implicit Monte Carlo (IMC)

which partially linearizes the system over a time step

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

# Our high-order low-order (HOLo) method improves on several drawbacks of standard IMC

## Standard IMC

Large **statistical noise** possible

**Effective scattering** can make MC tracking very expensive

Linearization can cause **non-physical** results (maximum principle violations)

Reconstruction of linear emission shape limits artificial energy propagation

## HOLo Method

ECMC is **efficient** for TRT

MC solution has **no scattering**

Fully **implicit** time-discretization and LO solution **resolves nonlinearities**

Linear-discontinuous FE for  $T(x)$  **preserving equilibrium diffusion limit**

# A HOLO Algorithm for Thermal Radiative Transfer

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Exponentially Convergent MC High-Order Solver

Derivation of the LO equations

Summary of algorithm

Computational Results

Monte Carlo time integration

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# Exponentially Convergent Monte Carlo (ECMC) can efficiently reduce noise globally

ECMC is an iterative form of residual Monte Carlo applied to particle transport problems

Each MC batch tallies the **error** in solution estimate

- ▶ standard MC particle transport, but a **complex** source
- ▶ ECMC requires a **functional** representation for all phase space variables being sampled

Can reduce solution error **globally**  $\propto e^{-\alpha N}$

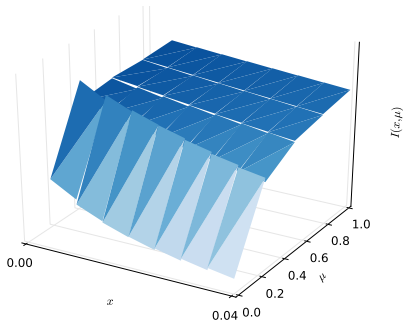
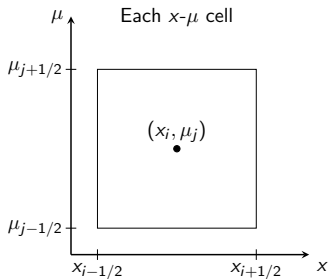
Adaptive *h*-refinement can help represent error



ECMC uses a **projection**  $\tilde{l}(x, \mu)$  onto a space-angle LDFF mesh to represent the solution

local volumetric tallies

$$\tilde{l}_{ij}(x, \mu) = \underset{\text{cell}}{I_a} + \frac{2}{h_x} \underset{\text{angle}}{I_x}(x - x_i) + \frac{2}{h_\mu} \underset{\text{angle}}{I_\mu}(\mu - \mu_i)$$



We apply the ECMC algorithm to the **pure-absorber** time-discrete transport equation

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

For each **batch**  $m$ :

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

## 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 each cell  
except for zero from cells in thermal equilibrium

# A HOLO Algorithm for Thermal Radiative Transfer



Exponentially Convergent MC High-Order Solver

Derivation of the LO equations

Summary of algorithm

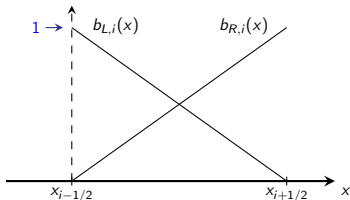
Computational Results

Monte Carlo time integration

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

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

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

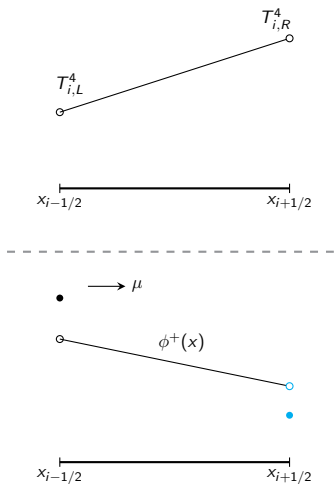
Ultimately, we get six **exact** moment equations  
for each spatial element  $i$

For example, apply  $\langle \cdot \rangle_{L,i}$  and  $(\cdot)^+$  to streaming term  
and perform algebra to form angular averages

