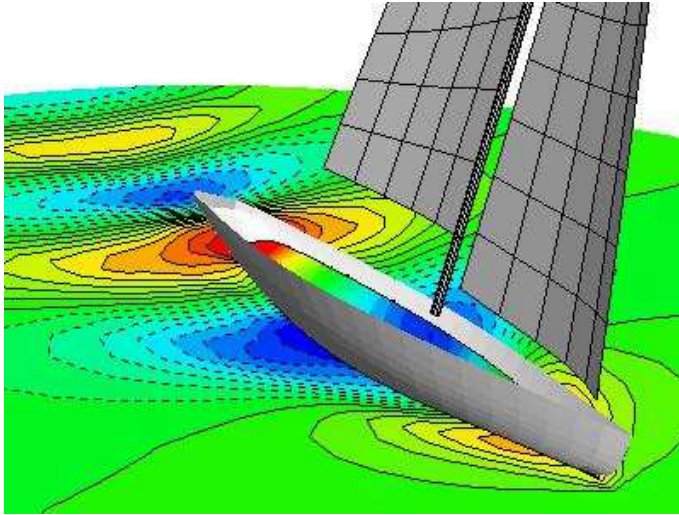


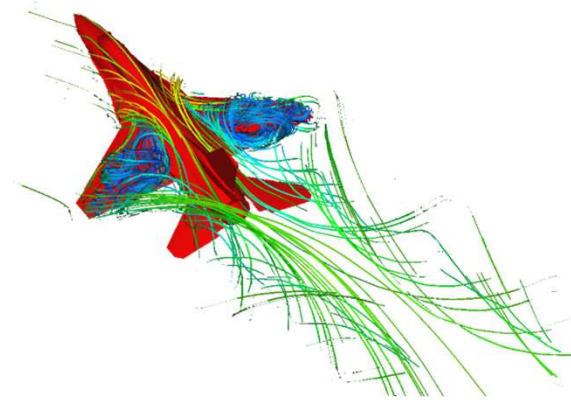
# CS 179: GPU Computing

## Lecture 18: Simulations and Randomness

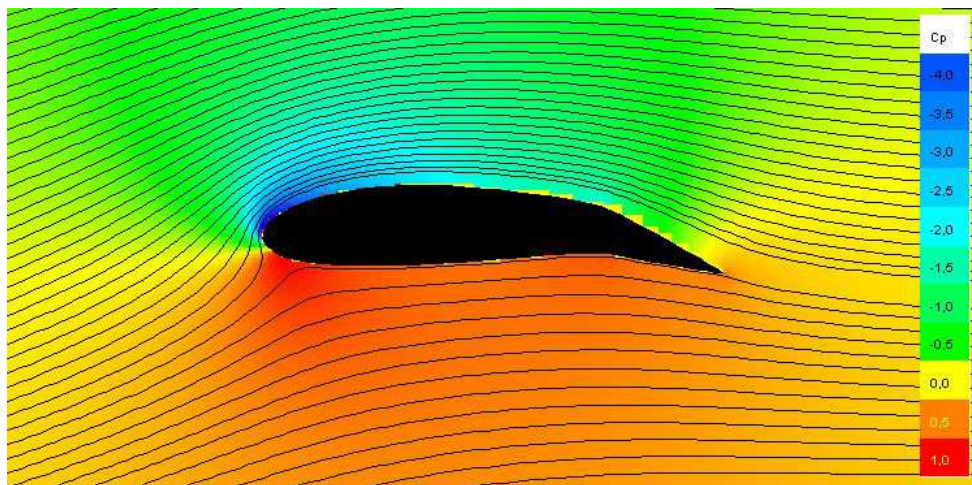
# Simulations



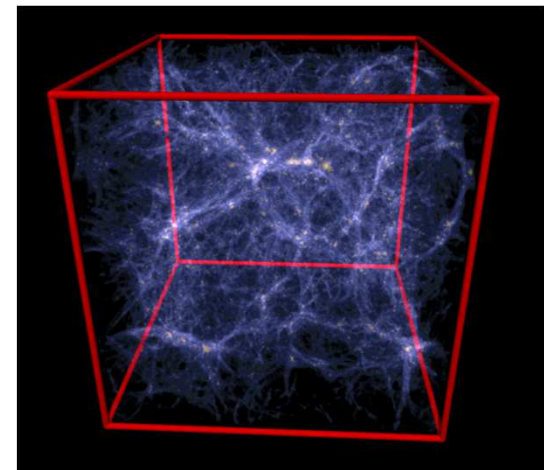
South Bay Simulations,  
<http://www.panix.com/~brosen/graphics/iacc.400.jpg>



Exa Corporation, <http://www.exa.com/images/f16.png>



Flysurfer Kiteboarding, <http://www.flysurfer.com/wp-content/blogs.dir/3/files/gallery/research-and-development/zwischenablage07.jpg>



Max-Planck Institut, <http://www.mpa-garching.mpg.de/gadget/hydrosims/>

# Simulations

- But what if your problem is hard to solve? e.g.
  - EM radiation attenuation
  - Estimating complex probability distributions
  - Complicated ODEs, PDEs
    - (e.g. option pricing in last lecture)
  - Geometric problems w/o closed-form solutions
    - Volume of complicated shapes

