

Random number generation (in C++)

- past, present and potential future

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What is Random?

It represents an event or entity, which cannot be determined but only described probabilistically,

For example:

- Falling rain drops which hit the ground in random pattern,
- Distribution of stars within the universe is a random, and
- Babies cry in random and so on

What are the use of Random Numbers?

They are mainly used in:

- Simulations be it a numerical calculation or a cartoon game,
- Binning analog data in channel format,
- Testing a product and so on.

What is the plan of this presentation?

- Historic evolution of random numbers and their applications,
- Modern development and implementation in C++, and beyond.

How did random numbers and their applications evolve?

Since pre-historic period, random numbers were generated from dice for gambling



In bronze age gambling became unethical



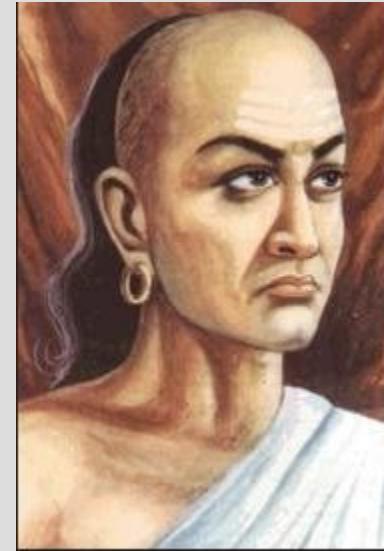
Random numbers from dice



Made Kings to give up their crowns to opponents



In 300 BC, Prof Chanakya of Takshashila University

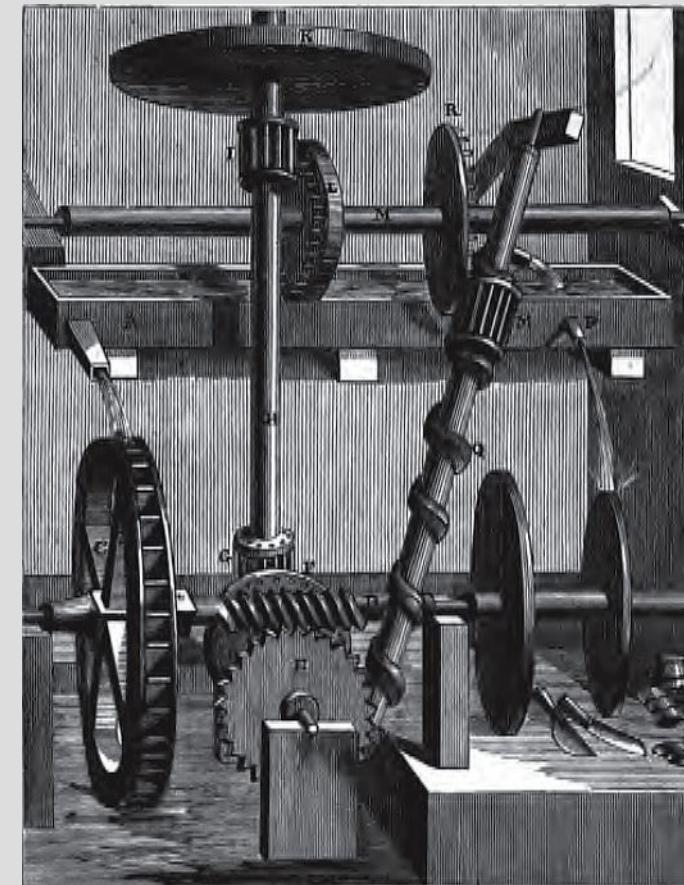


prescribed laws and taxations to regulate gambling.

