



Iterative Methods for Parallel Transport

Jae Chang

Nuclear Engineering

Texas A&M University

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Outline



- Transport Equation
- Solution Methods
- Parallel Implementations

The transport equation in operator notation leads naturally to
“source iteration” and “synthetic acceleration.”

$$\left[\vec{\Omega} \cdot \vec{\nabla} + \sigma_t(\vec{r}, E) \right] \psi(\vec{r}, E, \vec{\Omega}) = \int dE' \int d\Omega' \sigma_s(\vec{r}, E' \rightarrow E, \vec{\Omega}' \cdot \vec{\Omega}) \psi(\vec{r}, E', \vec{\Omega}') + q_{ext}(\vec{r}, E, \vec{\Omega})$$

Transport Equation: $L\psi = S\psi + q$

Source Iteration: $L\psi^{(l+1/2)} = S\psi^{(l)} + q$

Converged Solution: $\psi = \psi^{(l+1/2)} + (L - S)^{-1} S(\psi^{(l+1/2)} - \psi^{(l)})$

“Low-order” Operator: $M \approx (L - S)^{-1}$

“Synthetic” Iteration: $\psi^{(l+1)} = \psi^{(l+1/2)} + MS(\psi^{(l+1/2)} - \psi^{(l)})$

Different choices of low-order operator (M) give different synthetic methods.



- Diffusion synthetic acceleration (DSA)
 - Low-order operator is a diffusion approximation.
 - If discretized consistently, unconditionally convergent for all spatial meshes. (spectral radius < 1)
 - Effective for uniform meshes and homogeneous material properties (spectral radius $\ll 1$)
 - Efficiency of solving the diffusion approximation is a question.
- Transport synthetic acceleration (TSA)
 - Low-order operator is a simplified transport approximation.
 - Contains adjustable parameters:
 - Parameters determine spectral radius.
 - Parameters determine difficulty of low-order solution.
 - For many neutronics problems, parameters can be found to give effective and efficient scheme for uniform homogeneous problems.
 - Effectiveness degrades for optically thick problems.

TSA uses a simple transport operator to accelerate source iteration.

$$\vec{\Omega}_m \cdot \vec{\nabla} \psi_m^{(l+1/2)}(\vec{r}) + \sigma_t(\vec{r}) \psi_m^{(l+1/2)}(\vec{r}) = \frac{\sigma_s(\vec{r})}{4\pi} \phi^{(l)}(\vec{r}) + \frac{q(\vec{r})}{4\pi}$$

$$\phi^{(l+1/2)}(\vec{r}) = \sum_{m=1}^M w_m \psi_m^{(l+1/2)}(\vec{r})$$

$$\begin{aligned} \vec{\Omega}_n \cdot \vec{\nabla} f_n^{(k+1/2)}(\vec{r}) + [\sigma_t(\vec{r}) - \beta \sigma_s(\vec{r})] f_n^{(k+1/2)}(\vec{r}) \\ = \frac{\sigma_s(\vec{r})(1-\beta)}{4\pi} F^{(k)}(\vec{r}) + \frac{\sigma_s(\vec{r})}{4\pi} [\phi^{(l+1/2)}(\vec{r}) - \phi^{(l)}(\vec{r})], \end{aligned}$$

$$F^{(k+1)}(\vec{r}) = \sum_{n=1}^N w_n f_n^{(k+1/2)}(\vec{r})$$

$$\phi^{(l+1)}(\vec{r}) = \phi^{(l+1/2)}(\vec{r}) + F^{(k+1/2)}(\vec{r})$$

➤ Parameters:

- β
- N
- k

If we keep the diffusion length constant between high and low order problems, we have the stretched TSA.

$$\vec{\Omega}_m \cdot \vec{\nabla} \psi_m^{(l+1/2)}(\vec{r}) + \sigma_t(\vec{r}) \psi_m^{(l+1/2)}(\vec{r}) = \frac{\sigma_s(\vec{r})}{4\pi} \phi^{(l)}(\vec{r}) + \frac{q(\vec{r})}{4\pi}$$

$$\phi^{(l+1/2)}(\vec{r}) = \sum_{m=1}^M w_m \psi_m^{(l+1/2)}(\vec{r})$$

$$(\vec{\Omega}_n \cdot \vec{\nabla}_n) f_n^{(k+1/2)} + \frac{\sigma_t}{\varepsilon} f_n^{(k+1/2)} = \frac{\sigma_t - \varepsilon(\sigma_t - \sigma_s)}{4\pi} F^{(k+1/2)} + \frac{\varepsilon\sigma_s}{4\pi} (\phi^{(l+1/2)} - \phi^{(l)})$$

$$F^{(k+1/2)} = \sum_{n=1}^N w_n f_n^{(k+1/2)}$$

$$\phi^{(l+1)} = \phi^{(l+1/2)} + F^{(k+1)}$$

To damped the high frequency error modes, filtering steps were introduced.

$$\left(\vec{\Omega}_n \cdot \vec{\nabla}_n\right) f_n + \sigma_t \sqrt{1-c} f_n = \frac{\sigma_s}{4\pi\sqrt{1-c}} \left(\phi^{(l+1/2)} - \phi^{(l)}\right)$$

$$F = \sum_{n=1}^N w_n f$$

$$\pm\alpha \frac{\partial g_{x,\pm}}{\partial x} + \sigma_t g_{x,\pm} = \frac{\sigma_t}{2} F \quad G_x = g_{x,+} + g_{x,-}$$

$$\pm\alpha \frac{\partial g_{y,\pm}}{\partial y} + \sigma_t g_{y,\pm} = \frac{\sigma_t}{2} G_x \quad G_y = g_{y,+} + g_{y,-}$$

$$\phi^{(l+1)} = \phi^{(l+1/2)} + G_y$$

Wrapping a Krylov method around existing transport methods may improve convergence.



- Finds the “best” linear combination of previous iterates.
- Use DSA or TSA as a preconditioner to a Krylov iterative solvers.
- Easy to implement over existing transport codes.
- Conjugate gradient, BiCG, or GMRES.

Implementing transport sweeps in parallel.



- Problem decomposition
 - Angle
 - Space
 - Energy
- Communication
- Synchronization
- Scheduler
- Partitioner

GMRES in parallel



- Parallel implementation of inner products and norms
- Global reduce and broadcast
- DSA uses CG to solve the diffusion problem (low-order problem).

We use pArrays to implement GMRES.



- Use GMRES to solve the within groupset problem.
- pArray – parallel array
 - STAPL container
 - Built in parallel functions and operators, e.g. dot products and norms
 - Distribution of data handled automatically
- Independent of spatial discretization
 - Source iteration
 - TSA