Krylov Linear Solvers and Quasi Monte Carlo Methods for Transport Simulations

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Outline



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- Model Problem and QMC Model Problem Formulation as Linear System QMC overview
- 3. Preliminary Results
- 4. Conclusions
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Our Place in Project Roadmap



Model Problem



$$\mu \frac{\partial \psi}{\partial x}(x,\mu) + \Sigma_t(x)\psi(x,\mu) = \frac{1}{2} \left[\Sigma_s(x) \int_{-1}^1 \psi(x,\mu') d\mu' + q(x) \right]$$
$$\equiv \frac{1}{2} (\phi(x) + q(x)),$$

where $0 \le x \le \tau$ and

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

Subject to boundary conditions

$$\psi(0,\mu) = \psi_I(\mu), \mu > 0; \psi(\tau,\mu) = \psi_r(\mu), \mu < 0.$$

Source Iteration



Given an iteration n and $\phi_n(x)$ Compute ϕ_{n+1} by solving the fixed source problem

$$\mu \frac{\partial \psi_{n+1}}{\partial x}(x,\mu) + \Sigma_t(x)\psi(x,\mu) = \frac{1}{2} \left[\Sigma_s(x)\phi_n(x) + q(x) \right],$$

and then compute

$$\phi_{n+1}(x) = \int_{-1}^1 \psi_{n+1}(x,\mu') d\mu' \equiv \mathcal{S}(\phi_n,q,\psi_l,\psi_r)$$

Linear System



Source iteration is Picard iteration for the fixed point problem

$$\phi = \mathcal{S}(\phi, \mathbf{q}, \psi_I, \psi_r)$$

To use other solvers we must convert to a linear system via

$$\mathcal{K}(\phi) = \mathcal{S}(\phi, 0, 0, 0)$$
 and $f = \mathcal{S}(0, q, \psi_I, \psi_r)$

to get

$$A\phi \equiv (I - \mathcal{K})\phi = f,$$

which we can send to a linear solver.

Solving the problem



- Evaluating the Source iteration map
 - Conventional deterministic methods (SN . . .)
 - Monte Carlo
 - Quasi Monte Carlo (this talk)
- Solving the linear problem
 - Source (Picard) iteration (in most codes)
 - Krylov methods: matrix-free
 - needs **CONSISTENT** evaluation of ${\cal S}$ to get $A\phi$ MC has trouble with this
 - but no matrix assembly

Overview of Krylov Methods for Ax = b

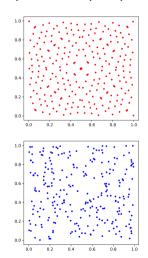


- Initialization
 - x₀: initial iterate
 - $\mathbf{r}_0 = \mathbf{b}_0 \mathbf{A}\mathbf{x}_0$: initial residual
- Find iteration \mathbf{x}_k in $\mathcal{K}_k = \text{span}(\mathbf{r}_0, \mathbf{A}\mathbf{r}_0, \dots, \mathbf{A}^{k-1}\mathbf{r}_0)$
- Example GMRES (Sadd/Schultz 1986): \mathbf{x}_k minimizes $\|\mathbf{b} \mathbf{A}\mathbf{x}\|$ over \mathcal{K}_k
- All is well if matrix-vector product is consistent.
 - Ax returns the same result with every call.
 - MC is not deterministic and hence not consistent.

QMC vs MC



QMC (red) and MC (blue)



A few facts

- MC and QMC are continuous in space/angle/energy
- With N samples/particles
 - QMC error *O*(1/*N*)
 - MC variance $O(1/\sqrt{N})$
- MC randomness
 - can break Krylov solvers
 - Willert et al 2013, 2014, 2015
 - Simoncini et al, 2003
- QMC gives consistent matrix-vector products

Preliminary Results



- All codes in Julia on github https://github.com/ctkelley/NDA_QMC.jl
- Model problem from Garcia/Siewert 1982

$$\tau = 5, \Sigma_s(x) = \omega_0 e^{-x/s}, \Sigma_t(x) = 1, q(x) = 0, \psi_I(\mu) = 1, \psi_r(\mu) = 0.$$

- ullet Two cases: s=1 (easy) and $s=\infty$ (hard)
- Evaluate source iteration map with QMC
- Compare GMRES and Bi-CGSTAB (van der Vorst 1992) to Picard
- Validation with with Garcia/Siewert exit distributions
 - QMC: N samples, Nx cells
 - Two figure accuracy with N = 1000 and Nx = 100

Validation: Easy case, N = 1000, Nx = 100



	Garcia/Siewert		QMC	
μ	$\psi(0,-\mu)$	$\psi(au, -\mu)$	$\psi(0,-\mu)$	$\psi(\tau, -\mu)$
5.00e-02	5.89664e-01	6.07488e-06	5.71197e-01	5.85487e-06
1.00e-01	5.31120e-01	6.92516e-06	5.22137e-01	6.66741e-06
2.00e-01	4.43280e-01	9.64232e-06	4.41567e-01	9.25261e-06
3.00e-01	3.80306e-01	1.62339e-05	3.81029e-01	1.54416e-05
4.00e-01	3.32964e-01	4.38580e-05	3.34673e-01	4.09691e-05
5.00e-01	2.96090e-01	1.69372e-04	2.98224e-01	1.57373e-04
6.00e-01	2.66563e-01	5.73465e-04	2.68871e-01	5.35989e-04
7.00e-01	2.42390e-01	1.51282e-03	2.44749e-01	1.42448e-03
8.00e-01	2.22235e-01	3.24369e-03	2.24583e-01	3.07431e-03
9.00e-01	2.05174e-01	5.96036e-03	2.07478e-01	5.67991e-03
1.00e+00	1.90546e-01	9.77123e-03	1.92789e-01	9.35351e-03