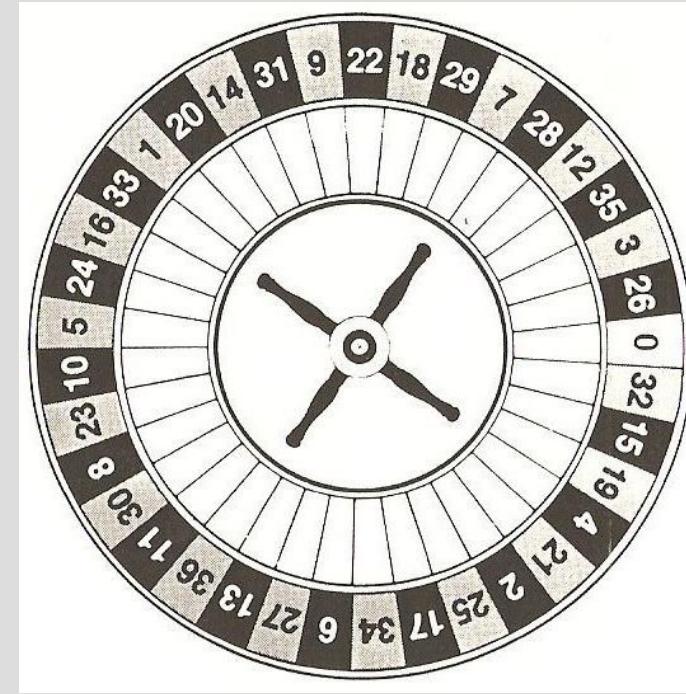
In 1650 Blaise Pascal, who introduced the computer in the form of his mechanical calculator



Designed perpetual motion experiment



That lead the way to the Roulette



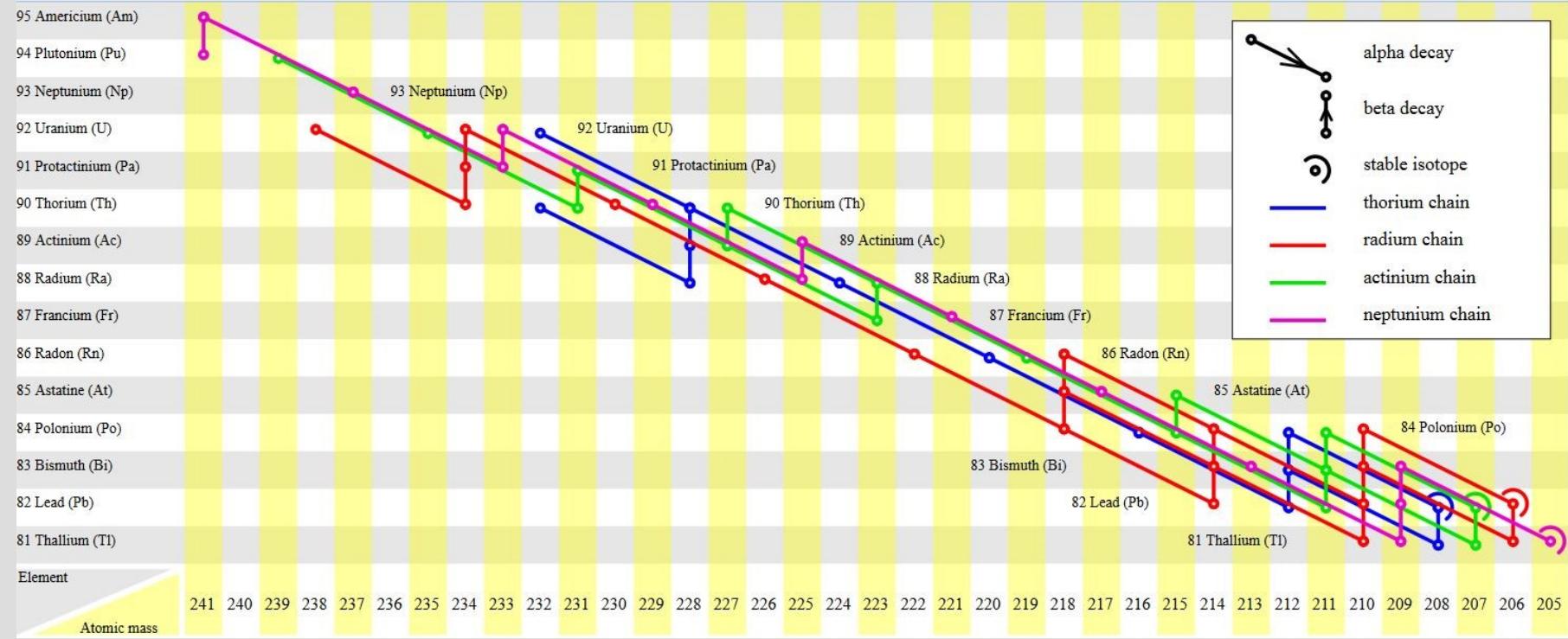
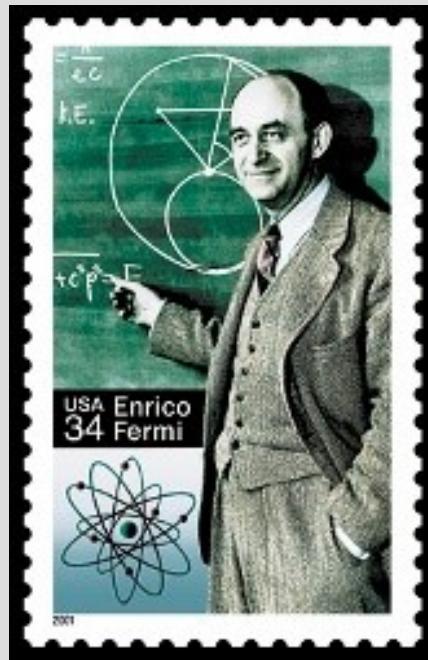