# Simulations

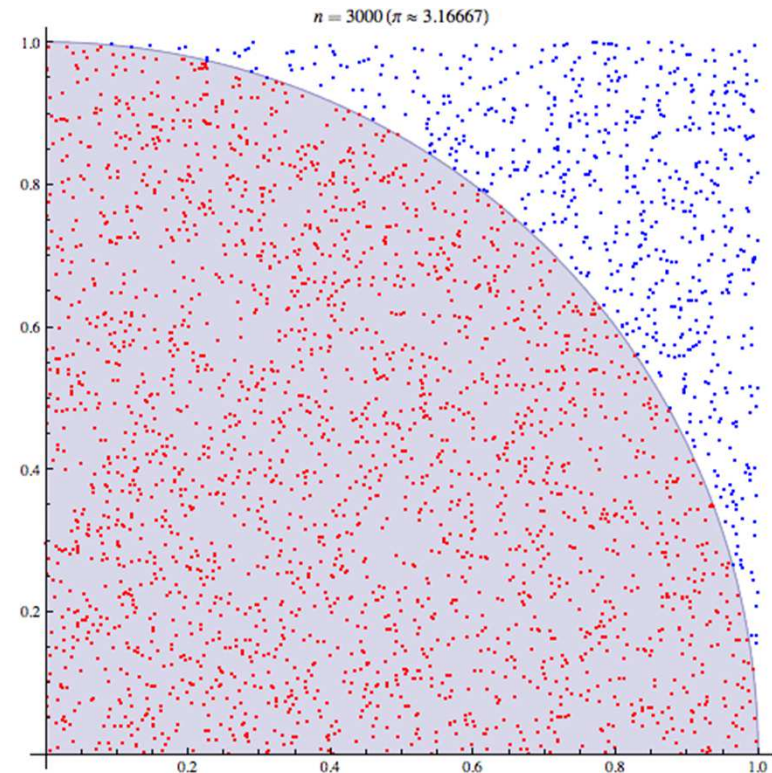
- Potential solution: **Monte Carlo methods**
  - Run simulation with randomly chosen inputs
    - (Possibly according to some distribution)
  - Do it again... and again... and again...
  - Aggregate results

# Monte Carlo example

- Estimating the value of  $\pi$

# Monte Carlo example

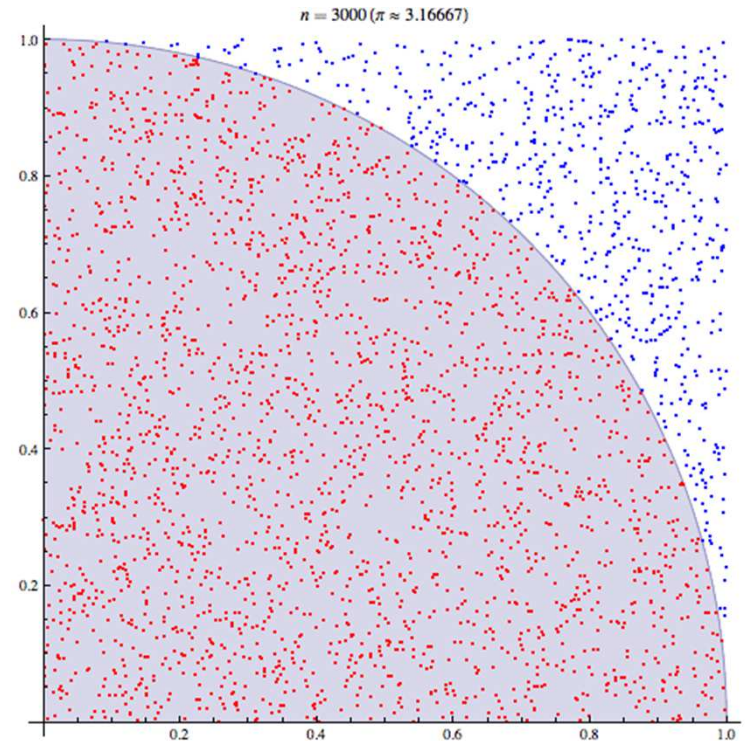
- Estimating the value of  $\pi$ 
  - Quarter-circle of radius  $r$ :
    - Area =  $(\pi r^2)/4$
  - Enclosing square:
    - Area =  $r^2$
  - Fraction of area:  $\pi/4$



"Pi 30K" by CaitlinJo - Own workThis mathematical image was created with Mathematica. Licensed under CC BY 3.0 via Wikimedia Commons - [http://commons.wikimedia.org/wiki/File:Pi\\_30K.gif#/media/File:Pi\\_30K.gif](http://commons.wikimedia.org/wiki/File:Pi_30K.gif#/media/File:Pi_30K.gif)

# Monte Carlo example

- Estimating the value of  $\pi$ 
  - Quarter-circle of radius  $r$ :
    - Area =  $(\pi r^2)/4$
  - Enclosing square:
    - Area =  $r^2$
  - Fraction of area:  $\pi/4 \approx 0.79$



- “Solution”: Randomly generate lots of points, calculate fraction within circle
  - Answer should be pretty close!

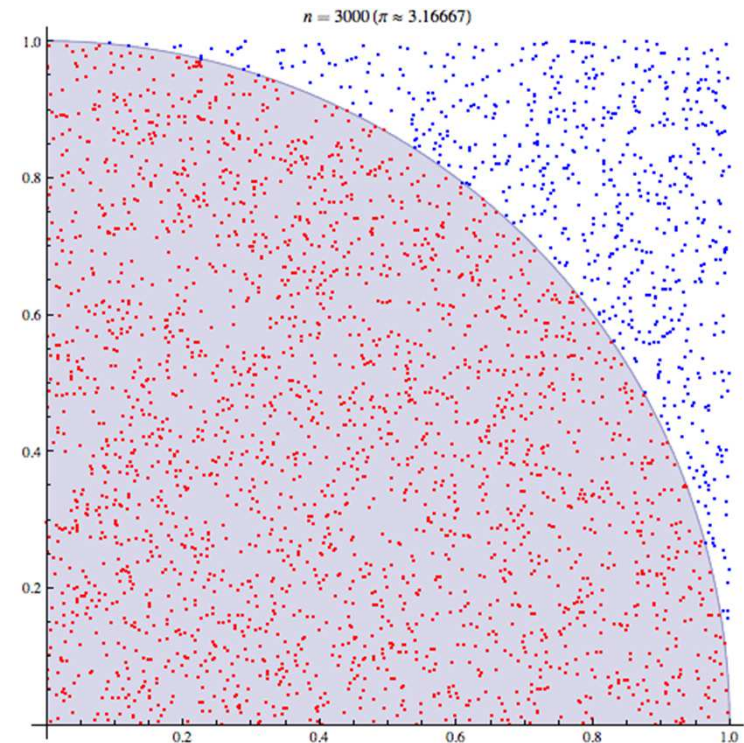
"Pi 30K" by CaitlinJo - Own workThis mathematical image was created with Mathematica. Licensed under CC BY 3.0 via Wikimedia Commons - [http://commons.wikimedia.org/wiki/File:Pi\\_30K.gif#/media/File:Pi\\_30K.gif](http://commons.wikimedia.org/wiki/File:Pi_30K.gif#/media/File:Pi_30K.gif)

