

# Monte Carlo Statistical Convergence

by

Dermott E. Cullen

1466 Hudson Way

Livermore, CA 94550

Tele: 925-321-4177

E.Mail: [RedCullen1@comcast.net](mailto:RedCullen1@comcast.net)

Website: <http://redcullen1.net/homepage.new>

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## Abstract

This document reviews Monte Carlo Statistical Convergence over the last 60 years, based on the real World state of computers over this time. Stress is placed on providing examples of what was and was not practically possible over this period of time. I present a few examples to illustrate convergence based not only on sample size, but also how cross sections can be modelled.

## Overview

In principle Monte Carlo has always been the most accurate method to model nuclear and atomic systems; most accurate to model in the sense of geometry and physical data. In practice it has always been limited by the time required to accurately calculate results on the computer facilities that we have had at any given time. Here I present a few real World example results based on using the TART Monte Carlo code [1].

## TART

The TART Monte Carlo code is designed both as a production code, and as a teaching tool. TART has many input options, and these have been used over many years to optimize results and simplify use of the code today. All input options have default values, and to use the code for production users need not change/define additional options. Here I will use TART as a teaching tool, to illustrate how sample size and nuclear data models affect running time and calculated results. I would encourage users/developers of all Monte Carlo codes to present similar results for the benefit of our entire community.

## Acknowledgments

I thank Dave Heinrichs (LLNL) for supporting the publication of this report and agreeing to distribute TART 2022 within LLNL. Next I acknowledge the cooperation with other neutron transport code designer/users, in particular, Bob MacFarlane (LANL, MCNP), Rich Procassini (LLNL, MERCURY), Ed Lent (LLNL, COG) and Maurice Greene (ORNL); coordinated code comparisons have led to improvements in TART as well as ALL of these other codes. **I cannot stress enough that modern transport codes are far too complex to allow true code verification without detailed comparisons to other similar codes. Only in our dreams may we be perfect.**

I also thank the many users of earlier versions of TART who have supplied extremely useful feedback to me. Since the general release of TART2005, TART2012 and TART2016, the response from users in terms of feedback has been extremely useful in improving the code. These improvements have been in terms of correcting problems in the earlier releases of TART, and in terms of users proposing new or improved options to meet their needs, now incorporated in TART 2022. **Today virtually ALL improvements to TART are based on feedback to me from TART users. Therefore, I highly encourage all users to supply their feedback to me, so that ALL TART users can benefit from your experience, including yourself by including your ideas in future versions if TART.**

## Convergence

Let me first define what I will use as my definition of convergence in this paper. Many publications claim that with today's codes and nuclear data we can calculate K-eff for critical systems to within three digits, e.g., K-eff  $\sim 1.000 \pm 0.001$ ; 0.1 %. Personally, I think this is far too optimistic, but in this paper I will use it as a measure of convergence. I will FLAG any differences in K-eff that we calculate using the various approximations used in this paper, as a simple means of indicating a result we should be concerned by.

In this paper I have presented a few example results to illustrate both the advantages and disadvantages of Monte Carlo. One obvious advantage is its ability to model geometry more precisely, in more detail. A second advantage is its ability to define nuclear data cross sections more precisely, on a continuous energy basis. On the other hand, an obvious disadvantage is that the estimated solutions slowly converge to accurate answers. For example, convergence can vary as  $1/\sqrt{\text{samples}}$ , so that to improve accuracy by a factor of 10 requires an increase in samples, and therefore running time, by a factor of 100. Even with the computer power that we have today, it may not be practical to increase sample size, and therefore running time, by a factor of 100. More to the point, it certainly wasn't practical many years ago with the computers we then had; I will briefly address this point.

A controversial point is that convergence speed depends on the number of degrees of freedom in each problem. Using continuous energy cross sections admittedly improves the detail to which nuclear data can be represented, but it can vastly increase the number of degrees of freedom in a problem. For example, in a multi-group calculation, the number of total cross sections to sample is equal to the number of groups, whereas using continuous energy cross sections there can be many thousands of tabulated cross sections, and using even simple linearly interpolation between tabulated values introduces vastly more degrees of freedom.

Here I present example results comparing continuous energy, multi-band, and multi-group results (both multi-band and multi-group use the same TART 616 energy groups). These results illustrate that even with 616 groups and a million neutron criticality samples, multi-group results differ significantly from results based on using continuous energy cross sections. We see this in particular for metal reflected critical assemblies; presumably due to neutron leakage through minima in the continuous energy cross sections, that are not included in the multi-group cross sections. However, even simple multi-band (2-band) results significantly improve agreement. This latter point should be of interest to users/developers of multi-group transport codes, since multi-band cross sections can be used in multi-group codes, but multi-group codes cannot easily be converted to use continuous energy data; the energy-to-energy transfer matrices become singular, and do not converge, as the number of groups are increased.

## History 101

I wrote my first FORTRAN (FORmula TRANslation) code in 1962. At the time FORTRAN was in its infancy, having been created only slightly earlier in 1957. There was no number after the word FORTRAN (e.g., FORTRAN-77), and there were no subroutines, and therefore no common ["no common" what?]; still, this was a great leap forward in making computers accessible to anyone.

At this time not only was the language new/primitive, so were the computers. The first computer I used was I think/remember as the largest generally available computer at the time, namely the IBM-7090/94, which had 32,000, 36 bit words of memory, and a cycle time of 2 microseconds. This computer cost a minimum of \$2.9 million. Storage was also extremely limited, with a 20 megabyte disk being the size of a washing machine; our main storage device was magnetic tape.

At the time our transport codes limited were to a few group diffusion calculations, and we saw the beginnings of discrete ordinates ( $S_n$ ) at major laboratories. This was a primitive beginning, but the situation changed very rapidly, both due to advances in computers and the codes available to use them and I must mention nuclear data (through the ENDF effort).

