# 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)

- 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 2<sup>1310</sup>
- 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), clsim

- Lag-1 MWC generator with period ~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

- Disjoint: yes
- Uncorrelated: maybe
- Independent of parallelization: no
- Quickly partitionable: maybe (requires efficient fast-forward by N)

#### **Block split**

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



[Mertens (2009)]

## Fast-forwarding a random number generator

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

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

Fast-forwarding through M positions by applying f() M times. If f is a linear function, this can be written as an M iterations of the matrix multiplication

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

and can be computed in O(n<sup>3</sup>logM) time rather than O(n<sup>3</sup>M). Since all finite or periodic sequences over a finite field can be generated by a linear recurrence, this is **always possible in principle**, but only practical for explicitly linear RNGs (linear congruential, general linear feedback shift registers, YARNs).

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

Tina's Random Number Generator (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:

 $f: S \to S$  (state transition function)  $g: S \to U$  (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
  - => 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)

- 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<sup>64</sup> possible streams, each with 2<sup>128</sup> 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**

- Gao, S., & Peterson, G. D. (2013). GASPRNG: GPU accelerated scalable parallel random number generator library. *Computer Physics Communications*, *184*(4), 1241–1249. <u>http://doi.org/10.1016/j.cpc.2012.12.001</u>
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