

HPCS'2015

A Runtime/Memory Trade-off of the Continuous Ziggurat Method on GPUs

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July 21, 2015







Topics

Motivation/Introduction

The Ziggurat method

Definition of the Ziggurat Algorithmic description

Memory/runtime trade-off

Results

Influence of the number of strips Comparison with alternative PRNGs

Conclusion





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Motivation/Introduction

Random numbers are required in many fields

- Cryptography
- Monte Carlo methods
- Simulating stochastic processes
- Stochastic/Random differential equations
- . .





Motivation/Introduction

Random numbers are required in many fields

- Cryptography
- Monte Carlo methods
- Simulating stochastic processes
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- ..

Generation of random numbers on a computer

- Determined by a deterministic rule
- Such random numbers are not really "random" but satisfy certain statistical criteria (pseudo random numbers)
- In general, such rules determine a uniformly distributed random number $u_{integer}^{uniform} \in \{0, ..., ULONG_MAX\}$ and $u_{finat}^{uniform} \in [0, 1]$, resp.
- A transformation function transforms this number to desired distribution, e.g. normal distribution u^{normal}, by "imitating" the inverse CDF
- ⇒ Ziggurat method [MT00]





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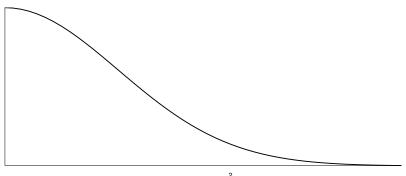
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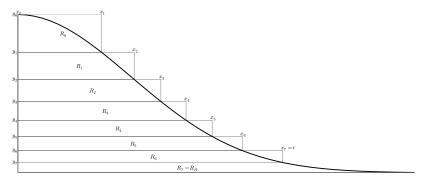
Definition of the Ziggurat: Gaussian bell function



$$f(x)=e^{-\frac{x^2}{2}}$$



Definition of the Ziggurat: Approximation by strips



 R_i strip

N number of strips

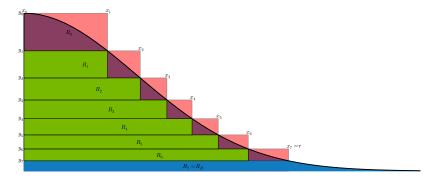
 (x_i, y_i) right bottom corner of all rectangular strips R_0, \ldots, R_{N-2} v common area of all strips R_0, \ldots, R_{N-1}

 $r = x_{N-1}$ right-most edge of a rectangular strip

Christoph Riesinger: A Runtime/Memory Trade-off of the Continuous Ziggurat Method on GPUs
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Definition of the Ziggurat: Four different regions



central region tail region cap region base strip rectangle totally under bell function remainder of strip under bell function remainder of strip over bell function non-rectangular strip R_{N-1}





Algorithmic description

The Ziggurat is stored by saving x_1, \ldots, x_{N-1} in a lookup table

 $x_1 = 0.738368917976448$

 $x_2 = 1.027386371780228$

 $x_3 = 1.262970198530832$

 $x_4 = 1.485358675643293$

 $x_5 = 1.716508125776779$

 $x_6 = 1.981904936400510$

 $x_7 = 2.338371698247252$





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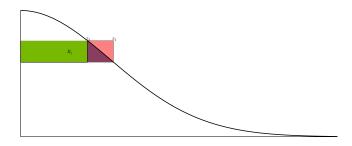
How the Ziggurat method works

- **1.** Generate a $u_{integer}^{uniform}$ and its corresponding $u_{float}^{uniform}$
- 2. To transform $u^{uniform}$ to u^{normal} select the k-th strip of the Ziggurat by $k = u_{integer}^{uniform} \& (2^N 1)$
- **3.** Depending on k, the transformation is done as follows...