**Casino de Monte-Carlo
Monaco - MC 98000**



**Casino de Monte-Carlo
Monaco - MC 98000**

In 1930



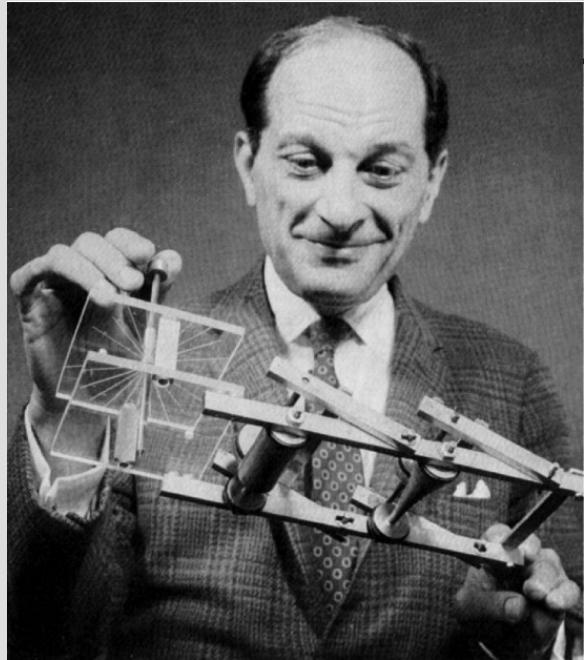
FERMIAC The Monte Carlo trolley

statistical sampling techniques



Random number generator

In 1940



Stan Ulam



While playing solitaire during his recovery from a surgery, he had thought about playing hundreds of games to estimate statistically the **probability** of a successful outcome