By the time I started my first post-graduate position in 1967 at Brookhaven National Laboratory (BNL) an effort was just beginning to improve our nuclear data (ENDF) (without accurate nuclear data we were in a "garbage in = garbage out" situation) , and by the time I started by second position in 1972 at Lawrence Livermore National Laboratory work had begun on Monte Carlo codes, MCNP at Los Alamos, and TART at Livermore, two codes that over the years have maintained close cooperation and development (fortunately, early on it was recognized that our particle transport codes are far too complicated to believe/rely on results without code comparisons).

In the relatively short period of time between 1962 and 1972 computers had advanced from the IBM-7090/94 to CRAY computers; improving both speed and memory/storage, allowing us to consider ever more complicated problems and more detail. Since then, we have undergone a revolution in our computer capabilities., as illustrated by the below tables.

## Running Time

The below table presents results obtained using a collection of 68 TART benchmark criticality problems. All 68 problems were run on each computer using only one processor, all with exactly the same run parameters, such as number of samples, and nuclear data models. This table summarizes timing results for the older TARTND code that only runs on CRAY computers, as well as all released versions of TART, on a variety of computers. The original TARTND running times from many years ago are included to illustrate how far we have come; ratios are normalized to these original CRAY-YMP running times.

Code	Computer	Running Time (Seconds)	Ratio to TARTNP CRAY-YMP	
TARTNP	CRAY-YMP	5396	1.0	1995 Beginning
TARTNP	CRAY-J90	7727	1.43	
<b>TART2022</b>	<b>XPS 15 7590/64bits/Windows10</b>	<b>14</b>	<b>0.0025</b>	<b>Today</b>
TART2016	XPS 9100/64bits/Windows10	19	0.0035	2016
TART2016	Inspiron 5759/64bits/Windows10	23	0.0043	
TART2005	AMD 3500+	47	0.0087	2005
TART2005	AMD 3400+	48	0.0089	
TART2005	IBM-PC Pentium IV/3600	58	0.0107	
TART 2002	Athlon XP1800/1520	89	0.0165	2002
TART 2002	IBM-PC Pentium IV/2000	132	0.025	
TART 2002	IBM-PC Lap Top III/1200	133	0.025	
TART 2002	IBM-PC Pentium III/1000	170	0.031	
TART 2002	IBM-PC Pentium III/500	500	0.09	
TART 2002	DEC-Alpha Model 5/625	516	0.10	
TART 2002	IBM-PC Pentium II/400	579	0.11	
TART 2002	PowerMAC/LapTop/500	683	0.126	
TART 2002	IBM-PC Pentium II/333	697	0.13	
TART 2002	DEC-Alpha Model 5/300	712	0.13	
TART 2002	IBM-PC Pentium II/266	855	0.16	
TART 2002	IBM-PC Pentium Pro/200	1185	0.22	
TART 2002	IBM-PC Lap Top/233	1301	0.24	
TART 2002	Power-MAC 7500/275	1350	0.25	
TART 2002	iMAC	1664	0.31	
TART 2002	HP-735/125	1834	0.34	
TART 2002	SUN E3000/166	2107	0.39	
TART 2002	IBM-PC LapTop/133	2990	0.58	
TART 2002	CRAY-YMP	4262	0.79	
TART 2002	IBM-RISC RS-6000	5739	1.06	
TART 2002	CRAY-J90	6095	1.13	
TART 2002	Meiko CS-2/66	6225	1.15	
TART95	CRAY-YMP	4912	0.91	
TART95	HP-350	4322	0.80	
TART95	DEC-Alpha	6130	1.14	
TART95	SUN	9673	1.79	

The latest TART2022 results shown above illustrate that today my humble \$2000 laptop can run the 68 criticality problems in 14 seconds; an incredible **400 times faster** than TARTNP could run these same problems on a multi-million dollar CRAY-YMP. For the most up-to-date list of running times see, <http://redcullen1.net/homepage.new/speed.htm>

It is worth noting that the above results are only for the last 28 years since TART's first public release outside LLNL. I cannot reliably guess how these times would compare to trying to run a code like TART 50 years ago on an IBM-7090/94 – but a wild guess would be that today we can run the same problem 10,000 times faster.

## Monte Carlo's Statistical Problems

Even with this incredible increase in computer power coupled with improved Monte Carlo codes and nuclear data, Monte Carlo with all of its advantages has one obvious problem: the results depend on how many histories are run, and the fact that convergence to the ultimate answer varies very slowly with the number of histories run. Basically, convergence varies as  $1/\sqrt{\text{samples}}$ , e.g., to increase the accuracy by a factor of ten requires that one hundred times as many samples be run. The below table uses the TART 68 fast critical assemblies to illustrate this point using 4 results, for  $10^8$ ,  $10^7$ ,  $10^6$ ,  $10^5$  neutron convergence samples.

For a criticality calculation TART uses two distinct steps. The first step is **settle cycles**, where we initially “guess” the answer by defining a starting neutron flux distribution in space, energy, time. TART starts from this initial guess and iterates neutron generation by generation assuming that highly level modes will dampen out and our “guess” will evolve into the correct final lowest mode flux distribution. The second step is the actual **criticality calculation**, where we start all over from our last generation of our settled flux and iterate generation by generation, defining the effective multiple (K-eff) as the ratio of neutrons produced in one generation compared to the number in the preceding generation, i.e., how fast the flux is multiplying per generation. With TART input the users controls,

Batch Size (sentl 3) = the number of neutron samples run in each batch (generation).

Settle Cycles (critcalc) = the number of **settle** cycles (generations) to run.

Critical Cycles (sentl 2) = the number of **criticality** cycles (generations) to run.

Settle Cycles X Batch Size = number of neutron samples in **settle** calculation.

Critical Cycles X Batch Size = number of neutron samples in **criticality** calculation.