# Monte Carlo example

- Pseudocode:

```
(simulate on N points)  
(assume r = 1)
```

```
points_in_circle = 0  
for i = 0,...,N-1:  
    randomly pick point (x,y) from  
        uniform distribution in  $[0,1]^2$   
    if (x,y) is in circle:  
        points_in_circle++  
  
return (points_in_circle / N) * 4
```



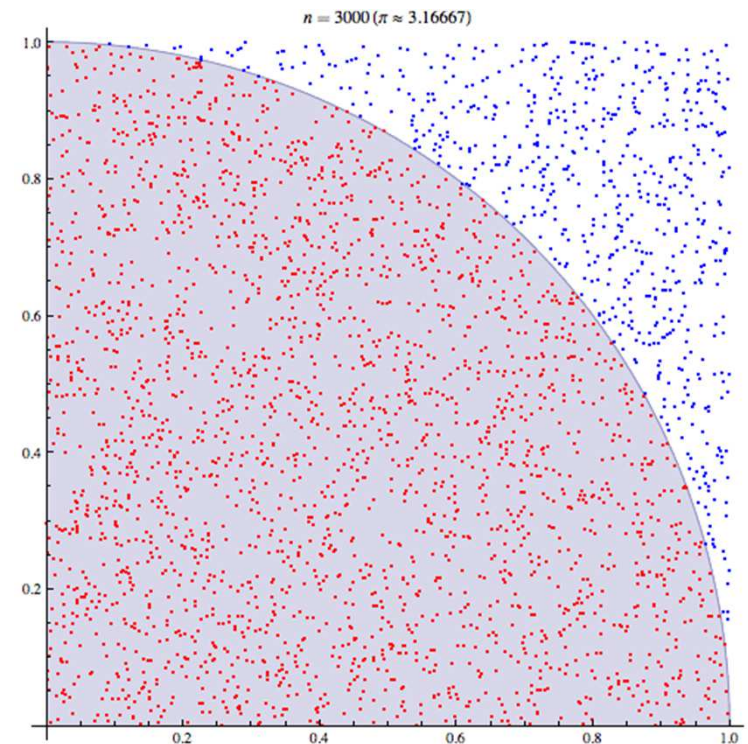


# Monte Carlo example

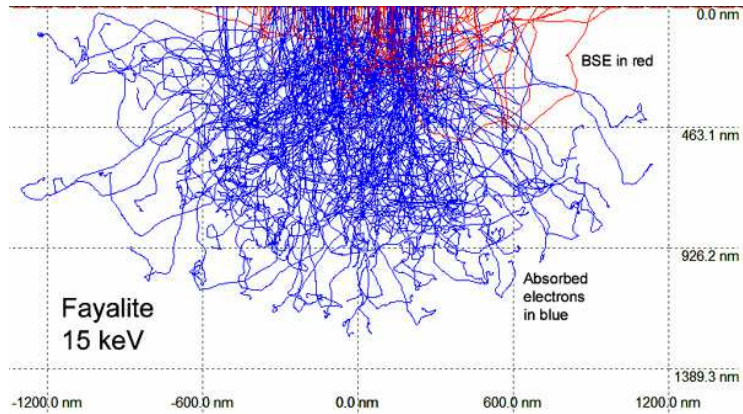
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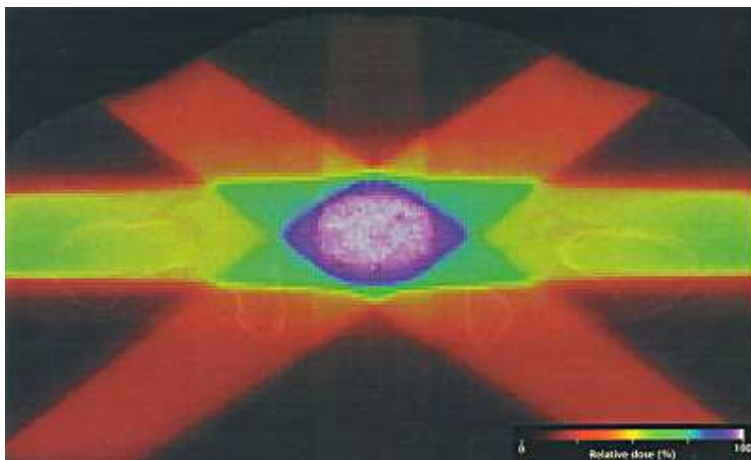
```
points_in_circle = 0  
for i = 0,...,N-1:  
    randomly pick point (x,y) from  
        uniform distribution in  $[0,1]^2$   
    if  $x^2 + y^2 < 1$ :  
        points_in_circle++  
  
return (points_in_circle / N) * 4
```



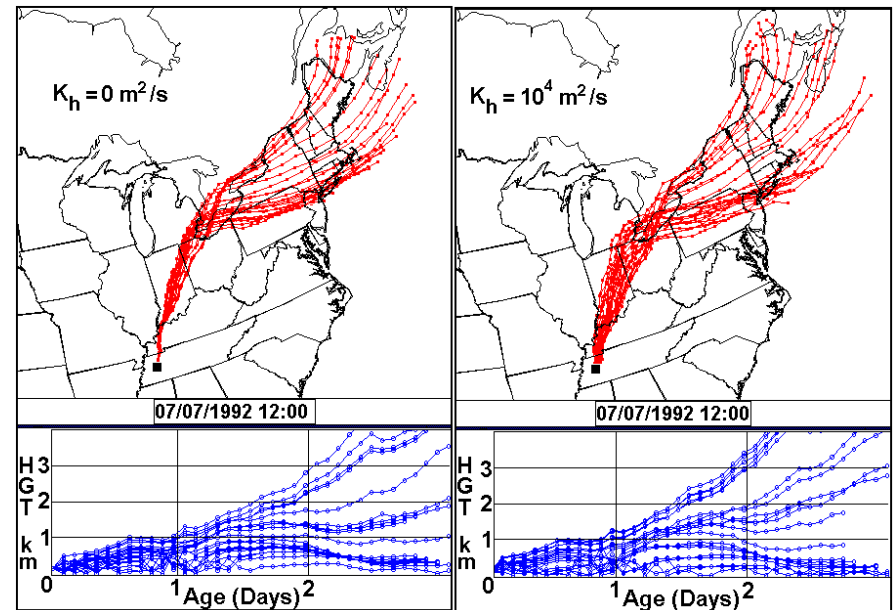
# Monte Carlo simulations



Planetary Materials Microanalysis Facility, Northern Arizona University, [http://www4.nau.edu/microanalysis/microprobe-sem/Images/Monte\\_Carlo.jpg](http://www4.nau.edu/microanalysis/microprobe-sem/Images/Monte_Carlo.jpg)