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

Now, approximate angular consistency terms  
with  $\tilde{I}_{HO}^{n+1}(x, \mu)$  from previous HO solve

# We eliminate auxillary spatial unknowns to close global system of equations



1. Assume  $T(x)$  and  $T^4(x)$  are LD preserving equi. diff. limit

2. Assume  $\phi^\pm(x)$  linear over each cell

3. Can eliminate outflows with parametric closure from HO solution, e.g.,

$$\phi_{i+1/2}^+ = \gamma_{i,HO}^+ \langle \phi \rangle_{a,i}^+ + \langle \phi \rangle_{x,i}^+$$

where  $\gamma_{i,HO}^+ = 1$  gives LD

HO closure improves consistency  
Requires face tallies in ECMC

# Apply source-iteration with linear diffusion-synthetic acceleration (DSA) to linearized LO system

Used source iteration (SI) with WLA-D  
for (effective) scattering source of each Newton step

1. Sweep for a new  $\phi^\pm$   
with a lagged scattering source
2. Solve approximate **spatially continuous** diffusion equation for error in scattering iterations
3. Update moments w/ approx. local balance equations over each spatial element



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Recast as a GMRES solution with DSA-preconditioning

# A HOLO Algorithm for Thermal Radiative Transfer



Exponentially Convergent MC High-Order Solver

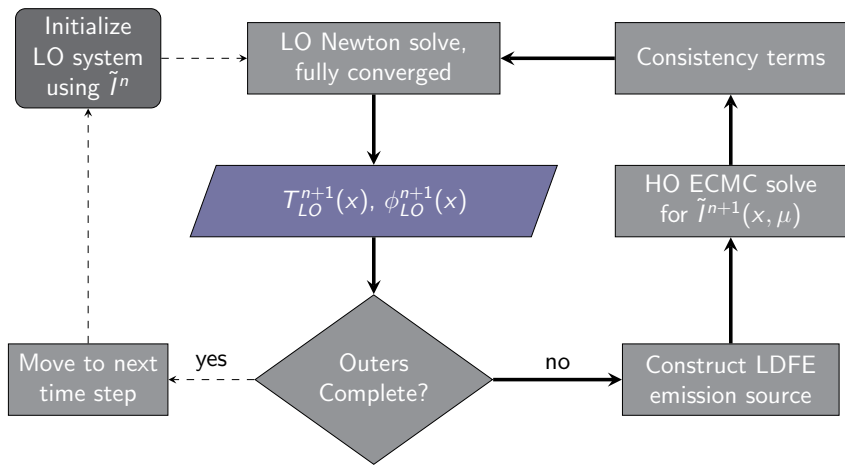
Derivation of the LO equations

Summary of algorithm

Computational Results

Monte Carlo time integration

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



# A HOLO Algorithm for Thermal Radiative Transfer

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## Implementation specifics for results are:

- ▶ HOLO method is written in stand-alone C++ (15k lines)  
IMC results from Jayenne (LANL code)
- ▶ Unless stated, one HO solve per time step,  
with two LO solves
  - ▶ *each HO solve* has 3 ECMC batches  
no adaptive mesh refinement

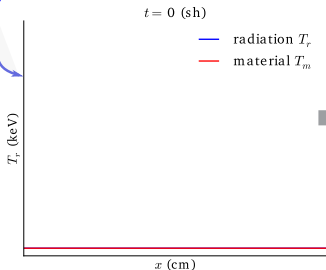
▶ Figure of Merit:

$$\text{FOM} = \frac{1}{\left( \frac{\|\sigma(\phi_i)\|}{\|\phi_i\|} \right)^2 N_{\text{total}}}$$

normalized so IMC FOM=1

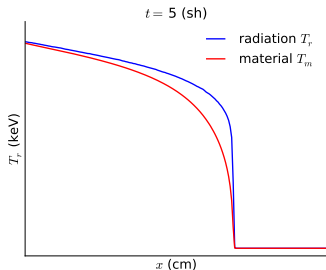
We will test our method with several standard **Marshak Wave** problems

constant radiation  
boundary source



$t = 0$  (sh)

