# **Random number generation for parallel Monte Carlo**

**Protocol of a temporary obsession**

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# **Motivating example**

**The best problems are the ones you create yourself**

- Factoring CORSIKA out into a service allows flexible scaling
- **Problem**: CORSIKA's RNG is explicit internal state => result depends on which server handles the request.
- **Solution**: client maintains and communicates desired RNG state
- How to communicate and apply state without sacrificing quality or efficiency?
	- a) Brute force, or
	- b) clever math



### **Requirements for parallel random numbers**

### **A hierarchy of needs**

- A single pseudorandom sequence should have:
	- deterministic output
	- an extremely long period (2<sup>128</sup> or more)
	- no autocorrelation
- Parallel pseudorandom sequences (streams) should be:
	- 1. Disjoint and uncorrelated (provably, if possible)
	- 2. Quickly partitionable into arbitrarily sized substreams
	- 3. Independent of the degree of parallelization
	- 4.Small (<< than 20kB state of MT19937)
	- 5. Fast (random numbers should be cheaper than the calculation they feed)

# **Partitioning strategies**

- 1. Use a single generator with different initial state (seed) for each stream and hope for the best
	- Disjoint and uncorrelated: **maybe**
	- Paritionable: **no**
	- Independent of parallelization: **no**
- 2. Use the same seed, but different parameter sets
	- Disjoint and uncorrelated: **yes**
	- Partitionable: **maybe** (partitioning strategy has to be fixed at the outset)
	- Independent of parallelization: **maybe** (given a sufficiently large number of parameter sets)

### **Parameterized RNGs: SPRNG**

### **Scalable Parallel Random Number Generator [\(sprng.org](http://sprng.org))**

- GPL v2 license
- C++/FORTRAN bindings (custom interface), 3rd-party CUDA implementation exists
- Creates independent "streams" of random numbers
- Independence of streams theoretically proven (for some generators)
- Default generator (lagged Fibonacci) has 2<sup>39648</sup> independent streams, each with period 21310
- Streams partitioned in a tree with fixed but user-specified arity. Example: with 64 streams, each root generator can spawn 64 substreams, each substream can spawn 64 substreams of its own, etc.
- Pitfalls:
	- **It is possible to exhaust the parameter space** if you try hard enough.
	- **Initializing a full-period RNG is expensive** (O(ms), equivalent to ~2e5 random numbers).

### **Bad example: Multiply-with-carry RNG**

**as used in MCML ([atomic.physics.lu.se\)](http://www.atomic.physics.lu.se/fileadmin/atomfysik/Biophotonics/Software/CUDAMCML.pdf), clsim**

- Lag-1 MWC generator with period  $\sim$ 2<sup>60</sup>, different prime multipliers lead to independent streams
- Good:
	- Very fast (3 floating-point operations per call)
	- Very small (8-byte state fits comfortably in GPU local memory)
- Bad:
	- Number of independent streams limited to number of prime multipliers generated prior to run **(not arbitrarily partitionable)**
	- RNG is attached to a thread rather than work item, so result depends on (nondeterministic) mapping of work items to threads **(result depends on parallelization)**

# **Partitioning strategies (continued)**

#### **Leapfrog**  *ri*

- Disjoint: **yes**
- Uncorrelated: **maybe**
- Independent of parallelization: **no** *t*
- Quickly partitionable: **maybe**

### **Block split**

- Disjoint and uncorrelated: **yes** bisjonit dhu dhoonelated. **yes**
- Independent of parallelization: **yes**
- Quickly partitionable: **maybe** (requires efficient fast-forward by block size)



[Mertens (2009)]  $\mathbf{L}^{\text{noncon}}$  is independent in Figure 4. It does not require an a priori estimate of  $\mathbf{L}$ 

#### **Fast-forwarding a random number generator** So how does a purely mathematical random number generator work? The main numbers that are not from the right block or from the right leap subsequence. St-forwarding a fandom number generator