[http://www.cancernetwork.com/sites/default/files/cn\\_import/n0011bf1.jpg](http://www.cancernetwork.com/sites/default/files/cn_import/n0011bf1.jpg)



Center for Air Pollution Impact & Trend Analysis, Washington University in St. Louis, [http://www4.nau.edu/microanalysis/microprobe-sem/Images/Monte\\_Carlo.jpg](http://www4.nau.edu/microanalysis/microprobe-sem/Images/Monte_Carlo.jpg)

# General Monte Carlo method

- Pseudocode:

```
for (number of trials):  
    randomly pick value from a probability distribution  
    perform deterministic computation on inputs
```

(aggregate results)

# General Monte Carlo method

- Why it works:
  - Law of large numbers!

$$\bar{X}_n \rightarrow \mu \quad \text{for} \quad n \rightarrow \infty,$$

# General Monte Carlo method

- Pseudocode:

```
for (number of trials):  
    randomly pick value from a probability distribution  
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(aggregate results)

- Can we parallelize this?

# General Monte Carlo method

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(aggregate results)
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← Trials are  
independent

- Can we parallelize this?

# General Monte Carlo method

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← Usually so  
(e.g. with reduction)

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# General Monte Carlo method

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```
for (number of trials):
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```
    randomly pick value from a probability distribution
```

```
    perform deterministic computation on inputs
```

```
(aggregate results)
```



What about this?



Trials are  
independent



Usually so  
(e.g. with reduction)

- Can we parallelize this?



# Parallelized Random Number Generation

# Early Credits

- Algorithm and presentation based on:
  - “Parallel Random Numbers: As Easy as 1, 2, 3”
    - (Salmon, Moraes, Dror, Shaw) at D. E. Shaw Research
    - Developed for biomolecular simulations on Anton (massively parallel ASIC-based supercomputer)
    - Also applicable to CPUs, GPUs

# Random Number Generation

- Generating random data computationally is hard
  - Computers are deterministic!



# Random Number Generation

- Two methods:
  - Hardware random number generator
    - aka TRNG (“True” RNG)
    - Uses data collected from environment (thermal, optical, etc)
    - Very slow!
  - Pseudorandom number generator (PRNG)
    - **Algorithm** that produces “random-looking” numbers
    - Faster – limited by computational power

# Demonstration

# Random Number Generation

- PRNG algorithm should be:
  - High-quality
    - Produce “good” random data
  - Fast
    - (In its own right)
  - Parallelizable!
- Can we do it?
  - (Assume selection from uniform distribution)

# A Very Basic PRNG

- “Linear congruential generator” (LCG)

– e.g. C’s rand()

```
//from glibc
```

```
int32_t val = state[0];  
val = ((state[0] * 1103515245) + 12345)  
      & 0x7fffffff;  
state[0] = val;  
*result = val;
```

– General formula:

$$X_{n+1} = (aX_n + c) \bmod m$$

- $X_0$  is the “seed” (e.g. system time)

# A Very Basic PRNG

- “Linear congruential generator” (LCG)
  - e.g. C’s rand()

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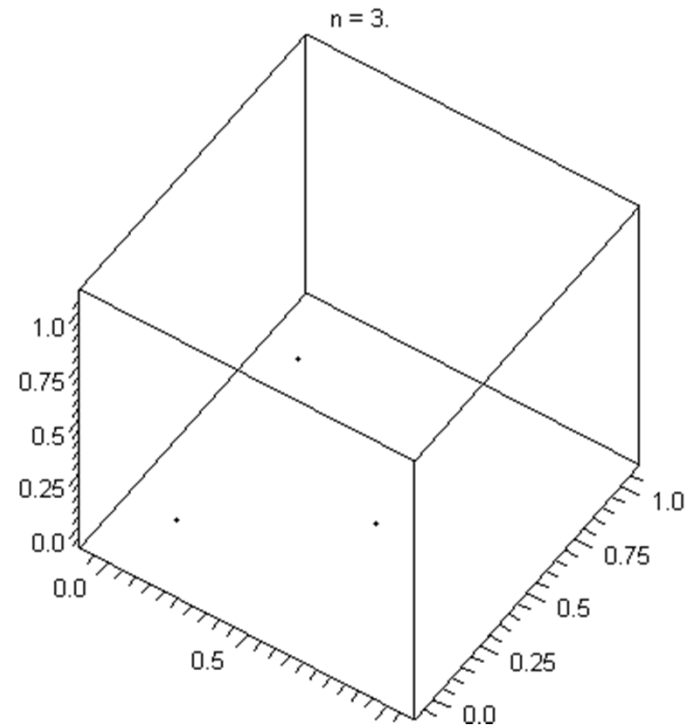
Non-parallelizable  
recurrence relation!



# Linear congruential generators

$$X_{n+1} = (aX_n + c) \bmod m$$

- Not high quality!
  - Clearly non-uniform
- Fast to compute
- Not parallelizable!



# Measures of RNG quality

- Impossible to prove a sequence is “random”
- Possible tests:
  - Frequency
  - Periodicity - do the values repeat too early?
  - Linear dependence
  - ...

# PRNG Parallelizability

- Many PRNGs (like the LCG) have a non-parallelizable appearance:

$$X_{n+1} = f(X_n)$$

- (Better chance of good data when):
  - All  $X_i$  in some large state space
  - Complicated function  $f$

# PRNG Parallelizability

- Possible “approach” to GPU parallelization:
  - Assign a PRNG to each thread!
    - Initialize with e.g. different  $X_0$
    - Thread 0 produces sequence  $X_{n+1,0} = f(X_{n,0})$
    - Thread 1 produces sequence  $X_{n+1,1} = f(X_{n,1})$
    - ...