— radiation  $T_r$   
— material  $T_m$



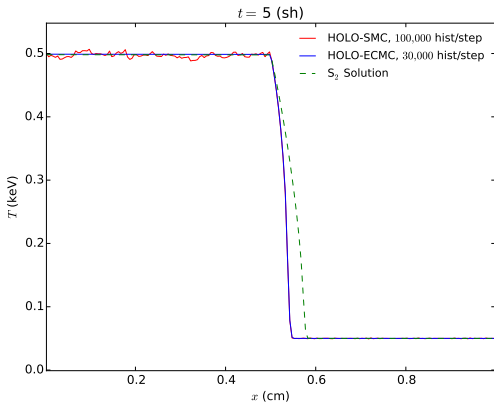
$t = 5$  (sh)

— radiation  $T_r$   
— material  $T_m$

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

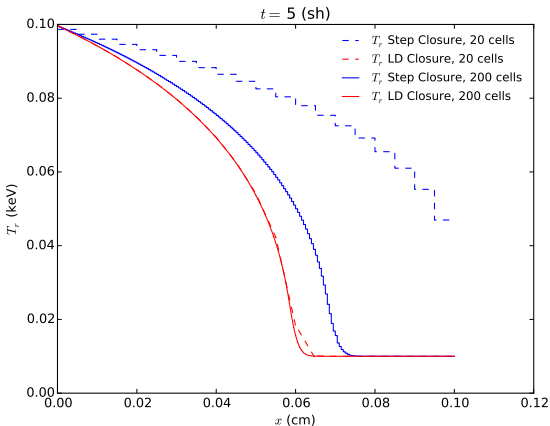
# ECMC is more efficient than standard MC as a HO solver

- ▶ Left half is optically thin ( $\sigma=0.2 \text{ cm}^{-1}$ ), right half is thick ( $\sigma_a=2000 \text{ cm}^{-1}$ ).  $8 \mu$  cells,  $\Delta t = 0.001 \text{ sh}$
- ▶ Results for HOLO with different HO solvers:  
ECMC (FOM=10,000), standard MC (FOM=0.46), and  $S_2$



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

- EDL Problem: Large, constant  $\sigma_a$  and small  $c_v$
- Apply HOLO algorithm, 12k histories per step



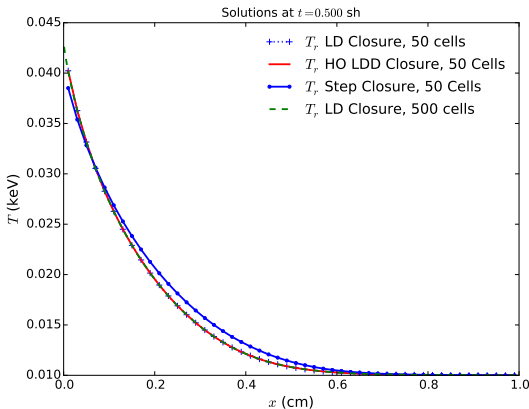


# Tested the HO spatial closure for a easier TRT problem

- Fairly diffusive domain so that linear discontinuous  $T(x)$  is positive

$\Delta t$  increased from 0.001 sh to  $\Delta t_{\max} = 0.01$  sh

- Use ECMC with adaptive refinement, 2 HOLO iters and 585k histories per HO solve



The HO spatial closure improves  $L_2$  error and consistency but does not decrease error in cell-averaged quantities

- Compute errors for 50 cells, compared to 500 cell reference.  
Average results from 20 independent simulations

	$\ e(x)\ _{2,rel}$	$\ e_i\ _{2,rel}$	$\ \phi_{HO}(x) - \phi_{LO}(x)\ _{2,rel}$
LD FE Closure	1.60%	0.59%	0.76%
HO Closure	1.40%	0.67%	0.013%