For the TART 68 fast critical assemblies the below table defines how each TART parameter was defined. Here is a very brief explanation. Experience has shown that the settle cycles are just as important as the criticality cycles, and it is recommended that both be 100 or more. Batch size is then set to define the final total samples used in the criticality calculation. Note, that in the first case the settle cycles take 10% as much time (samples) as the criticality calculation, and in the other 3 cases settle cycles take 100% as much time (samples). This settle cycle calculation is overhead required to ensure that the flux distribution has settled sufficiently to allow averaging of the following criticality cycles to define the correct average K-eff.

**CAVEAT EMPTOR:** Always use Howerton's first law: “**We are in no rush for the wrong answer**” [2]. When using TART or any other Monte Carlo code do not make the mistake of trying to save computer time if it means sacrificing accuracy. If you do sacrifice accuracy you are not saving time, you are wasting it. This is particularly true of settle cycles; if you do any use enough settle generations to relax close to right answer, you will end up averaging incorrect generations to define your final estimate of the answer. The can greatly – make that GREATLY – bias the final average result.



The below table summarizes the results shown in the two tables that follow. First results are presented using **continuous energy cross section** for four criticality sample sizes varying from 100,000,000 down to 100,000, in step sizes of 10. Next results are presented **varying the cross section model**, comparing results using continuous energy cross sections, to multi-band [3, 4, 5], and multi-group cross sections.

It is worth noting how far we have come: today's time to sample (settle + criticality) 1.1 x (problems ) 68 X 10<sup>8</sup> samples per problem is 11,153 seconds. In this case TART is calculating over **670,000 samples per second**. 11,153 seconds is 186 minutes, or about 3 hours. The above table of running times shows today's TART is above 400 times faster than 30 years ago. In other words, in 1995 it would have taken TART about 1,200 hours or 50 days to do this calculation. Or if you want to go to the extreme of using my guess of 10,000 faster than a 1962 IBM-7090/94, back then it would have taken 1,250 days. Please remember that this set of 68 fast critical assemblies is only a small part of the 1,173 critical assemblies today routinely calculated by TART [1]. In our dreams if I had started these calculations in 1962 on a IBM-7090/94 by today they might not yet be finished; we could still be running today (my sense of humor).

**Continuous Energy Cross Sections**

Total Samples	Settle Cycles	Criticality Cycles	Batch Size	Time (Seconds) 68 Problems
10 <sup>8</sup>	100	1,000	100,000	11153.88
10 <sup>7</sup>	100	100	100,000	2161.40
10 <sup>6</sup>	100	100	10,000	136.57
10 <sup>5</sup>	100	100	1,000	1.68

**Multi-Band or Multi-Group Cross Sections**

10 <sup>6</sup> M-Band	100	100	10,000	119.29
10 <sup>6</sup> M-Group	100	100	10,000	63.43

Returning to how I will define **convergence** in this paper, I will assume any K-eff that differs from the 10<sup>8</sup> results by more than 0.1 % (3 digits accuracy) is questionable. In the below tables I have marked all such K-eff estimates in **YELLOW**. When we compare results to the 10<sup>8</sup> sample size results,

What we see from the first below table based on **sample size**,

- 1) 10<sup>7</sup>: ALL 68 results agree – to 0.1% 10<sup>7</sup> is as good as 10<sup>8</sup> = can save 10 times.
- 2) 10<sup>6</sup>: About 25% of the results differ = marginally o.k.
- 3) 10<sup>5</sup>: Most of the results differ = TOO FEW SAMPLES = CAVEAT EMPTOR

What we see from the second table based on **nuclear data models** and 10<sup>6</sup> samples,

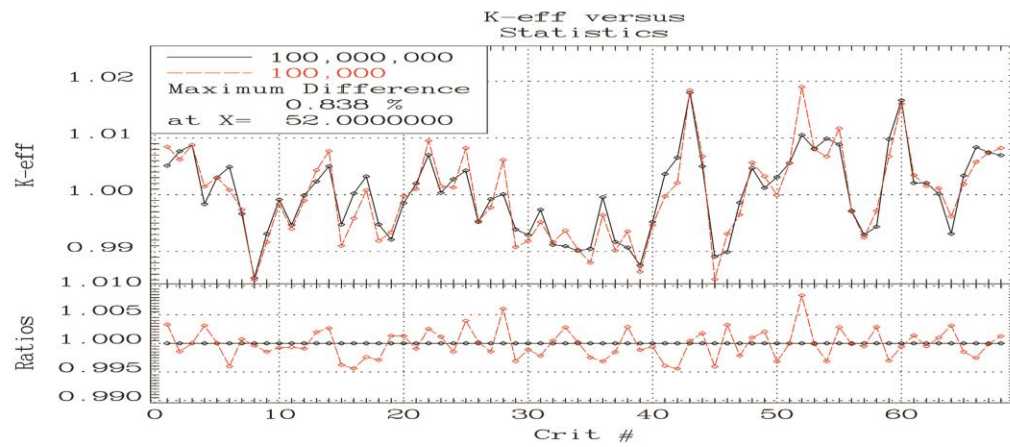
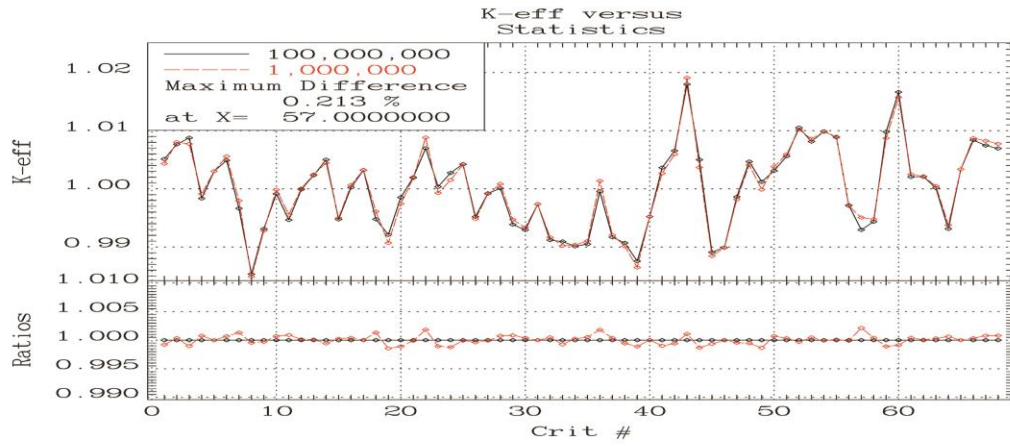
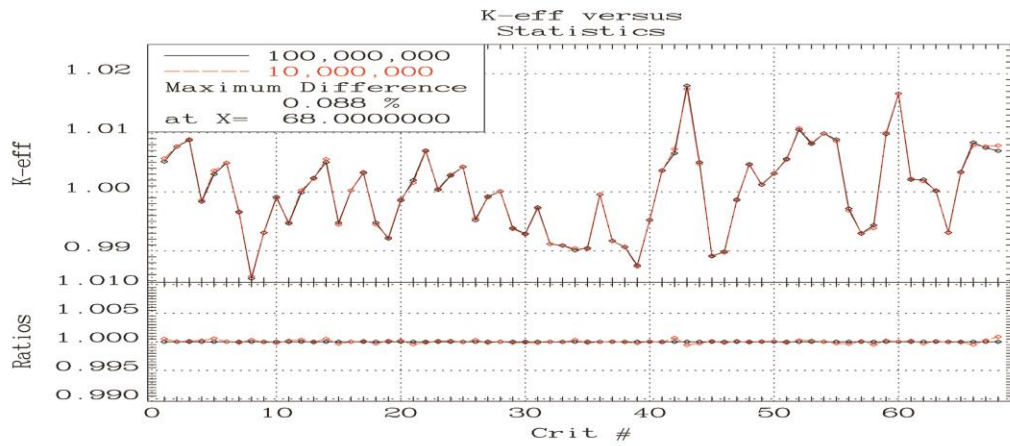
- 1) Continuous results are the same of the first table = marginally o.k.
- 2) Multi-band results are statistically better than continuous
- 3) Multi-Group is terrible - even the 68 case average exceeding 0.1%.