# PRNG Parallelizability

- Possible “approach” to GPU parallelization:
  - Assign a PRNG to each thread!
    - Initialize with e.g. different  $X_0$
    - Thread 0 produces sequence  $X_{n+1,0} = f(X_{n,0})$
    - Thread 1 produces sequence  $X_{n+1,1} = f(X_{n,1})$
    - ...
  - In practice, often cannot get high quality
    - Repeated values, lack of good, enumerable parameters

# PRNG Parallelizability

- Instead of:

$$X_{n+1} = f(X_n)$$

- Suppose we had:

$$X_{n+1} = b(n)$$

– This is parallelizable! (Without our previous “trick”)

- Is this possible?

# More General PRNG

- “Keyed” PRNG given by:
  - Transition function:  $f: S \rightarrow S$
  - Output function:  $g: K \times S \rightarrow U$
- S: Internal (hidden) state space
- U: Output space
- K: “Key space”
  - Can “seed” output behavior without relying on  $X_0$  alone – useful for scientific reproducibility!

# More General PRNG

- “Keyed” PRNG given by:

- Transition function:  $f: S \rightarrow S$

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- U: Output space

- K: “Key space”

- Can “seed” output behavior without relying on  $X_0$  alone –  
useful for scientific reproducibility!

If S has  $J$  times more bits than U, can produce  $J$  outputs per transition.

Assume  $J = 1$  in this lecture



# More General PRNG

- “Keyed” PRNG given by:

– Transition function:  $f: S \rightarrow S$

– Output function:  $g: K \times S \rightarrow U$

– “Trivial” example: LCG  $X_{n+1} = (aX_n + c) \bmod m$

- $f(X_n) = aX_n + c$

- $g(X_n) = X_n$

- $S$  is (for example) the space of 32-bit integers

- $U = S$

- $K$  is “trivial” (no keys used)

# More General PRNG

- “Keyed” PRNG given by:

- Transition function:  $f: S \rightarrow S$

- Output function:  $g: K \times S \rightarrow U$

- “Trivial” example: LCG  $X_{n+1} = (aX_n + c) \bmod m$

- $f(X_n) = aX_n + c$

- $g(X_n) = X_n$

- $f$  is more complicated than  $g$ !

# More General PRNG

- “Keyed” PRNG given by:

- Transition function:  $f: S \rightarrow S$

- Output function:  $g: K \times S \rightarrow U$

- General theme:  $f$  is complicated,  $g$  is simple

- What if we flipped that?

# More General PRNG

- “Keyed” PRNG given by:
  - Transition function:  $f: S \rightarrow S$
  - Output function:  $g: K \times S \rightarrow U$
- General theme:  $f$  is complicated,  $g$  is simple
  - What if we flipped that?
  - What if  $f$  were so simple that it could be evaluated *explicitly*?

# More General PRNG

- i.e. what if we had:
  - Simple transition function (p-bit integer state space):
$$f(s) = (s + 1) \bmod 2^p$$
    - This is just a counter! Can expand into explicit formula
$$f(n) = (n + n_0) \bmod 2^p$$
      - These form **counter-based PRNGs**
  - Complicated output function  $g$
- Would this work?

# More General PRNG

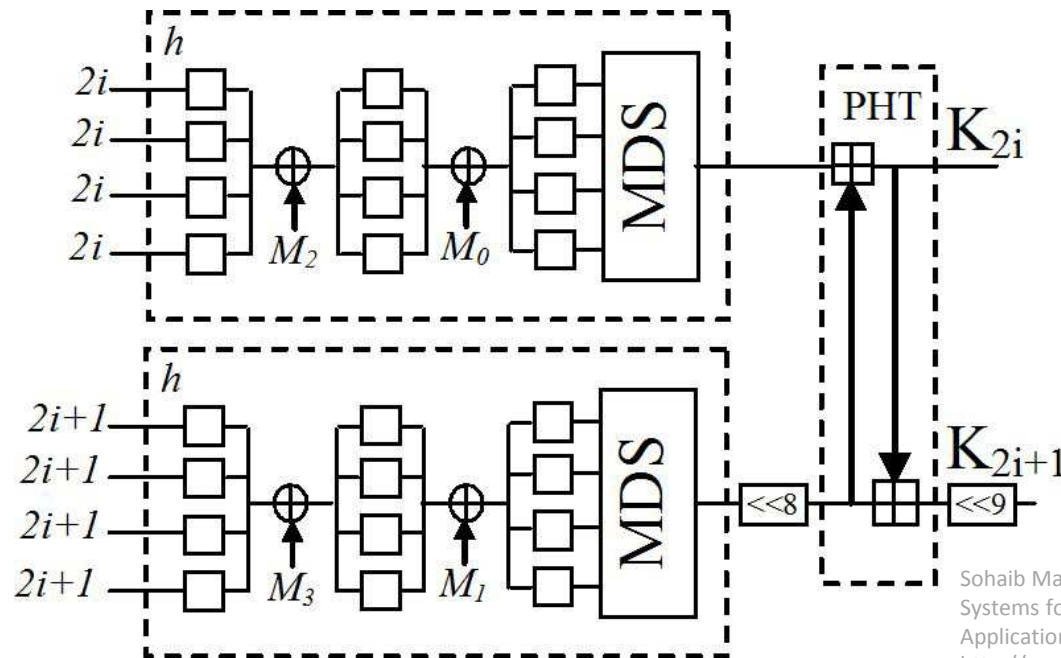
- i.e. what if we had:
  - Simple transition function  $f$
  - Complicated output function  $g(k, n)$ 
    - Should be *bijective* w/r/to  $n$ 
      - Guarantees period of  $2^p$
    - Shouldn't be *too* difficult to compute

# Bijjective Functions

- Cryptographic block ciphers!
  - AES (Advanced Encryption Standard), Threefish, ...
  - Must be bijective!
    - (Otherwise messages can't be encrypted/decrypted)