\* $1\sigma \leq 0.01\%$  for all results

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- For ECMC,  $\tilde{I}_{HO}(x, \mu)$  does *not satisfy* moment equations

$$\mathbf{L}\tilde{I}(x, \mu) = q - r(x, \mu)$$

- Issues with using HO spatial closure and lumping  
but standard lumped LD FE is accurate

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

- Desire a positive  $\tilde{l}_{HO}(x, \mu)$  for consistency terms to produce a physical, stable LO solution
- Rotate negative  $\tilde{l}(x, \mu)$  above  $l_{\text{floor}}$  at end of batch:

$$\tilde{l}_{\text{pos}} = l_a + \mathbf{C} \left[ \frac{2}{h_x} l_x (x - x_i) + \frac{2}{h_\mu} l_\mu (\mu - \mu_j) \right], \quad (x, \mu) \in \mathcal{D}_{ij},$$

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- Optionally add artificial **source** to next batch as attempt to mitigate stagnation

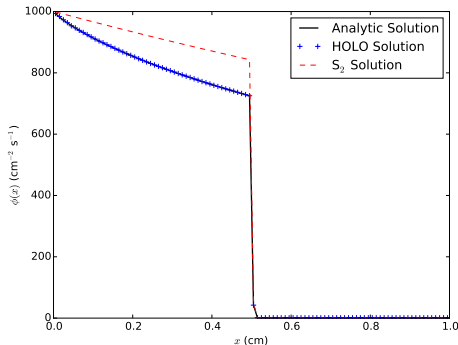
$$\mathbf{L}I^{(m+1)} = q + \mathbf{L} \left( \tilde{l}^{n+1,(m)} - \tilde{l}_{\text{pos}}^{n+1,(m)} \right)$$

If we apply  $\mathbf{L}^{-1}$  to both sides

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

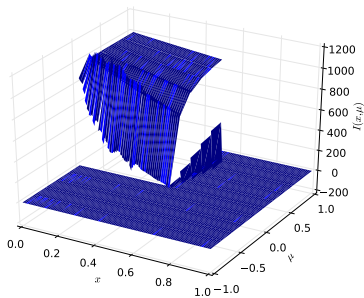
# Apply rotation and artificial source methods to steady state, pure-absorber neutronics problem

- ▶ Thin ( $\sigma_a = 0.2 \text{ cm}^{-1}$ ) and thick ( $\sigma_a = 1000 \text{ cm}^{-1}$ ) regions
- ▶  $q(x) = I_{\text{floor}} \sigma_a(x)$

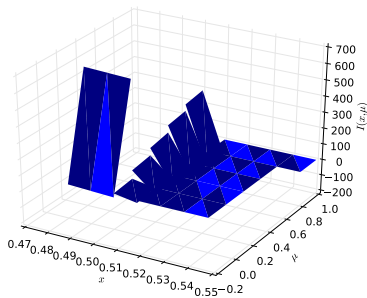


The angular intensity is difficult to resolve at the material interface and near  $\mu = 0$

$$\tilde{I}(x, \mu)$$



$$\tilde{I}(x, \mu) < I_{\text{floor}} \text{ (zoom)}$$



## Artificial source does not improve solution for analytic neutronics problem

- Compute cell-averaged error norms for HO and LO solutions, with 4 batches,  $10^5$  histories per batch

Fixup Method	$\ e_i^{LO}\ _{2,rel}$	$\ e_i^{HO}\ _{a,rel}$	FOM
Rotate every Batch	0.232%	0.265%	1.76
Rotate Last Batch	0.267%	0.302%	1.30
Artificial Source	0.261%	0.304%	1.34