Hopefully you can take away from these results that 10<sup>5</sup> is far too few samples to use and expect accurate answers. At least 10<sup>6</sup> samples are needed to obtain even marginally acceptable answers. Note, that statistically the 10<sup>6</sup> multi-band results are better than the continuous (Average difference from 10<sup>8</sup>: Multi-Band 0.002% vs. Continuous 0.005%); which is what we expect since multi-band has far fewer degrees of freedom.

# 10<sup>8</sup>, 10<sup>7</sup>, 10<sup>6</sup>, 10<sup>5</sup> Convergence Samples Comparison

Crit. ID.	Fuel	Reflect	K-expect			K-expect			K-expect		
			100 Mill	10 Mill	% Diff.	1 Mill	% Diff.	0.1 Mill	% Diff.		
c10100	pu-a	be	5.222	1.005130	1.005650	-0.052	1.004330	0.080	1.008450	-0.332	
c20100	pu-a	be	8.170	1.007660	1.007700	-0.004	1.008020	-0.036	1.006220	0.144	
c30100	pu-a	be	13.00	1.008770	1.008920	-0.015	1.007730	0.104	1.008760	0.001	
c40100	pu-d			0.998348	0.998578	-0.023	0.999115	-0.077	1.001450	-0.310	
c50100	pu-d	be	3.690	1.003010	1.003600	-0.059	1.003020	-0.001	1.003060	-0.005	
c60100	pu-d	be	5.250	1.004900	1.004920	-0.002	1.005590	-0.069	1.000850	0.405	
c70100	pu-d	c	3.830	0.996636	0.996487	0.015	0.997957	-0.132	0.997371	-0.074	
c80100	pu-d	ti	8.000	0.985341	0.985670	-0.033	0.984891	0.045	0.984978	0.036	
c90100	pu-d	w	4.700	0.993105	0.993045	0.006	0.992816	0.029	0.991646	0.146	
c10010	pu-d	u-235	0.660	0.999140	0.998936	0.020	0.999839	-0.070	0.998351	0.079	
c11010	pu-d	u-238	1.930	0.994653	0.994847	-0.019	0.995566	-0.091	0.993983	0.067	
c12010	pu-d	u-238	6.740	0.999892	1.000260	-0.037	1.000040	-0.015	0.998975	0.092	
c13010	pu-d	u	4.130	1.002340	1.002230	0.011	1.002440	-0.010	1.004320	-0.198	
c14010	pu-d	u	19.60	1.005010	1.005530	-0.052	1.004450	0.056	1.007680	-0.267	
c10100	u-233			0.994754	0.994437	0.032	0.994997	-0.024	0.991002	0.375	
c20100	u-233	be	2.050	1.000220	1.000210	0.001	1.000630	-0.041	0.995827	0.439	
c30100	u-233	be	4.200	1.003220	1.003350	-0.013	1.003250	-0.003	1.000810	0.241	
c40100	u-233	w	2.440	0.994778	0.994475	0.030	0.996098	-0.132	0.991873	0.291	
c50100	u-233	w	5.790	0.992107	0.992280	-0.017	0.990685	0.142	0.993404	-0.130	
c60100	u-233	u-235	1.210	0.998533	0.998745	-0.021	0.997441	0.109	0.999813	-0.128	
c70100	u-233	u-235	1.980	1.001970	1.001540	0.043	1.001870	0.010	1.001030	0.094	
c80100	u-233	u-235	4.820	1.006990	1.006860	0.013	1.008820	-0.183	1.009520	-0.253	
c90100	u-233	u	2.300	1.000350	1.000510	-0.016	0.999280	0.107	1.001490	-0.114	
c10010	u-233	u	5.310	1.002740	1.002910	-0.017	1.001510	0.123	1.001270	0.147	
c11010	u-233	u	19.91	1.004240	1.004190	0.005	1.004160	0.008	1.008210	-0.397	
c001	u-235	be	1.27	0.995201	0.995523	-0.032	0.994833	0.037	0.995395	-0.019	
c002	u-235	be	2.54	0.999217	0.999044	0.017	0.999133	0.008	0.997761	0.146	
c003	u-235	c	1.27	1.000080	1.000110	-0.003	1.000840	-0.076	1.006120	-0.604	
c004	u-235	c	2.54	0.993876	0.993722	0.015	0.994717	-0.084	0.990783	0.309	
c005	u-235	mg	1.27	0.992930	0.992806	0.012	0.993336	-0.041	0.991841	0.109	