# AES-128 Algorithm

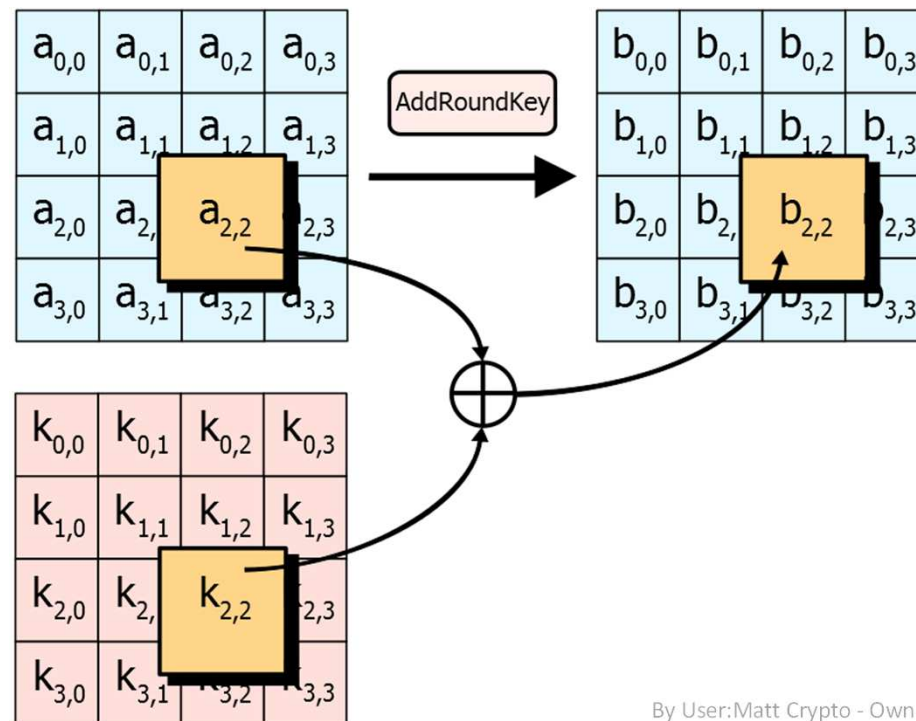
- 1) Key Expansion
  - Determine all keys  $k$  from initial cipher key  $k_B$ 
    - Used to strengthen weak keys





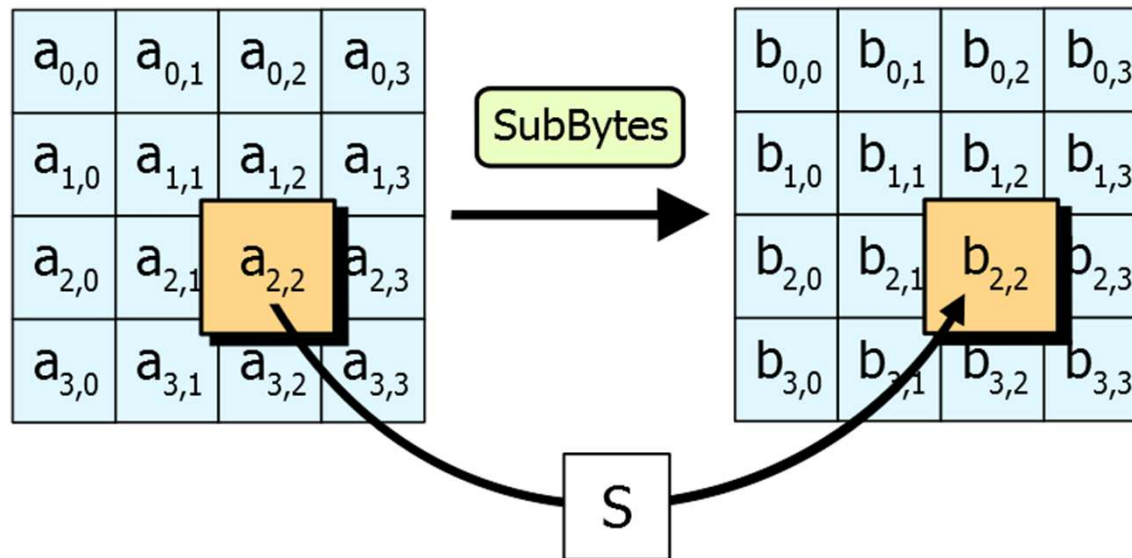
# AES-128 Algorithm

- 2) Add round key
  - Bitwise XOR state  $s$  with key  $k_0$



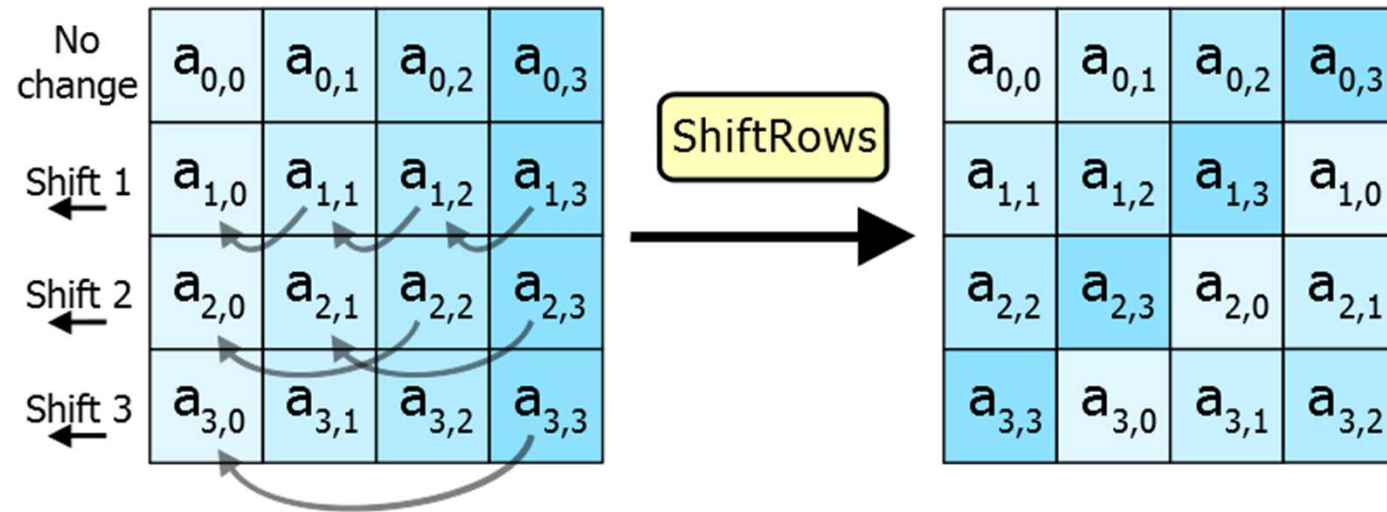
# AES-128 Algorithm

- 3) For each round... (10 rounds total)
  - a) Substitute bytes
    - Use lookup table to switch positions