\* $1\sigma \leq 0.01\%$  for all results

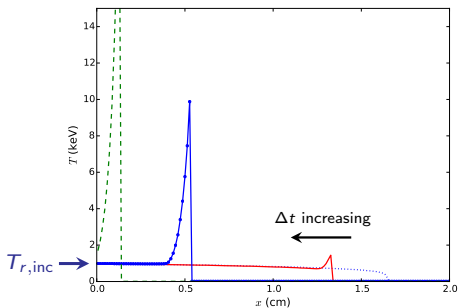
- The artificial source reduces magnitude of local residual, but can produce inaccurate solutions down stream
- Mixed results for TRT problems  
but all results have similar and high efficiency



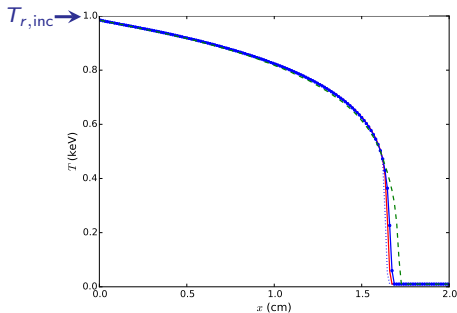
# Our HOLO method preserves the maximum principle with sufficient nonlinear convergence

- ▶ **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

IMC  $T_m$



HOLO  $T_m$



## DSA allows for efficient iterative solution of the low-order equations

Apply iterative solution methods to TRT two material problem with diffusivity of the thick region increased

Method	Avg. Sweeps/Newton Iter.
SI	1037
SI-DSA	10.9
GMRES	11.6
GMRES-DSA	6

\*25.1 **damped** Newton iterations per time step  
Scattering iteration relative tolerance  $10^{-10}$

# A HOLO Algorithm for Thermal Radiative Transfer

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Exponentially Convergent MC High-Order Solver

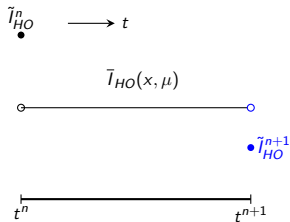
Derivation of the LO equations

Summary of algorithm

Computational Results

Monte Carlo time integration

The time variable is included in the ECMC trial space allowing for accurate time integration of HO intensity



Introduce trial space representation for  $I(x, \mu, t)$   
step doubly discontinuous in  $t$ , LDFE in  $x$  &  $\mu$

Include continuous  $\frac{1}{c} \frac{\partial}{\partial t} (\cdot)$  in  $\mathbf{L}$  for residual source  
leaving  $T(x)$  and  $T^4(x)$  implicit:

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

Sample and track particle histories in time.  
Tally time-averaged and  $t^{n+1}$  errors

The LO equations must be closed consistently by eliminating  $t^{n+1}$  unknowns with HO information

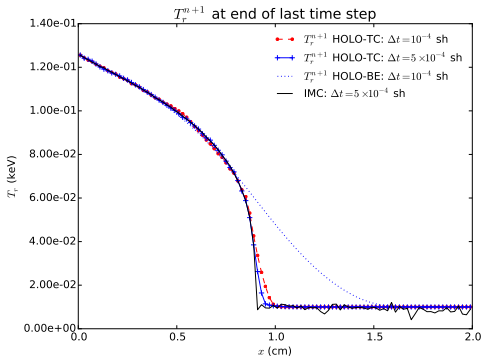
1. Form LO Equations with space-angle moments as before, but integrated over a time step  
Treat  $T(x)$  terms implicitly
2. Angular consistency terms are now time-averaged  
evaluated with time-averaged projection  $\bar{T}_{HO}(x, \mu)$
3. Eliminate space-angle moments of  $I_{LO}^{n+1}$  in terms of **time-averaged** moments  $\bar{I}_{LO}$ , e.g.,

$$\langle \phi \rangle_{L,i}^{n+1,+} = \gamma_{L,i}^{HO,+} \langle \bar{\phi} \rangle_{L,i}^+$$