c006	u-235	mg	2.54	0.997364	0.997178	0.019	0.997353	0.001	0.995186	0.218	
c007	u-235	al	1.27	0.991186	0.991167	0.002	0.991661	-0.048	0.991646	-0.046	
c008	u-235	al	2.54	0.990938	0.990874	0.006	0.990207	0.073	0.993689	-0.275	
c009	u-235	ti	1.27	0.990121	0.990455	-0.033	0.990370	-0.025	0.990326	-0.021	
c010	u-235	ti	2.54	0.990491	0.990325	0.017	0.991024	-0.053	0.988037	0.245	
c011	u-235	fe	1.27	0.999562	0.999522	0.004	1.001360	-0.180	0.996415	0.315	
c012	u-235	fe	2.54	0.991684	0.991752	-0.007	0.992060	-0.038	0.990170	0.151	
c013	u-235	ni	1.27	0.990690	0.990622	0.007	0.990139	0.055	0.993540	-0.285	
c014	u-235	ni	2.54	0.987563	0.987346	0.022	0.986477	0.109	0.986415	0.115	
c015	u-235	cu	1.27	0.995208	0.995286	-0.008	0.995245	-0.004	0.994625	0.058	
c016	u-235	cu	2.54	1.003640	1.003550	0.009	1.002650	0.099	0.999706	0.393	
c017	u-235	mo	1.27	1.006540	1.007240	-0.070	1.005950	0.059	1.002090	0.445	
c018	u-235	mo	2.54	1.017970	1.017400	0.057	1.019080	-0.111	1.018400	-0.043	
c019	u-235	mo-allo		1.005000	1.004740	0.026	1.003680	0.132	1.006760	-0.176	
c020	u-235	w	1.27	0.989099	0.989213	-0.011	0.988451	0.065	0.985083	0.402	
c021	u-235	w	2.54	0.989894	0.989719	0.018	0.989853	0.004	0.993111	-0.322	
c10100	u-235			0.998608	0.998732	-0.012	0.998168	0.044	0.996463	0.215	
c20100	u-235			1.004680	1.004570	0.011	1.004130	0.055	1.005670	-0.099	
c30100	u-235			1.001210	1.001290	-0.008	0.999867	0.134	1.003250	-0.204	
c40100	u-235	be	2.222	1.003100	1.003170	-0.007	1.003820	-0.072	0.999906	0.319	
c50100	u-235	be	3.260	1.005590	1.005430	0.016	1.005940	-0.035	1.005540	0.005	
c60100	u-235	be	4.710	1.010520	1.010780	-0.026	1.010210	0.031	1.018990	-0.847	
c70100	u-235	be	5.440	1.008130	1.008310	-0.018	1.008600	-0.047	1.008000	0.013	
c80100	u-235	be	9.270	1.009880	1.009850	0.003	1.009810	0.007	1.006690	0.319	
c90100	u-235	be	11.79	1.008860	1.008600	0.026	1.008960	-0.010	1.011700	-0.284	
c10010	u-235	c	10.16	0.997181	0.996818	0.036	0.997079	0.010	0.997041	0.014	
c11010	u-235	c	15.24	0.992946	0.993044	-0.010	0.995065	-0.212	0.992447	0.050	
c12010	u-235	ni	4.940	0.994357	0.993871	0.049	0.994772	-0.041	0.997189	-0.283	
c13010	u-235	cu	5.030	1.009800	1.010020	-0.022	1.008710	0.109	1.006740	0.306	
c14010	u-235	cu	10.56	1.016640	1.016620	0.002	1.015750	0.089	1.016030	0.061	
c15010	u-235	w	5.080	1.002060	1.002240	-0.018	1.002470	-0.041	1.003440	-0.138	
c16010	u-235	w	10.16	1.002080	1.001800	0.028	1.002150	-0.007	1.001570	0.051	
c17010	u-235	pb	8.990	1.000150	1.000260	-0.011	1.000470	-0.032	1.001130	-0.098	
c18010	u-235	pb	17.22	0.993122	0.993054	0.007	0.993752	-0.063	0.996199	-0.308	
c19010	u-235	u	1.760	1.003370	1.003270	0.010	1.003370	0.000	1.001870	0.150	
c20010	u-235	u	4.470	1.008370	1.007890	0.048	1.008720	-0.035	1.005790	0.258	
c21010	u-235	u	9.960	1.007460	1.007710	-0.025	1.008260	-0.080	1.007330	0.013	
c22010	u-235	u	18.01	1.006950	1.007840	-0.089	1.007750	-0.080	1.008220	-0.127	
Average			1.000105	1.000127	-0.002	1.000158	-0.005	0.999980	0.013		
Minimum			0.985341	0.985670		0.984891		0.984978			
Maximum			1.017970	1.017400		1.019080		1.018990			

