

Magnitude of Bias in Monte Carlo Eigenvalue Calculations

Prof. H. Bowsher*, E. M. Gelbard, P. Gemmel**, G. Rack†

Most Monte Carlo eigenvalue calculations are based on power iteration methods, like those used in analytical algorithms. But if N_H , the number of histories in each generation is fixed, then such Monte Carlo calculations will be biased.^(1,2) Various arguments lead to the conclusion that eigenvalue and shape biases are both proportional to $1/N_H$, but little more is known about their magnitudes. Numerical experiments on simple matrices suggest that the biases are small, but information more relevant to real reactor calculations is very sparse.

In fact to determine the bias in real reactor calculations is quite expensive. It seems worthwhile, therefore, to try to understand the Monte Carlo biases in systems more realistic than arbitrary matrices, but simpler than real reactors. For this reason we have computed biases in simple one-group model problems by the following procedure.

- a. Each problem configuration is split into I subzones, and one calculates the zone-to-zone collision probabilities p_{ij}^c .
- b. One then computes, deterministically, the probability, p_{ij}^a , that a neutron formed in subzone j will be absorbed in subzone i.
- c. From p_{ij}^c , and the problem cross sections one gets the leading eigenvalue λ^D , and the corresponding eigenvector, ϕ_D , also deterministically.
- d. Define the source vector s_D , with components $s_D^i \equiv v \sum_j \phi_D^j / \sum_j \phi_D^j$, (where the superscripts are subzone numbers) and choose N_H neutron birth-sites from this source.

*Dept. of Chemistry & Physics, August College, Augusta, GA 30910

**4061 Linden Avenue, Western Springs, IL 60558

†1800 N. Noland Avenue, Independence, MO 64050

NOTICE

PORTIONS OF THIS REPORT ARE RELEASABLE.

It has been reproduced from the best available copy to permit the broadest possible availability.


 DISTRIBUTION OF THIS DOCUMENT IS UNLIMITED

MASTER

c. For each birth-site, j , choose an absorption site from the p.d.f. p_{ij}^a .

Accumulate $v\Sigma_f^i$ in the i 'th fission-source bin. After all N_H starters have been processed the total fission rate, summed over all source bins, is the first-generation eigenvalue, $\lambda^{(1)}$, and the fission rate per bin, divided by $\lambda^{(1)}$, is the p.d.f. for source sites for the second generation, etc.

In vection notation

$$1) \quad \Sigma_t \phi = \underline{P}^c [\Sigma_s + 1/\lambda \underline{v\Sigma_f}] \phi.$$

Here Σ_t , Σ_s and $\underline{v\Sigma_f}$ are diagonal matrices whose elements are, respectively,

Σ_t^i , Σ_s^i and $v\Sigma_f^i$. Further \underline{P}^c is the matrix of collision probabilities and ϕ the

flux vector. It is easy to show, from Eq. (1), that $\lambda \underline{a} = \underline{P}^a \underline{f}$, where

$$2) \quad \underline{P}^a \equiv \Sigma_a (\Sigma_t^c - \underline{P} \Sigma_s)^{-1} \underline{P},$$

\underline{f} is the fission source, \underline{P}^a the matrix of absorption probabilities and \underline{a} the absorption rate. We use Eq. (2) to get the p_{ij}^a 's from the collision probabilities, p_{ij}^c .

Test problems have been run in various model problem configurations, with $N_H = 6$ and 12. Such low values of N_H were used to make the biases more easily detectable. For all test problems (a) in the core $\Sigma_s = 0.9$, $\Sigma_a = 0.1$, $v\Sigma_f = 0.1$, and (b) in the reflector $\Sigma_s = 1.98$, $\Sigma_a = 0.02$, $v\Sigma_f = 0$, all in inverse cm.

In test problems one through four, slab problems without control rods, there are, respectively 10, 20, 30 and 60 subzones in the core, all 1 cm wide.

All slab configurations have identical reflectors composed of six subzones. Of these the first subzone, nearest the core, is 1 cm wide, while all others are 3 cm wide.