# AES-128 Algorithm

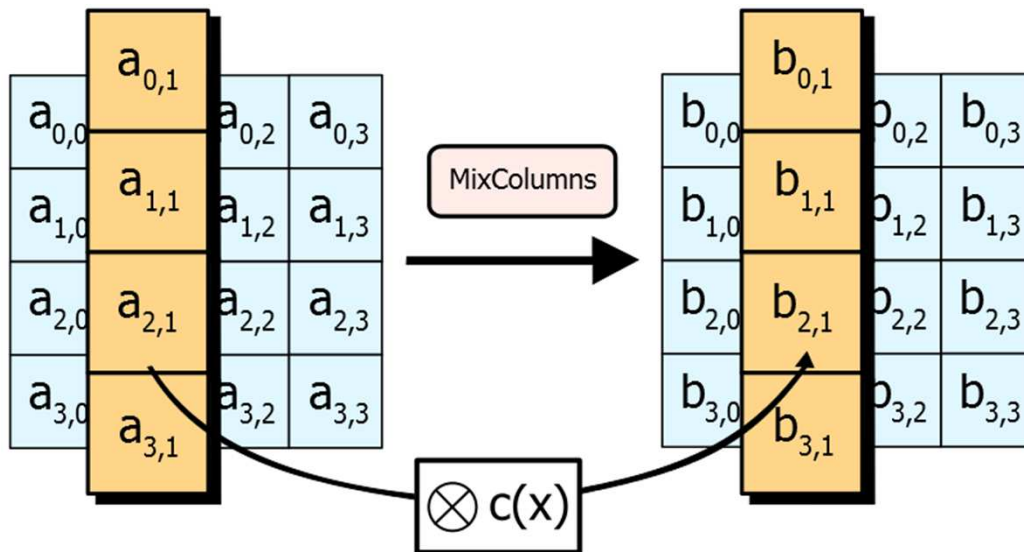
- 3) For each round...
  - b) Shift rows



# AES-128 Algorithm

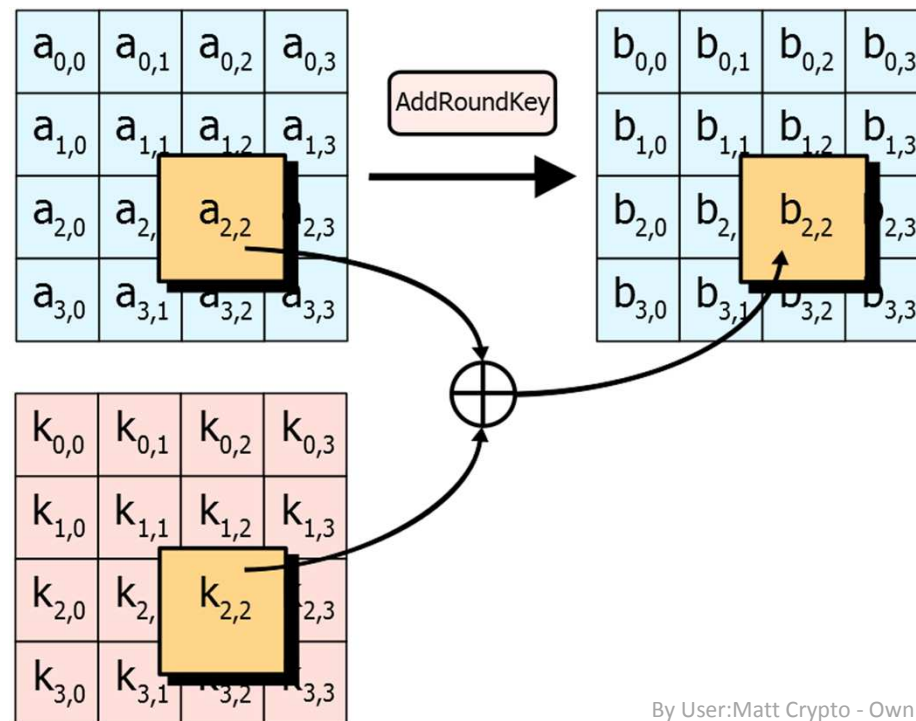
- 3) For each round...
  - c) Mix columns
    - Multiply by constant matrix

$$\begin{bmatrix} 2 & 3 & 1 & 1 \\ 1 & 2 & 3 & 1 \\ 1 & 1 & 2 & 3 \\ 3 & 1 & 1 & 2 \end{bmatrix}$$



# AES-128 Algorithm

- 3) For each round...
  - d) Add round key (as before)



# AES-128 Algorithm

- 4) Final round
  - Do everything in normal round except mix columns

# AES-128 Algorithm

- Summary:
  - 1) Expand keys
  - 2) Add round key
  - 3) For each round (10 rounds total)
    - Substitute bytes
    - Shift rows
    - Mix columns
    - Add round key
  - 4) Final round:
    - (do everything except mix columns)

# Algorithmic Improvements

- We have a good PRNG!
  - Simple transition function  $f$ 
    - Counter
  - Complicated output function  $g(k, n)$ 
    - AES-128



# Algorithmic Improvements

- We have a good PRNG!
  - Simple transition function  $f$ 
    - Counter
  - Complicated output function  $g(k, n)$ 
    - AES-128
  - High quality!
    - Passes Crush test suite (more on that later)
  - Parallelizable!
    - $f$  and  $g$  only depend on  $k, n$  !
  - Sort of slow to compute
    - AES is sort of slow without special instructions (which GPUs don't have)

# Algorithmic Improvements

- Can we “make AES go faster”?
  - AES is a cryptographic algorithm, but we’re using it for PRNG
  - Can we change the algorithm for our purposes?