RNGs produce a recurrent sequence, i.e. the next state depends on the previous N

$$
r_i = f(r_{i-1}, r_{i-2}, \ldots, r_{i-n}),
$$

r det forwarding unodgen in positions by applying it) in times. In the difficult function function, this can be written as an M iterations of the matrix multiplication that we need to provide the first n numbers to provide the first n numbers to get this recurrence of the ground.<br>The first numbers to get this recurrence of the ground. The ground of the ground of the ground. The ground of Fast-forwarding through M positions by applying f() M times. If f is a linear

$$
\begin{pmatrix}\n r_{i-(n-1)} \\
\vdots \\
r_{i-1} \\
r_i\n\end{pmatrix}\n=\n\begin{pmatrix}\n 0 & 1 & \dots & 0 \\
\vdots & \vdots & \ddots & \vdots \\
0 & 0 & \dots & 1 \\
a_n & a_{n-1} & \dots & a_1\n\end{pmatrix}\n\begin{pmatrix}\n r_{i-n} \\
\vdots \\
r_{i-2} \\
r_{i-1}\n\end{pmatrix}\n\mod p
$$

 $A$ periodic sequences over a finite field can be generated by a linear recurrence, this<br>is always nessible in principle, but solv prestical for evolucity linear DNCs. No dividyo possibilo in principio, but only practical for explicitly integratives.<br>(linear congruential, general linear feedback shift registers, YARNs). between 0 and some prime number prime number prime number p is the linear recurrence prime number p is the lin<br>Distribution of the linear recurrence prime number p is the linear recurrence prime number of the linear recor and can be computed in  $O(n^3 \log M)$  time rather than  $O(n^3M)$ . Since all finite or is **always possible in principle,** but only practical for explicitly linear RNGs

[Mertens (2009)]

# **Block-splitting/leapfrogging RNGs: TRNG**

**Tina's Random Number Generator [\(numbercrunch.de/trng](https://www.numbercrunch.de/trng/))**

- 3-clause BSD license
- Passes full suite of empirical tests in TestU01
- C++11 random\_number\_engine and CUDA bindings
- Some engines with efficient split and skip operations
- Partitioning left to the user

### **Counter-based RNGs**

• An RNG is built out of two functions:

 $f\colon S\to S$  (state transition function)

 $g: S \rightarrow U$  (output function)

- Conventional RNGs have a complicated f() that produces integers over some range, and a simple g() that scales those integers to [0,1).
- Counter-based RNGs make f() simple (a counter!) and a g() that
	- Maps arbitrarily sequences of integers onto another set whose distribution is indistinguishable from noise
	- Is reasonable fast to evaluate
	- $\approx$  g() has the same properties as a good cryptographic block cypher!

### **Counter-based RNGs: Random123**

**"Random numbers: as easy as 1, 2, 3" [\(deshawresearch.com](http://www.deshawresearch.com/downloads/download_random123.cgi/))**

- 3-clause BSD license
- Passes full suite of empirical tests in TestU01
- C, C++11 random\_number\_engine, CUDA bindings
- Faster than MT19937 on CPUs with AES-NI support
- $2^{64}$  possible streams, each with  $2^{128}$  period
- Skip and split operations naturally supported, and practically free
- Partitioning left to the user

### **Summary**

- Massively parallel random number generation is a common problem, and there are known solutions.
- In all cases, random number generation should be deterministic and independent of granularity of parallelism, execution order, etc.
	- Attach RNG stream/block to particle (or whatever other atomic unit you have in your simulation)
	- Ensure that the conditions for creating a new stream/block are deterministic
- The implementation depends on the characteristics of the simulation
	- For explicit parallelism with rare, predictable branching and no restrictions on local memory: use SPRNG streams
	- For implicit (dynamically load-balanced) parallelism, or with unpredictable workloads, assign a (dynamically sized) block to each work item
		- TRNG: fast-forward blocks in logarithmic time
		- Random123: fast-forward blocks in constant time

### **Further reading**

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