Results for these configurations (and for other tests discussed below) are shown in Table I. The listed biases do seem to be proportional to $1/N_H$, though statistical uncertainties still leave some room for doubt. If biases are proportional to $1/N_H$, then eigenvalue biases would be negligible even for $N_H = 50$. A higher $N_H (\approx 200)$ would be required to reduce the shape-bias in the core to $\approx 5\%$, but $N_H = 200$ is still near the lower bound of normal batch sizes.

Similar slab computations also have been run with absorbing subzones used to simulate control rods. Biases seem to be reduced by the presence of such "rods" apparently because they increase dominance ratios.

Three X Y computations have been run in configurations shown in Fig. 1. These configurations differed only in the sizes of the core subzones which were, respectively, one, two and three cm squares in problems 5, 6 and 7. One sees, from, Table I, that the biases again seem to be proportional to $1/N_H$, and would be acceptable for $N_H \gtrsim 100$.

It seems, then, that it is permissible to use fairly small batch sizes in Monte Carlo eigenvalue calculations. Whether this is advantageous or not may depend on possible effects of N_H on the convergence rate. Preliminary results suggest, however, that the convergence rate does not depend on N_H .

References

1. R. C. Gast and N. R. Candelore, (W-BAPL, USA) "Monte Carlo Eigenvalue Strategies and Uncertainties".
2. E. M. Gelbard and R. E. Prael (ANL, USA), Monte Carlo Work at Argonne National Laboratory".

TABLE I.

Monte Carlo Calculations Based on 960,000 Histories								
Problem Number	Deterministic Calculation		Percent Eigenvalue Bias		Percent Shape Bias			
					Core Center		Outer Edge or Corner	
			Histories/ Generation	Histories/ Generation	Histories/ Generation	Histories/ Generation	Histories/ Generation	Histories/ Generation
	λ_0	λ_1	6	12	6	12	6	12
1	0.95135	0.64940	-0.276 ± 0.030	-0.129 ± 0.030	-4.2 ± 0.5	-2.3 ± 0.4	8.4 ± 1.3	3.8 ± 0.4
2	0.98329	0.86236	-0.417 ± 0.029	-0.199 ± 0.029	-12.2 ± 2.3	-5.9 ± 1.1	31.0 ± 2.1	13.5 ± 2.1
3	0.99163	0.92830	-0.406 ± 0.029	-0.212 ± 0.028	-17.69 ± 1.6	-9.2 ± 1.8	65.4 ± 20.4	26.5 ± 4.2
4	0.99762	0.97897	-0.385 ± 0.029	-0.218 ± 0.028	-25.1 ± 3.8	-18.9 ± 5.0	156.3 ± 12.5	90.0 ± 14.3
5	0.90749	0.62580	-0.540 ± 0.049	-0.247 ± 0.038	-8.1 ± 1.4	-4.9 ± 1.4	13.7 ± 2.1	6.2 ± 2.4
6	0.96008	0.81312	-0.662 ± 0.037	-0.343 ± 0.037	-16.7 ± 1.5	-10.5 ± 1.6	36.6 ± 3.3	19.3 ± 3.9
7	0.97447	0.87508	-0.602 ± 0.032	-0.323 ± 0.042	-21.2 ± 1.7	-14.4 ± 1.9	57.1 ± 5.2	35.2 ± 6.1

Eigenvalue and fission-source-shape biases in simple one-group test problems.