Algorithmic description: Central region



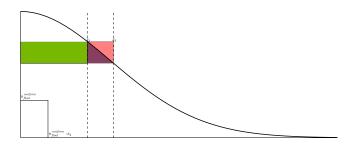
A central region is hit if

- $k \neq N 1$
- $u_{float}^{uniform} \leq x_{k+1}/x_k$





Algorithmic description: Central region



A central region is hit if

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Then, transformation is done by

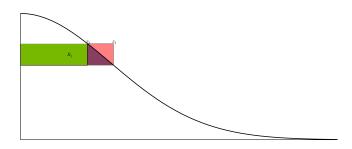
$$u^{normal} = u_{float}^{uniform} \cdot x_{k+1}$$

 \Rightarrow great, because cheap





Algorithmic description: Tail region



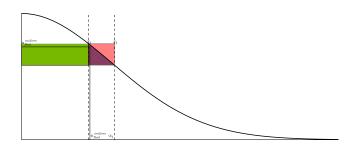
A tail region is hit if

- $k \neq N 1$
- · the central region is not hit
- $u_{float}^{uniform} \cdot (f(x_{k+1}) f(x_k)) < f(u_{float}^{uniform} \cdot x_{k+1}) f(x_{k+1})$





Algorithmic description: Tail region



A tail region is hit if

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Then, transformation is done by

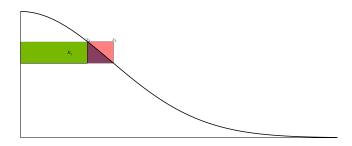
$$u^{normal} = u_{float}^{uniform} \cdot x_{k+1}$$

 \Rightarrow great, because cheap





Algorithmic description: Cap region



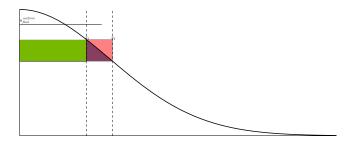
A cap region is hit if

- $k \neq N 1$
- neither the central nor tail region are hit





Algorithmic description: Cap region



A cap region is hit if

- $k \neq N 1$
- neither the central nor tail region are hit

Then, transformation is done by

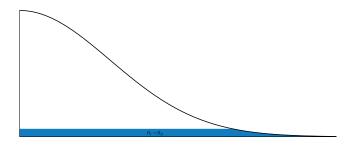
Restart the Ziggurat method with new $u^{uniform}$

⇒ bad, because expensive





Algorithmic description: Base strip



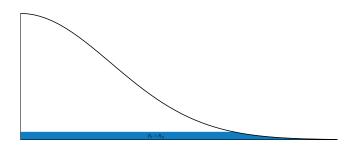
The base strip is hit if

•
$$k = N - 1$$





Algorithmic description: Base strip



The base strip is hit if

•
$$k = N - 1$$

Then, transformation is done by

$$x = (v \cdot u_{float}^{uniform}) / f(r)$$

• If
$$x < r$$
, then $u^{normal} = x$

• If
$$x \ge r$$
, then [Mar64]

⇒ bad, because expensive





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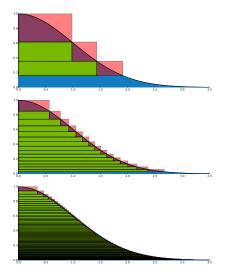
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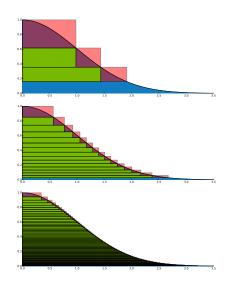
Influence of the number of strips







Influence of the number of strips



| Number of strips | % of covered area | | |
|------------------|-------------------|--|--|
| $2 = 2^1$ | 0.0% | | |
| $4 = 2^2$ | \sim 42.98% | | |
| $8 = 2^3$ | \sim 69.75% | | |
| $16 = 2^4$ | \sim 84.13% | | |
| $32 = 2^5$ | ~ 91.71% | | |
| $64 = 2^6$ | \sim 95.64% | | |
| $128 = 2^7$ | ~ 97.71% | | |
| $256 = 2^8$ | \sim 98.80% | | |
| $512 = 2^9$ | $\sim 99.37\%$ | | |
| $1024 = 2^{10}$ | \sim 99.67% | | |
| $2048 = 2^{11}$ | $\sim 99.83\%$ | | |
| $4096 = 2^{12}$ | \sim 99.91% | | |
| $8192 = 2^{13}$ | \sim 99.95% | | |





HPC aspects of number of used strips

- The more strips are used for the Ziggurat, the bigger the ratio of the sum of all central regions to the sum of all strips gets
- The bigger this ratio gets, the higher the likelihood to hit a (cheap) central region gets
- In addition, on GPUs, this reduces the likelihood for warp divergence