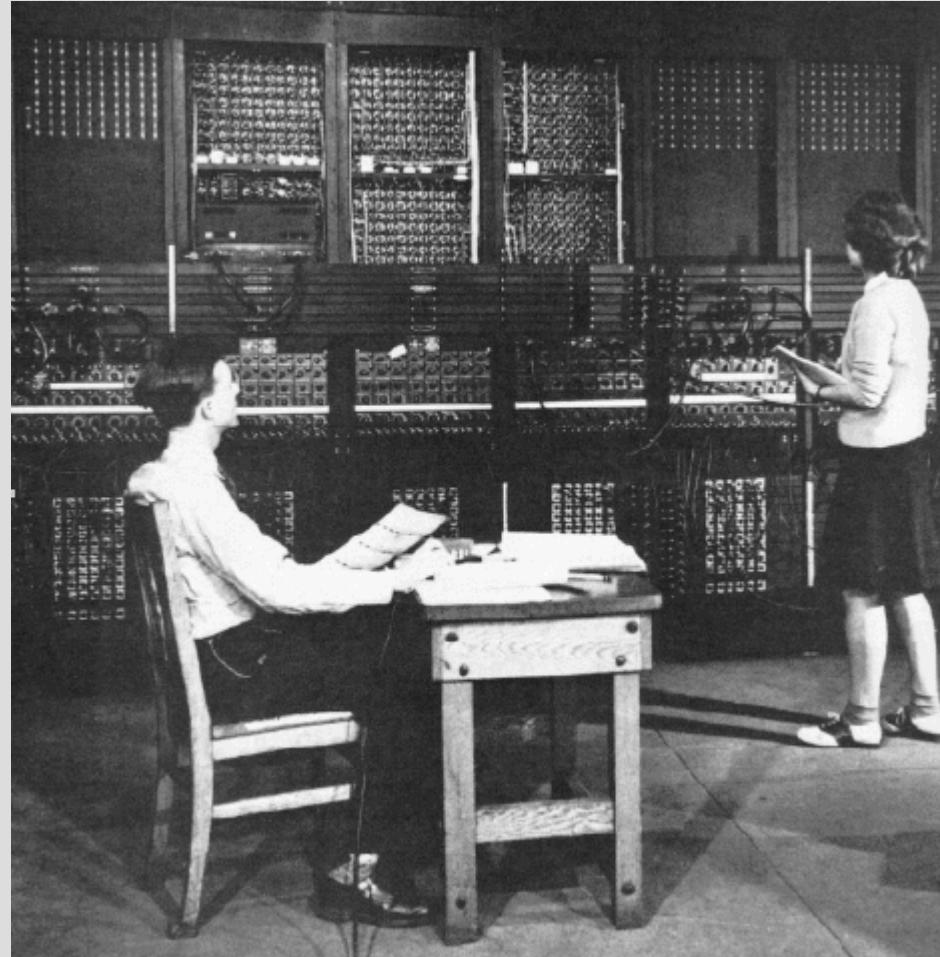
This lead to the idea of Monte Carlo Method!

ENIAC - first electronic general-purpose computer

John von Neumann



**Used Monte Carlo Method and
employed random numbers to solve
complicated problems.**



Obtained random numbers from:

- nuclear radioactivity;
- voltage fluctuation;
- Solar flare and

stored in punch cards.

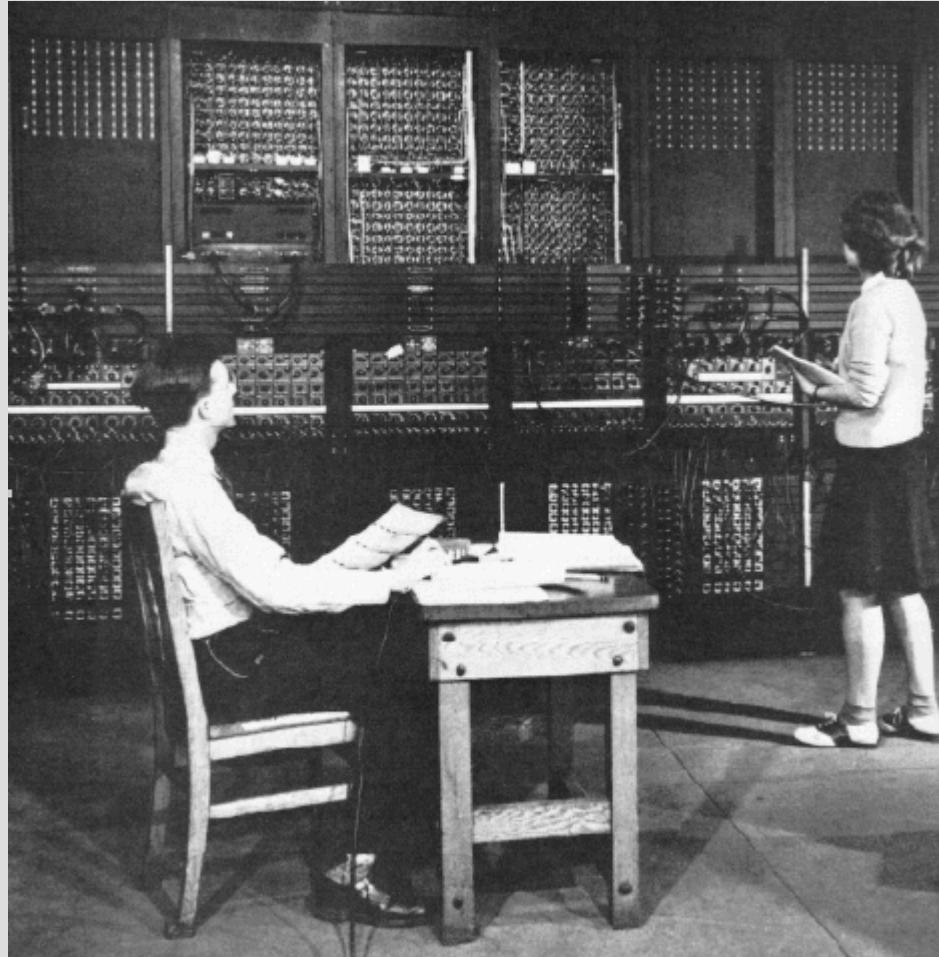
But reading from punch card was very slow!