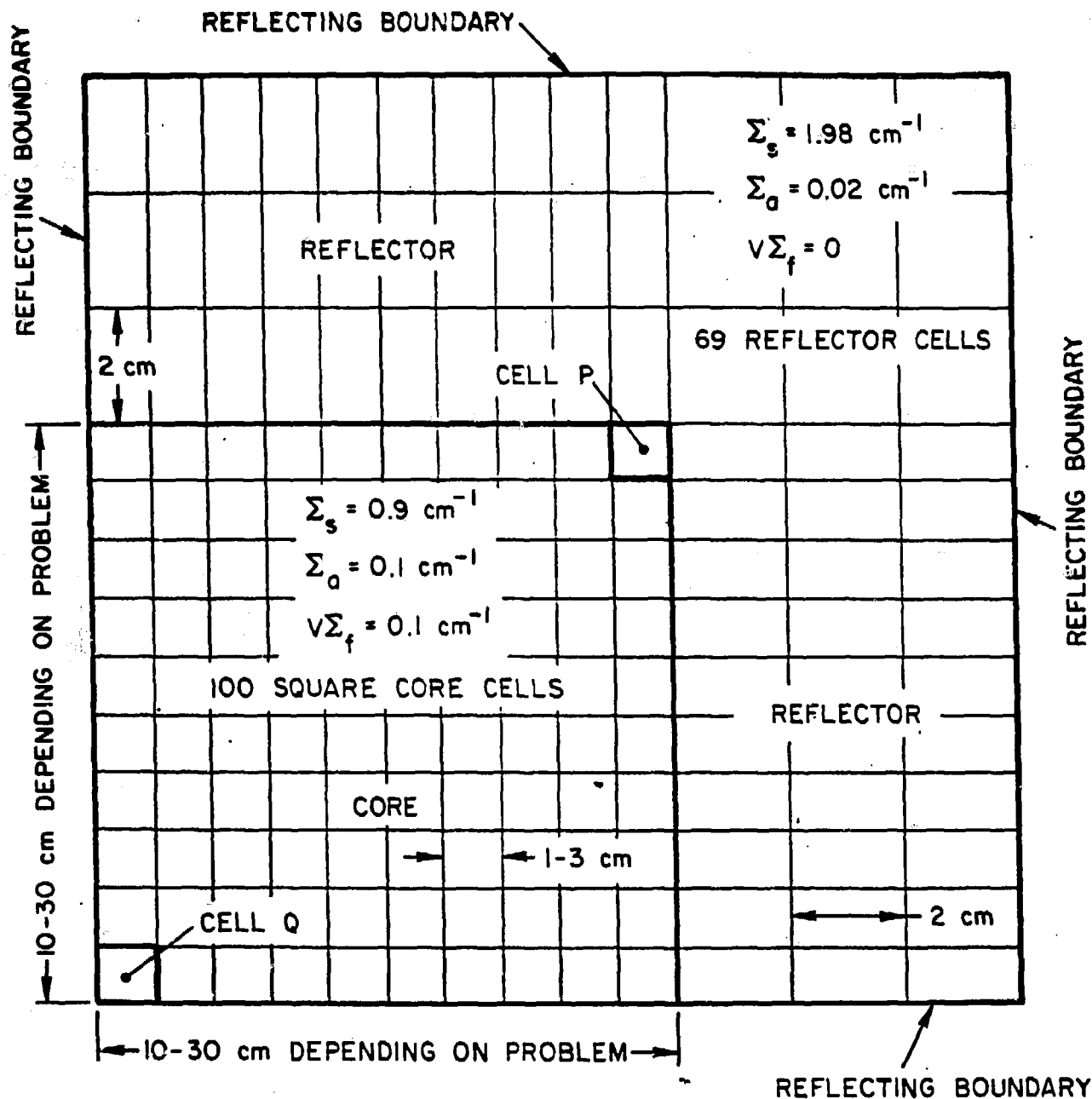


Fig. 1. Two-Dimensional Test Problem Configurations.

DISCLAIMER

This report was prepared as an account of work sponsored by an agency of the United States Government. Neither the United States Government nor any agency thereof, nor any of their employees, makes any warranty, express or implied, or assumes any legal liability or responsibility for the accuracy, completeness, or usefulness of any information, apparatus, product, or process disclosed, or represents that its use would not infringe privately owned rights. Reference herein to any specific commercial product, process, or service by trade name, trademark, manufacturer, or otherwise does not necessarily constitute or imply its endorsement, recommendation, or favoring by the United States Government or any agency thereof. The views and opinions of authors expressed herein do not necessarily state or reflect those of the United States Government or any agency thereof.