# AES-128 Algorithm

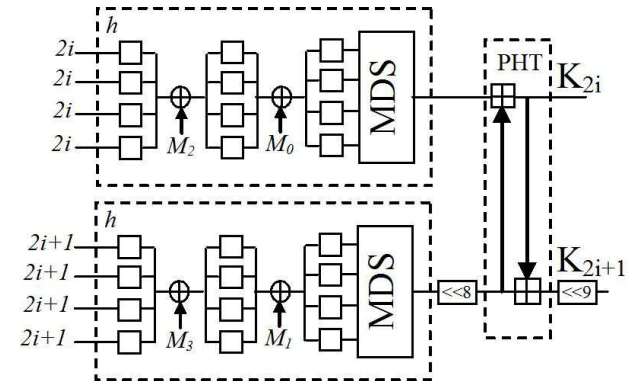
- Summary:
  - 1) Expand keys
  - 2) Add round key
  - 3) For each round (10 rounds total)
    - Substitute bytes
    - Shift rows
    - Mix columns
    - Add round key
  - 4) Final round:
    - (do everything except mix columns)

# AES-128 Algorithm

Purpose of this step is to  
hide key from attacker  
using chosen plaintext.  
Not relevant here.

- Summary:

- 1) Expand keys
- 2) Add round key
- 3) For each round (10 rounds total)
  - Substitute bytes
  - Shift rows
  - Mix columns
  - Add round key
- 4) Final round:
  - (do everything except mix columns)

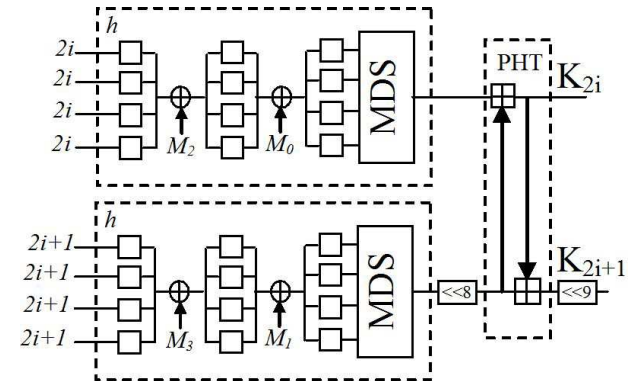


# AES-128 Algorithm

- Summary:

- 1) Expand keys
- 2) Add round key
- 3) For each round (**10 rounds total**)
  - Substitute bytes
  - Shift rows
  - Mix columns
  - Add round key
- 4) Final round:
  - (do everything except mix columns)

Purpose of this step is to hide key from attacker using chosen plaintext. Not relevant here.



Do we really need this many rounds?

Other changes?

# Key Schedule Change

- Old key schedule:

- The first  $n$  bytes of the expanded key are simply the encryption key.
- The rcon iteration value  $i$  is set to 1
- Until we have  $b$  bytes of expanded key, we do the following to generate  $n$  more bytes of expanded key:
  - We do the following to create 4 bytes of expanded key:
    - We create a 4-byte temporary variable,  $t$
    - We assign the value of the previous four bytes in the expanded key to  $t$
    - We perform the key schedule core (see above) on  $t$ , with  $i$  as the rcon iteration value
    - We increment  $i$  by 1
    - We exclusive-OR  $t$  with the four-byte block  $n$  bytes before the new expanded key. This becomes the next 4 bytes in the expanded key
  - We then do the following three times to create the next twelve bytes of expanded key:
    - We assign the value of the previous 4 bytes in the expanded key to  $t$
    - We exclusive-OR  $t$  with the four-byte block  $n$  bytes before the new expanded key. This becomes the next 4 bytes in the expanded key
  - If we are processing a 256-bit key, we do the following to generate the next 4 bytes of expanded key:
    - We assign the value of the previous 4 bytes in the expanded key to  $t$
    - We run each of the 4 bytes in  $t$  through [Rijndael's S-box](#)
    - We exclusive-OR  $t$  with the 4-byte block  $n$  bytes before the new expanded key. This becomes the next 4 bytes in the expanded key.

- New key schedule:

- $k_0 = k_B$
- $k_{i+1} = k_i + \text{constant}$ 
  - e.g. golden ratio

# AES-128 Algorithm

- Summary:
  - 1) Expand keys using simplified algorithm
  - 2) Add round key
  - 3) For each round (~~10~~ 5 rounds total)
    - Substitute bytes
    - Shift rows
    - Mix columns
    - Add round key
  - 4) Final round:
    - (do everything except mix columns)

Other simplifications possible!

# Algorithmic Improvements

- We have a good PRNG!
  - Simple transition function  $f$ 
    - Counter
  - Complicated output function  $g(k, n)$ 
    - Modified AES-128 (known as ARS-5)
  - High quality!
    - Passes Crush test suite (more on that later)
  - Parallelizable!
    - $f$  and  $g$  only depend on  $k, n$  !
  - Moderately faster to compute



# Even faster parallel PRNGs

- Use a different  $g$ , e.g.
  - Threefish cipher
    - Optimized for PRNG – known as “Threefry”
  - “Philox”
    - (see paper for details)
    - 202 GB/s on GTX580!
      - Fastest known PRNG in existence

# General Monte Carlo method

- Pseudocode:

```
for (number of trials):
```

```
    randomly pick value from a probability distribution
```

```
    perform deterministic computation on inputs
```

```
(aggregate results)
```



What about this?



Trials are  
independent



Usually so  
(e.g. with reduction)

- Can we parallelize this?

# General Monte Carlo method

- Pseudocode:

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for (number of trials):
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    perform deterministic computation on inputs
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```
(aggregate results)
```

Yes!

Trials are  
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Usually so  
(e.g. with reduction)

- Can we parallelize this?

- Yes!

- Part of cuRAND

# Summary

- Monte Carlo methods
  - Very useful in scientific simulations
  - Parallelizable because of...
- Parallelized random number generation
  - Another story of “parallel algorithm analysis”

# Credits (again)

- Parallel RNG algorithm and presentation based on:
  - “Parallel Random Numbers: As Easy as 1, 2, 3”
    - (Salmon, Moraes, Dror, Shaw) at D. E. Shaw Research