ENIAC - first electronic general-purpose computer

John von Neumann



Used Monte Carlo Method and employed random numbers to solve complicated problems.



"Any one who considers arithmetical methods of producing random digits is, of course, in a state of sin."

So, he developed pseudo-random numbers, using the 13th century mid-square method

Let, Seed $x_0 = 0.7891$, then

$$\begin{aligned}x_0^2 &= 0.62 \textcolor{blue}{2678} \ 81 \\ \Rightarrow x_1 &= 0.2678\end{aligned}$$

$$\begin{aligned}x_1^2 &= 0.07 \textcolor{blue}{1716} \ 84 \\ \Rightarrow x_2 &= 0.1716\end{aligned}$$

$$\begin{aligned}x_2^2 &= 0.02 \textcolor{blue}{9446} \ 56 \\ \Rightarrow x_3 &= 0.9446\end{aligned}$$

He applied Monte Carlo Methods and

John von Neumann



He applied Monte Carlo Methods and

246

J. G. CHARNEY, R. FÖRTOFT, J. VON NEUMANN

John von Neumann

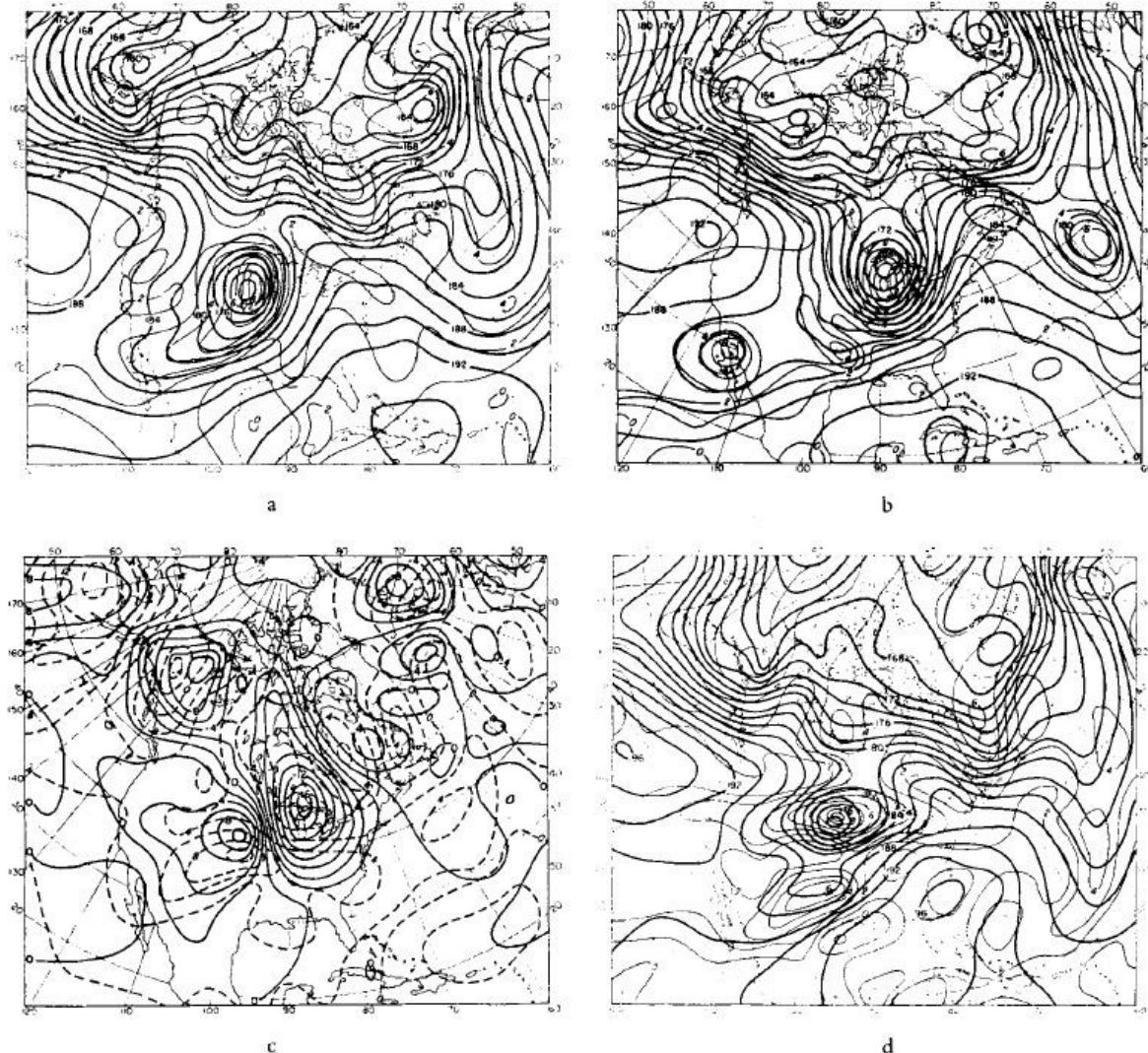
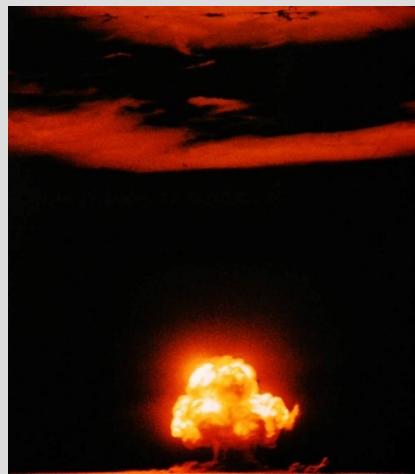
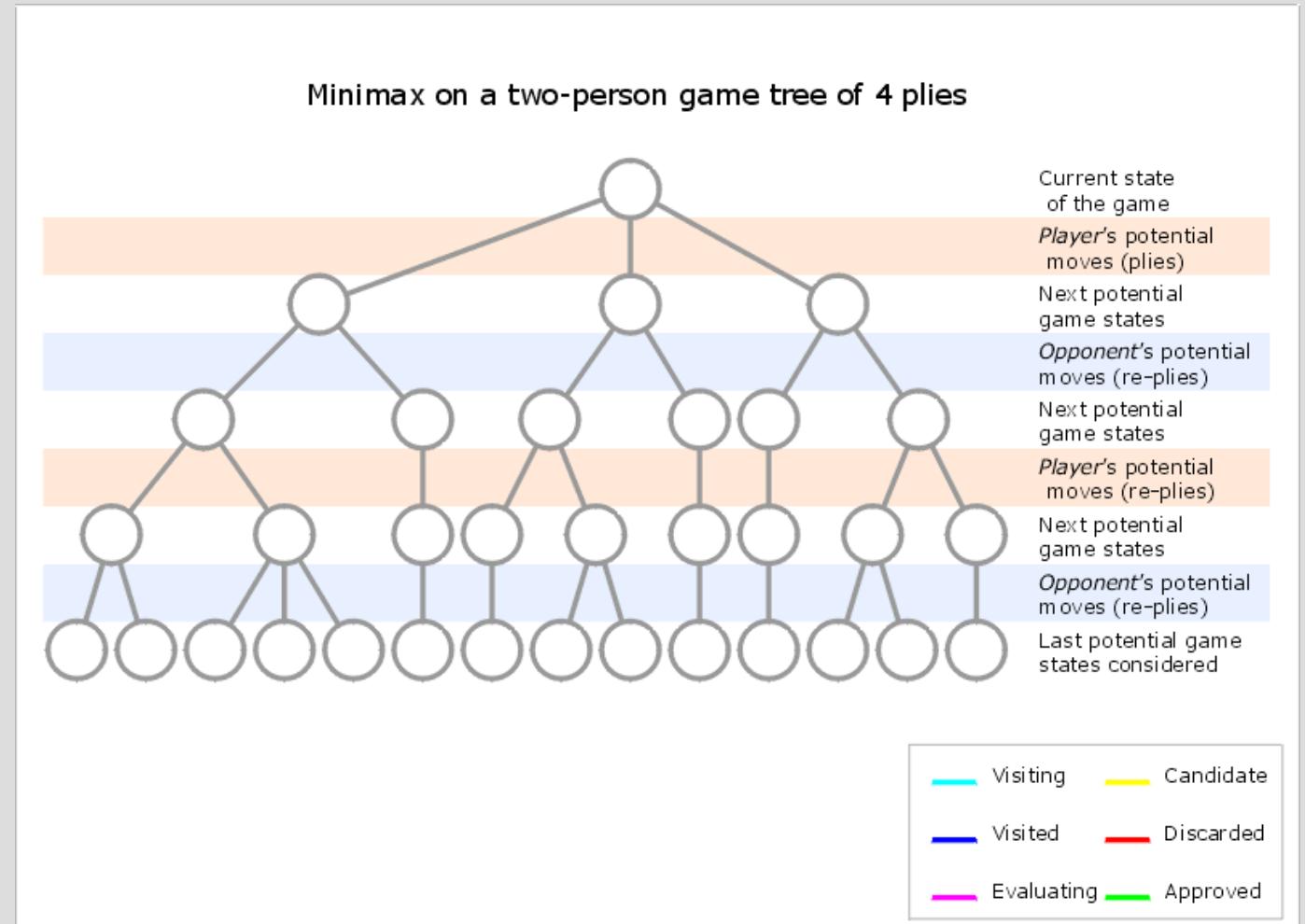
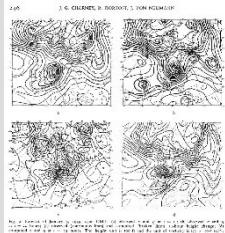
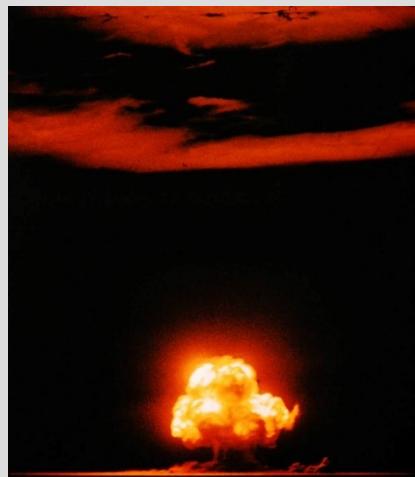