4. Use closure to advance to next time step  
after solving for time-averaged unknowns

# The time closure preserves the accuracy of MC time integration in LO solution

- ▶ Material has  $\sigma_a = 0.2 \text{ cm}^{-1}$ , temperature mostly uncouples  
Plots depict  $T_r^{n+1}$  at  $t = 0.1 \text{ sh}$
- ▶ For HOLO w/ time closure (HOLO-TC)  
smaller time steps decrease noise but increase projection error
- ▶ HOLO Backward Euler (HOLO-BE) is inaccurate



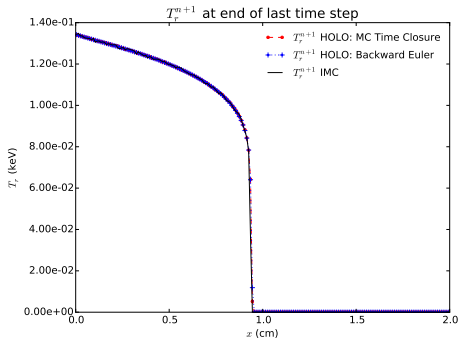
The HOLO-TC method is more efficient than IMC with sufficiently fine mesh

- Results for  $200 \times$  cells, HOLO-TC has  $60 \mu$  cells  
 $\Delta t = 0.001$  sh
- Error computed against IMC reference answer with  $4 \times 10^8$  histories/step,  $100 \times$  cells

hists./step	$\ \mathbf{e}_i\ _{2,\text{rel}}$		FOM	
	IMC	HOLO-TC (1)	IMC	HOLO-TC(1)
30,000	2.93%	14.00%	1	0.10
300,000	0.99%	0.37%	0.92	14.2
1,000,000	0.49%	0.18%	1.02	81.7

# The HO temporal closure is stable in a mix of optical thicknesses with sufficient histories

- ▶ Marshak wave problem,  $10^6$  hists/step over 2 batches



- ▶ Multiple batches are more efficient at estimating census
- ▶ HOLO-BE (FOM=1800) more efficient than HOLO-TC (FOM=15)  
step doubly-discontinuous trial space inaccurate



# A HOLO Algorithm for Thermal Radiative Transfer

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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

## Future work. . . very distant future

Need consistent way to resolve negativities  
strictly positive trial space for moments

Extend time treatment to a linear trial space  
sampling is significantly complicated

Improve efficiency of ECMC  
with asymmetric mesh refinement

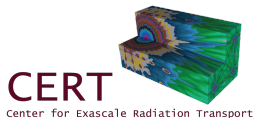
Consider Kernel Density Estimators for ECMC  
to improve efficiency within a batch

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Simon Bolding

Advisor: Jim Morel

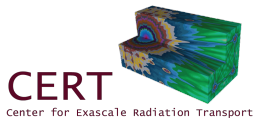
22 November 2016



# Backup Slides

Simon Bolding

Advisor: Jim Morel



# FOM and error norm definitions

Cell-averaged error norms

$$\|e_i\|_{rel}^{(l)} = \left( \frac{\sum_{i=1}^{N_c^{(l)}} \left( \phi_i^{n+1,(l)} - \phi_i^{n+1,ref} \right)^2}{\sum_{i=1}^{N_c^{(l)}} \left( \phi_i^{n+1,ref} \right)^2} \right)^{1/2}, \quad (1)$$

FOM is based on the following variance

$$\sigma(\phi_i)^2 = \frac{1}{N_{sim} - 1} \sum_{l=1}^{N_{sim}} \left( \overline{\phi_i} - \phi_i^{(l)} \right)^2, \quad (2)$$