### $10^8, 10^7, 10^6, 10^5$ Convergence Samples Comparison



# 10<sup>8</sup>, 10<sup>6</sup>, 10<sup>6</sup> Multi-Band, 10<sup>6</sup> Multi-Group Convergence Samples Comparison

Crit. ID.	Fuel	Reflect	K-expect 100 Mill	K-expect 1 Mill	% Diff.	K-expect Bands	% Diff.	K-expect Groups	%	
Groups										
c10100	pub-a	be	5.222	1.005130	1.004330	0.080	1.003970	0.116	1.005570	-0.044
c20100	pu-a	be	8.170	1.007660	1.008020	-0.036	1.008580	-0.092	1.005190	0.247
c30100	pu-a	be	13.00	1.008770	1.007730	0.104	1.010050	-0.128	1.006290	0.248
c40100	pu-d			0.998348	0.999115	-0.077	0.999187	-0.084	0.997640	0.071
c50100	pu-d	be	3.690	1.003010	1.003020	-0.001	1.002620	0.039	1.003130	-0.012
c60100	pu-d	be	5.250	1.004900	1.005590	-0.069	1.003520	0.138	1.004090	0.081
c70100	pu-d	c	3.830	0.996636	0.997957	-0.132	0.996820	-0.018	0.998055	-0.142
c80100	pu-d	ti	8.000	0.985341	0.984891	0.045	0.985318	0.002	0.989815	-0.447
c90100	pu-d	w	4.700	0.993105	0.992816	0.029	0.995053	-0.195	0.994216	-0.111
c10010	pu-d	u-235	0.660	0.999140	0.999839	-0.070	0.999111	0.003	0.999096	0.004
c11010	pu-d	u-238	1.930	0.994653	0.995566	-0.091	0.994328	0.033	0.993655	0.100
c12010	pu-d	u-238	6.740	0.999892	1.000040	-0.015	1.000150	-0.026	1.000570	-0.068
c13010	pu-d	u	4.130	1.002340	1.002440	-0.010	1.002160	0.018	1.003400	-0.106
c14010	pu-d	u	19.60	1.005010	1.004450	0.056	1.004420	0.059	1.008050	-0.304
c10100	u-233			0.994754	0.994997	-0.024	0.994519	0.023	0.994887	-0.013
c20100	u-233	be	2.050	1.000220	1.000630	-0.041	0.998618	0.160	1.000270	-0.005
c30100	u-233	be	4.200	1.003220	1.003250	-0.003	1.002670	0.055	1.004020	-0.080
c40100	u-233	w	2.440	0.994778	0.996098	-0.132	0.994078	0.070	0.994385	0.039
c50100	u-233	w	5.790	0.992107	0.990685	0.142	0.992892	-0.079	0.990730	0.138
c60100	u-233	u-235	1.210	0.998533	0.997441	0.109	0.998305	0.023	0.998610	-0.008
c70100	u-233	u-235	1.980	1.001970	1.001870	0.010	1.001650	0.032	1.000780	0.119
c80100	u-233	u-235	4.820	1.006990	1.008820	-0.183	1.006000	0.099	1.006670	0.032
c90100	u-233	u	2.300	1.000350	0.999280	0.107	1.000830	-0.048	1.000170	0.018
c10010	u-233	u	5.310	1.002740	1.001510	0.123	1.003570	-0.083	1.002580	0.016
c11010	u-233	u	19.91	1.004240	1.004160	0.008	1.003330	0.091	1.004970	-0.073
c001	u-235	be	1.27	0.995201	0.994833	0.037	0.994637	0.056	0.995590	-0.039
c002	u-235	be	2.54	0.999217	0.999133	0.008	0.999669	-0.045	0.997613	0.160
c003	u-235	c	1.27	1.000080	1.000840	-0.076	0.998850	0.123	1.000430	-0.035
c004	u-235	c	2.54	0.993876	0.994717	-0.084	0.994266	-0.039	0.993075	0.080
c005	u-235	mg	1.27	0.992930	0.993336	-0.041	0.991861	0.107	0.993096	-0.017
c006	u-235	mg	2.54	0.997364	0.997353	0.001	0.998272	-0.091	0.998617	-0.125
c007	u-235	al	1.27	0.991186	0.991661	-0.048	0.990965	0.022	0.991389	-0.020
c008	u-235	al	2.54	0.990938	0.990207	0.073	0.991841	-0.090	0.990189	0.075
c009	u-235	ti	1.27	0.990121	0.990370	-0.025	0.990711	-0.059	0.990715	-0.059
c010	u-235	ti	2.54	0.990491	0.991024	-0.053	0.990928	-0.044	0.993513	-0.302
c011	u-235	fe	1.27	0.999562	1.001360	-0.180	0.998895	0.067	1.004400	-0.484
c012	u-235	fe	2.54	0.991684	0.992060	-0.038	0.991828	-0.014	1.000280	-0.860
c013	u-235	ni	1.27	0.990690	0.990139	0.055	0.990452	0.024	0.996786	-0.610
c014	u-235	ni	2.54	0.987563	0.986477	0.109	0.986948	0.061	1.003390	-1.583
c015	u-235	cu	1.27	0.995208	0.995245	-0.004	0.994339	0.087	0.996295	-0.109
c016	u-235	cu	2.54	1.003640	1.002650	0.099	1.001680	0.196	1.006860	-0.322
c017	u-235	mo	1.27	1.006540	1.005950	0.059	1.006520	0.002	1.008700	-0.216
c018	u-235	mo	2.54	1.017970	1.019080	-0.111	1.017800	0.017	1.018680	-0.071
c019	u-235	mo-allo		1.005000	1.003680	0.132	1.005010	-0.001	1.003840	0.116
c020	u-235	w	1.27	0.989099	0.988451	0.065	0.988973	0.013	0.990684	-0.159
c021	u-235	w	2.54	0.989894	0.989853	0.004	0.989500	0.039	0.990881	-0.099
c10100	u-235			0.998608	0.998168	0.044	0.999182	-0.057	0.998641	-0.003
c20100	u-235			1.004680	1.004130	0.055	1.003660	0.102	1.005610	-0.093
c30100	u-235			1.001210	0.999867	0.134	1.000430	0.078	1.001630	-0.042
c40100	u-235	be	2.222	1.003100	1.003820	-0.072	1.003510	-0.041	1.003750	-0.065
c50100	u-235	be	3.260	1.005590	1.005940	-0.035	1.005880	-0.029	1.005170	0.042
c60100	u-235	be	4.710	1.010520	1.010210	0.031	1.012290	-0.177	1.008970	0.155
c70100	u-235	be	5.440	1.008130	1.008600	-0.047	1.008470	-0.034	1.009000	-0.087
c80100	u-235	be	9.270	1.009880	1.009810	0.007	1.010810	-0.093	1.008240	0.164
c90100	u-235	be	11.79	1.008860	1.008960	-0.010	1.009800	-0.094	1.007650	0.121
c10010	u-235	c	10.16	0.997181	0.997079	0.010	0.997536	-0.035	0.996982	0.020
c11010	u-235	c	15.24	0.992946	0.995065	-0.212	0.994051	-0.111	0.991671	0.128
c12010	u-235	ni	4.940	0.994357	0.994772	-0.041	0.993762	0.060	1.010100	-1.574
c13010	u-235	cu	5.030	1.009800	1.008710	0.109	1.008550	0.125	1.014710	-0.491
c14010	u-235	cu	10.56	1.016640	1.015750	0.089	1.015500	0.114	1.027350	-1.071
c15010	u-235	w	5.080	1.002060	1.002470	-0.041	1.002530	-0.047	1.002860	-0.080
c16010	u-235	w	10.16	1.002080	1.002150	-0.007	1.002290	-0.021	1.004920	-0.284
c17010	u-235	pb	8.990	1.000150	1.000470	-0.032	1.000180	-0.003	1.002090	-0.194
c18010	u-235	pb	17.22	0.993122	0.993752	-0.063	0.993551	-0.043	0.997023	-0.390
c19010	u-235	u	1.760	1.003370	1.003370	0.000	1.004350	-0.098	1.003200	0.017
c20010	u-235	u	4.470	1.008370	1.008720	-0.035	1.008770	-0.040	1.009070	-0.070
c21010	u-235	u	9.960	1.007460	1.008260	-0.080	1.006740	0.072	1.006060	0.140
c22010	u-235	u	18.01	1.006950	1.007750	-0.080	1.008300	-0.135	1.006760	0.019
Average			1.000105	1.000158	-0.005	1.000085	0.002	1.001372	-0.127	
Minimum			0.985341	0.984891		0.985318		0.989815		
Maximum			1.017970	1.019080		1.017800		1.027350		