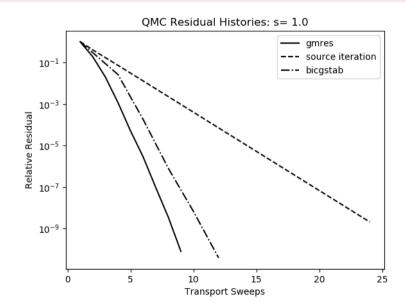
Validation: Hard case, N = 1000, Nx = 100



	Garcia/Siewert		QMC	
μ	$\psi(0,-\mu)$	$\psi(au, -\mu)$	$\psi(0,-\mu)$	$\psi(\tau, -\mu)$
5.00e-02	8.97798e-01	1.02202e-01	8.47454e-01	1.00663e-01
1.00e-01	8.87836e-01	1.12164e-01	8.52822e-01	1.10325e-01
2.00e-01	8.69581e-01	1.30419e-01	8.47710e-01	1.29064e-01
3.00e-01	8.52299e-01	1.47701e-01	8.35879e-01	1.46849e-01
4.00e-01	8.35503e-01	1.64497e-01	8.22291e-01	1.64034e-01
5.00e-01	8.18996e-01	1.81004e-01	8.08044e-01	1.80827e-01
6.00e-01	8.02676e-01	1.97324e-01	7.93459e-01	1.97336e-01
7.00e-01	7.86493e-01	2.13507e-01	7.78672e-01	2.13625e-01
8.00e-01	7.70429e-01	2.29571e-01	7.63768e-01	2.29725e-01
9.00e-01	7.54496e-01	2.45504e-01	7.48818e-01	2.45642e-01
1.00e+00	7.38721e-01	2.61279e-01	7.33889e-01	2.61361e-01

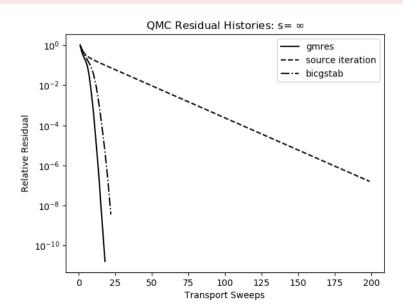
Convergence: Easy Case





Convergence: Hard Case





Conclusions



- QMC for neutron transport simulations
 - Completely deterministic
 - Avoids problems with MC and Krylov methods
 - Krylov solvers very promising
- Test on benchmark problem
- Next up: NDA

Links



- Jupyter notebook draft of paper: https://github.com/ctkelley/NDA_QMC.jl
- Julia package: https://github.com/ctkelley/SIAMFANLEquations.jl
- Jupyter notebook draft of solver book: https://github.com/ctkelley/NotebookSIAMFANL

References I: Background for this talk



- SAMUEL PASMANN, ILHAM VARIANSYAH, AND RYAN G. McClarren, Convergent Transport Source Iteration Calculations with Quasi-Monte Carlo, in 2021 ANS Virtual Annual Meeting, 2021, pp. 192–195.
- R.D.M. GARCIA AND C.E. SIEWERT, Radiative transfer in finite inhomogeneous plane-parallel atmospheres, J. Quant. Spectrosc. Radiat. Transfer, 27 (1982), pp. 141–148.
- C. T. Kelley, Solving Nonlinear Equations with Iterative Methods: Solvers and Examples in Julia, 2020. Unpublished book ms, under contract with SIAM.
- C. T. Kelley, Notebook for Solving Nonlinear Equations with Iterative Methods: Solvers and Examples in Julia. https://github.com/ctkelley/NotebookSIAMFANL, 2020.

References II: MC + Krylov problems



- J. WILLERT, XIAOJUN CHEN, AND C. T. KELLEY, Newton's method for Monte Carlo-based residuals, SIAM J. Numer. Anal., 53 (2015), pp. 1738–1757.
- J. WILLERT, C. T. KELLEY, D. A. KNOLL, AND H. K. PARK, *Hybrid deterministic/Monte Carlo neutronics*, SIAM J. Sci. Comp., 35 (2013), pp. S62–S83.
- J. WILLERT, C. T. KELLEY, D. A. KNOLL, AND H. K. PARK, A hybrid deterministic/Monte Carlo method for solving the k-eigenvalue problem with a comparison to analog Monte Carlo solutions, J. Comp. Th. Transport, 43 (2014), pp. 50–67.
- V. SIMONCINI AND D. B. SZYLD, Theory of inexact Krylov subspace methods and applications to scientific computing, SIAM J. Sci. Comput., 25 (2003), pp. 454–477.

References III: Solvers



- C. T. Kelley, *SIAMFANLEquations.jl.* https://github.com/ctkelley/SIAMFANLEquations.jl, 2020. Julia Package.
- C. T. Kelley, *Iterative Methods for Linear and Nonlinear Equations*, no. 16 in Frontiers in Applied Mathematics, SIAM, Philadelphia, 1995.
- C. T. Kelley, Solving Nonlinear Equations with Newton's Method, no. 1 in Fundamentals of Algorithms, SIAM, Philadelphia, 2003.
- Y. SAAD AND M.H. SCHULTZ, GMRES a generalized minimal residual algorithm for solving nonsymmetric linear systems, SIAM J. Sci. Stat. Comp., 7 (1986), pp. 856–869.
- H. A. VAN DER VORST, Bi-CGSTAB: A fast and smoothly converging variant to Bi-CG for the solution of nonsymmetric systems, SIAM J. Sci. Statist. Comput., 13 (1992), pp. 631–644.