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

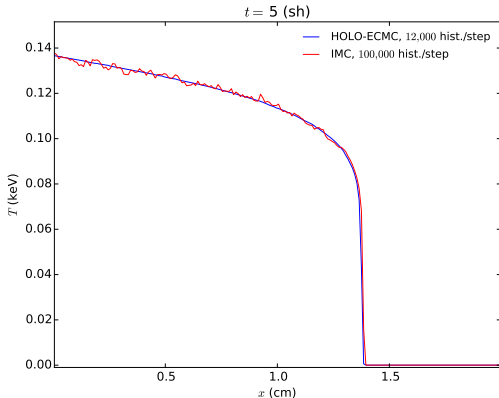
The system is *linearized* over a time step  $t \in [t^n, t^{n+1}]$   
Opacities are evaluated with  $T(t^n)$

- ▶ 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  $\Delta t$  for intensity

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

►  $\sigma_a \propto T^{-3}$

- Transient solution after 5 shakes ( $\sim 520$  steps)  
200 x cells (and 4  $\mu$  cells for ECMC)



## Time-integrated moment equation for $L$ , +

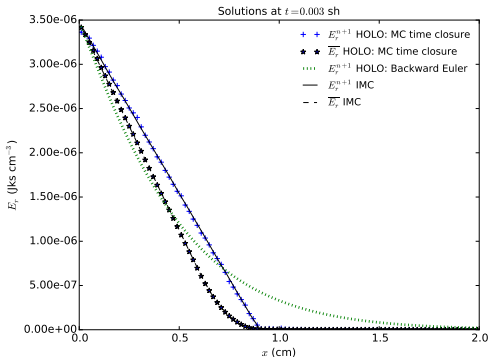
$$\begin{aligned}
 & \frac{\langle \phi \rangle_{L,i}^{+,n+1} - \langle \phi \rangle_{L,i}^{+,n}}{c\Delta t} - 2\bar{\mu}_{i-1/2}^+ \bar{\phi}_{i-1/2}^+ + \overline{\{\mu\}}_{L,i}^+ \langle \bar{\phi} \rangle_{L,i}^+ + \overline{\{\mu\}}_{R,i}^+ \langle \bar{\phi} \rangle_{R,i}^+ \\
 & + \sigma_{t,i}^{n+1} h_i \langle \bar{\phi} \rangle_{L,i}^{n+1,+} - \frac{\sigma_{s,i} h_i}{2} \left( \langle \bar{\phi} \rangle_{L,i}^+ + \langle \bar{\phi} \rangle_{L,i}^- \right) \\
 & = \frac{h_i}{2} \langle \sigma_a^{n+1} a c T^{n+1,4} \rangle_{L,i}, \quad (3)
 \end{aligned}$$



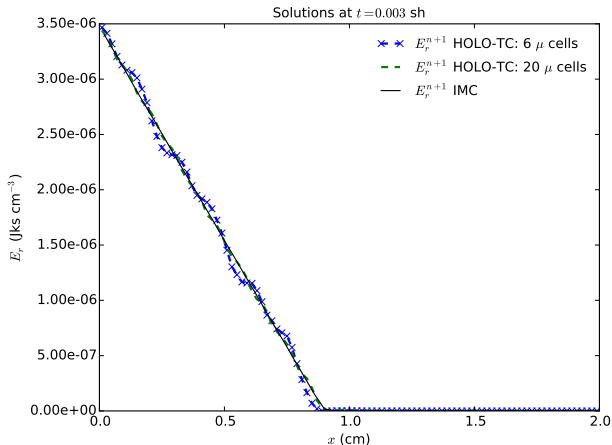
# The time-closure parameters preserve accuracy of MC time integration in the LO solution

Near-void problem,  $\sigma_a = 10^{-6} \text{ cm}^{-1}$   
take 3 **large** time steps

- Comparison radiation energy densities  $E_r = \phi(x)/c$  for **time-averaged** and **census** values
- 3 batches of 100,000 hists./step, 100 x cells,  
**FOM=0.53**