## Effect of Cross Sections: Measured vs. Theory

The below table illustrates the effect of two different sets of nuclear data. In this case both were run with exactly the same converged sample size,  $10^8$  neutron samples, and continuous cross sections; they differ only in the nuclear data used by each. Here I will not name the two sets of data, I will only mention that they differ in among other ways by the fact that one contains much more elemental data, and the other isotopic data.

The elemental data has the advantage of being better known from experiments, while the more theoretical isotopic data has the advantage of allowing each isotopic to be defined in more detail, such as different resolved and unresolved resonance regions. Which is better? Which produces more accurate results? Or does it matter?

In an attempt to answer these questions, I again return to the definition of **convergence** that I am using throughout this paper: that differences of more than 0.1% in K-eff is significant.

The below table illustrates that K-eff for all 68 cases differs by well in excess of 0.1%; indeed, even the average of all 68 cases is 0.5% different, with maximum differences approaching 1.0%. Note that the difference is not statistical, with some higher and some lower. Here in ALL 68 cases one set has a higher K-eff. Which one is better? Only time and testing will answer that question.

The point that I hope to make here is that even with the great strides we have made over the last more than 50 years to improve our nuclear data, it still remains a MAJOR factor and concern in our nuclear applications. Please always remember that even with a perfect code and methods, without good data we are in a “**Garbage In = Garbage Out**” situation.