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| #strips | likelihood of warp div. |
|-----------------|----------------------------------|
| $2 = 2^1$ | $1 - 0.0^{32} = 100.0\%$ |
| $4 = 2^2$ | $1 - 0.4298^{32} \approx 99.9\%$ |
| $8 = 2^3$ | $1 - 0.6975^{32} \approx 99.9\%$ |
| $16 = 2^4$ | $1 - 0.8413^{32} \approx 99.9\%$ |
| $32 = 2^5$ | $1 - 0.9171^{32} \approx 93.7\%$ |
| $64 = 2^6$ | $1 - 0.9564^{32} \approx 76.0\%$ |
| $128 = 2^7$ | $1-0.9771^{32}\approx 52.3\%$ |
| $256 = 2^8$ | $1-0.9880^{32}\approx 32.0\%$ |
| $512 = 2^9$ | $1 - 0.9937^{32} \approx 18.3\%$ |
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- In addition, on GPUs, this reduces the likelihood for warp divergence
- So runtime can be reduced by using more strips which results in larger lookup tables
 - ⇒ runtime/memory trade-off

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Setup

GPUs

| GPU architecture | Fermi | Kepler | Maxwell |
|------------------------------|---------|------------|------------|
| Model | M2090 | Tesla K40m | GTX 750 Ti |
| Compute capability | 2.0 | 3.5 | 5.0 |
| #Processing elements | 16 × 32 | 15 × 192 | 5 × 128 |
| Size of shared memory (KB) | 48 | 48 | 48 |
| SP peak performance (TFLOPS) | 1.3312 | 3.84192 | 1.6384 |



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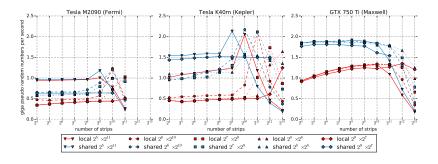
Setup

- For every run, 2²⁸ float numbers are generated
- Each thread produces 2¹² random numbers
- All floating point operations are done in single precision
- Uniform input u^{uniform} is generated by cuRAND XORWOW [Mar03]
- Execution times contain generation of $u_{integer}^{uniform}$ and time for transformation





Influence of the number of strips

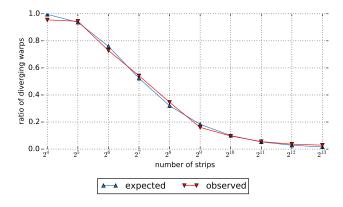


- States for XORWOW can be stored in local or shared memory
- Value of interest is giga pseudo random numbers per second (GPRNs/s)
- For most configurations, a single peak of performance is observable





Influence of the number of strips: Warp divergence

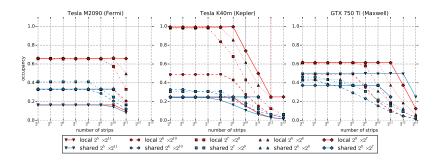


⇒ For low strip numbers, performance is limited by high warp divergence





Influence of the number of strips: Occupancy

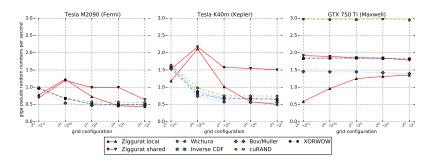


⇒ For large strip numbers, performance is limited by low occupancy





Comparison with alternative PRNGs



- For Ziggurat, best configuration (in terms of used strips) is used
- Wichura [Wic88] is a direct generator for random numbers, inverse CDF uses normcdfinvf(), and Box/Muller [BM58] is the most popular transformation function
- Ziggurat shows better performance then all other methods on Fermi (+19.9%) and Kepler (+24.3%), but not on Maxwell (-35.8%)





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Conclusion

- The Ziggurat method is a rejection method to transform uniformly distributed random numbers to normally distributed random numbers
- A runtime/memory trade-off allows reducing the execution time of the transformation by spending more memory for larger lookup tables
- Especially GPUs benefit from this trade-off since it leads to lower warp divergence
- While an implementation for CPUs typically uses 128 or 256 strips, a flat recommendation for GPUs is not possible
- Our Ziggurat implementation is up to 19.9% faster on Fermi and up to 24.3% faster on Kepler





Bibliography



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Final slide