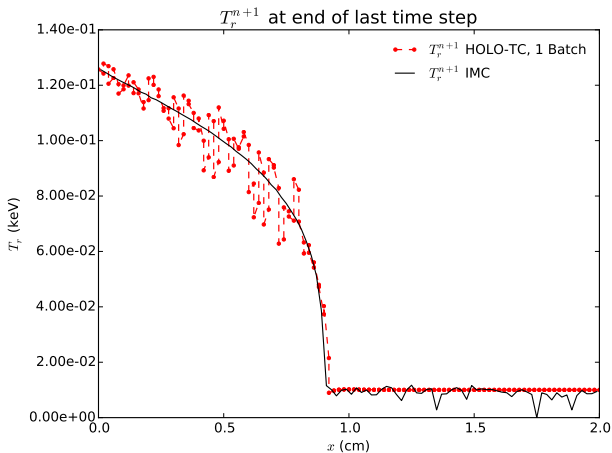


# Projection error for near-void problem

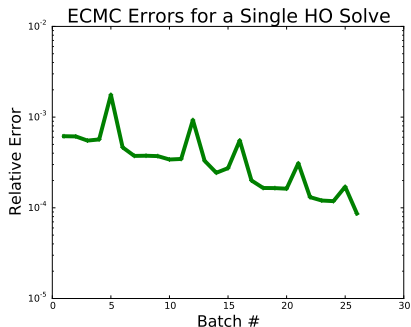
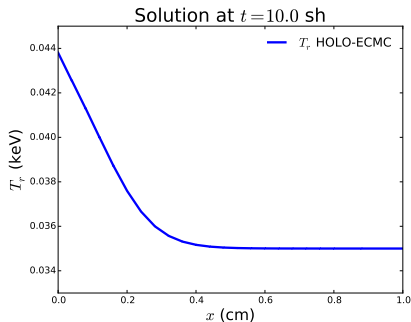


**Figure :** Comparison of radiation energy densities for the HOLO method with different numbers of  $\mu$  cells.  $\Delta t = 0.001$  sh, for near-void problem.

Without sufficient histories, time closure can introduce instabilities



Exponential convergence can be maintained if the LDFFE mesh resolves the solution reasonably



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

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

$$\begin{array}{lcl} \mathbf{L}\tilde{l}^{(m)} = q - r^{(m)} & \longrightarrow & \delta^{(m+1)} = \mathbf{L} \left( \tilde{l}^{(m)} - \tilde{l}_{pos}^{(m)} \right) \\ \mathbf{L}\tilde{l}_{pos}^{(m)} = q - r^{(m)} + \delta^{(m+1)} & & q \rightarrow q + \delta^{(m+1)} \end{array}$$

We can delay error stagnation

Investigating alternative positive projection of  $l$

# Tried importance sampling on the interior of the time step

Ensure that  $p_{surv}$  of particles sampled from interior are 2 mfp from census

$p_{surv}$	FOM
No Bias	1
0.05	0.001
0.1	0.005
0.25	0.179
0.5	0.003

# 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

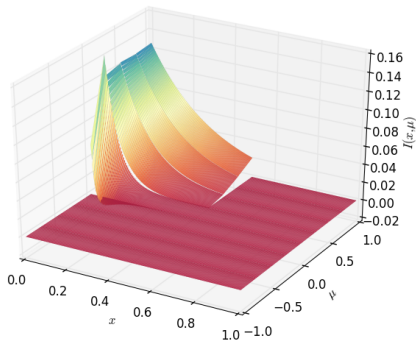
$$[\mathbf{D}(\mu^\pm) - \sigma_a^*(1 - f^*)] \underline{\Phi}^{n+1} = f^* \underline{B}(T^*) + \frac{\underline{\Phi}^n}{c\Delta t}$$

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

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

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

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





## Timing Results For Two Material Problem

hists./step	$\Delta t(sh)$	IMC ( $\mu s/hist.$ )	HOLO-ECMC ( $\mu s/hist$ )	Newt
100,000	0.001	17	3.5	
30,000	0.001	18	6.9	
30,000	0.005	59	7.4	

# Forming the LO System

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

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

- 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)