# 10<sup>8</sup> Samples Comparison of two Different Sets of Nuclear Data

Crit. ID.	Fuel	Reflector		Expected K		% Diff
				Data A	Data B	
c10100	pu-a	be	5.222	1.0051300	0.9987940	0.634 %
c20100	pu-a	be	8.170	1.0076600	1.0014900	0.617 %
c30100	pu-a	be	13.000	1.0087700	1.0026600	0.611 %
c40100	pu-d			0.9983480	0.9915910	0.676 %
c50100	pu-d	be	3.690	1.0030100	0.9964720	0.654 %
c60100	pu-d	be	5.250	1.0049000	0.9985310	0.637 %
c70100	pu-d	c	3.830	0.9966360	0.9896390	0.700 %
c80100	pu-d	ti	8.000	0.9853410	0.9801100	0.523 %
c90100	pu-d	w	4.700	0.9931050	0.9862920	0.681 %
c10010	pu-d	u-235	0.660	0.9991400	0.9925030	0.664 %
c11010	pu-d	u-238	1.930	0.9946530	0.9870730	0.758 %
c12010	pu-d	u-238	6.740	0.9998920	0.9914460	0.845 %
c13010	pu-d	u	4.130	1.0023400	0.9943420	0.800 %
c14010	pu-d	u	19.600	1.0050100	0.9957360	0.927 %
c10100	u-233			0.9947540	0.9905680	0.419 %
c20100	u-233	be	2.050	1.0002200	0.9966000	0.362 %
c30100	u-233	be	4.200	1.0032200	1.0000200	0.320 %
c40100	u-233	w	2.440	0.9947780	0.9911950	0.358 %
c50100	u-233	w	5.790	0.9921070	0.9887550	0.335 %
c60100	u-233	u-235	1.210	0.9985330	0.9946780	0.386 %
c70100	u-233	u-235	1.980	1.0019700	0.9979380	0.403 %
c80100	u-233	u-235	4.820	1.0069900	1.0027800	0.421 %
c90100	u-233	u	2.300	1.0003500	0.9961960	0.415 %
c10010	u-233	u	5.310	1.0027400	0.9984810	0.426 %
c11010	u-233	u	19.910	1.0042400	0.9990690	0.517 %
c001	u-235	be	1.27	0.9952010	0.9906170	0.458 %
c002	u-235	be	2.54	0.9992170	0.9946110	0.461 %
c003	u-235	c	1.27	1.0000800	0.9952350	0.485 %
c004	u-235	c	2.54	0.9938760	0.9891320	0.474 %
c005	u-235	mg	1.27	0.9929300	0.9881520	0.478 %
c006	u-235	mg	2.54	0.9973640	0.9927210	0.464 %
c007	u-235	al	1.27	0.9911860	0.9860450	0.514 %
c008	u-235	al	2.54	0.9909380	0.9865740	0.436 %
c009	u-235	ti	1.27	0.9901210	0.9854960	0.462 %
c010	u-235	ti	2.54	0.9904910	0.9862920	0.420 %
c011	u-235	fe	1.27	0.9995620	0.9946920	0.487 %
c012	u-235	fe	2.54	0.9916840	0.9867870	0.490 %
c013	u-235	ni	1.27	0.9906900	0.9856940	0.500 %
c014	u-235	ni	2.54	0.9875630	0.9826570	0.491 %
c015	u-235	cu	1.27	0.9952080	0.9906600	0.455 %
c016	u-235	cu	2.54	1.0036400	0.9986830	0.496 %
c017	u-235	mo	1.27	1.0065400	1.0018700	0.467 %
c018	u-235	mo	2.54	1.0179700	1.0134100	0.456 %
c019	u-235	mo-alloy		1.0050000	1.0001800	0.482 %
c020	u-235	w	1.27	0.9890990	0.9846640	0.443 %
c021	u-235	w	2.54	0.9898940	0.9854630	0.443 %
c10100	u-235			0.9986080	0.9939070	0.470 %
c20100	u-235			1.0046800	1.0006900	0.399 %
c30100	u-235			1.0012100	0.9965570	0.465 %
c40100	u-235	be	2.222	1.0031000	0.9985040	0.460 %
c50100	u-235	be	3.260	1.0055900	1.0009900	0.460 %
c60100	u-235	be	4.710	1.0105200	1.0059100	0.461 %
c70100	u-235	be	5.440	1.0081300	1.0035700	0.456 %
c80100	u-235	be	9.270	1.0098800	1.0053600	0.452 %
c90100	u-235	be	11.790	1.0088600	1.0043400	0.452 %
c10010	u-235	c	10.160	0.9971810	0.9931480	0.403 %
c11010	u-235	c	15.240	0.9929460	0.9888790	0.407 %
c12010	u-235	ni	4.940	0.9943570	0.9898860	0.447 %
c13010	u-235	cu	5.030	1.0098000	1.0048500	0.495 %
c14010	u-235	cu	10.560	1.0166400	1.0121700	0.447 %
c15010	u-235	w	5.080	1.0020600	0.9976830	0.438 %
c16010	u-235	w	10.160	1.0020800	0.9977310	0.435 %
c17010	u-235	pb	8.990	1.0001500	0.9957320	0.442 %
c18010	u-235	pb	17.220	0.9931220	0.9886030	0.452 %
c19010	u-235	u	1.760	1.0033700	0.9984420	0.493 %
c20010	u-235	u	4.470	1.0083700	1.0028700	0.550 %
c21010	u-235	u	9.960	1.0074600	1.0008100	0.665 %
c22010	u-235	u	18.010	1.0069500	0.9998690	0.708 %
Averages				1.0001052	0.9950455	0.506 %
Minimum				0.9853410	0.9801100	
Maximum				1.0179700	1.0134100	



## Conclusions

### Convergence

I first defined what I will use as my definition of convergence in this paper. Many publications claim that with today's codes and nuclear data we can calculate K-eff for critical systems to within three digits, e.g., K-eff  $\sim 1.000 \pm 0.001$ ; 0.1 %. Personally, I think this is far too optimistic, but in this paper I will use it as a measure of convergence. I will FLAG any differences in K-eff that we calculate using the various approximations used in this paper, as a simple means of indicating a result we should be concerned by.

In this paper I have presented a few example results to illustrate both the advantages and disadvantages of Monte Carlo. One obvious advantage is its ability to model geometry more precisely, in more detail. A second advantage is its ability to define nuclear data cross sections more precisely, on a continuous energy basis. On the other hand, an obvious disadvantage is that the estimated solutions slowly converge to accurate answers. For example, convergence can vary as  $1/\sqrt{\text{samples}}$ , so that to improve accuracy by a factor of 10 requires an increase in samples, and therefore running time, by a factor of 100. Even with the computer power that we have today, it may not be practical to increase sample size, and therefore running time, by a factor of 100. More to the point, it certainly wasn't practical many years ago with the computers we then had; I will briefly address this point.

A controversial point is that convergence speed depends on the number of degrees of freedom in each problem. Using continuous energy cross sections admittedly improves the detail to which nuclear data can be represented, but it can vastly increase the number of degrees of freedom in a problem. For example, in a multi-group calculation, the number of total cross sections to sample is equal to the number of groups, whereas using continuous energy cross sections there can be many thousands of tabulated cross section, and using even simple linearly interpolation between tabulated values introduces vastly more degrees of freedom.

Here I present example results comparing continuous energy, multi-band, and multi-group results (both multi-band and multi-group use the same TART 616 energy groups). These results illustrate that even with 616 groups and a million neutron criticality samples, multi-group results differ significantly from results based on using continuous energy cross sections. We see this in particular for metal reflected critical assemblies; presumably due to neutron leakage through minima in the continuous energy cross sections, that are not included in the multi-group cross sections. However, even simple multi-band (2-band) results significantly improve agreement. This latter point should be of interest to users/developers of multi-group transport codes, since multi-band cross sections can be used in multi-group codes, but multi-group codes cannot easily be converted to use continuous energy data; the energy-to-energy transfer matrices become singular, and do not converge, as the number of groups are increased.

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**Lawrence Livermore National Laboratory  
Technical Information Department  
Livermore, CA 94551**