Fig. 2. Forecast of January 5, 1949, 0300 GMT: (a) observed z and η at $t = 0$; (b) observed z and η at $t = 24$ hours; (c) observed (continuous lines) and computed (broken lines) 24-hour height change; (d) computed z and η at $t = 24$ hours. The height unit is 100 ft and the unit of vorticity is $1/3 \times 10^{-4} \text{ sec}^{-1}$.

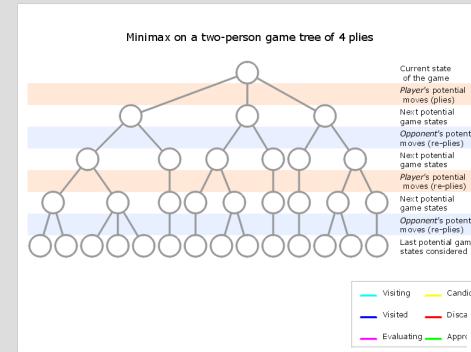
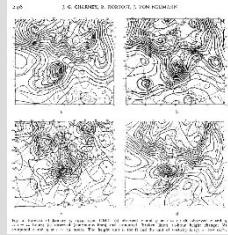
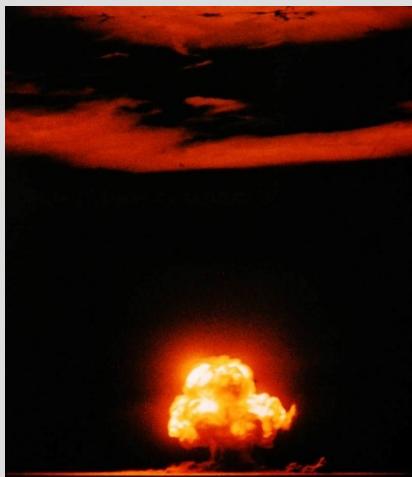
He applied Monte Carlo Methods and

John von Neumann

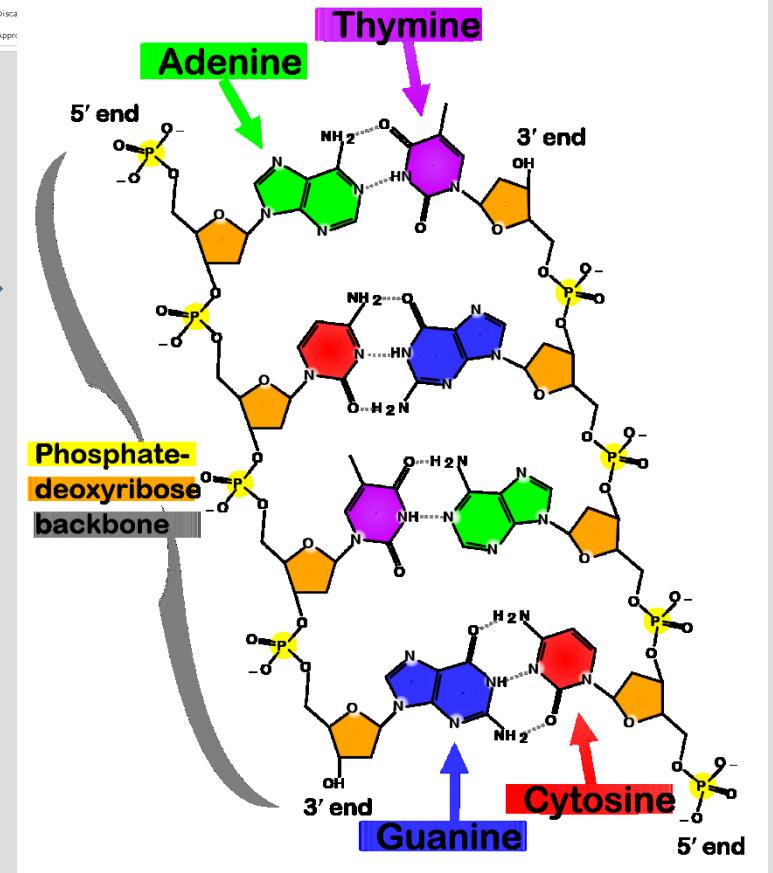


He applied Monte Carlo Methods and

John von Neumann

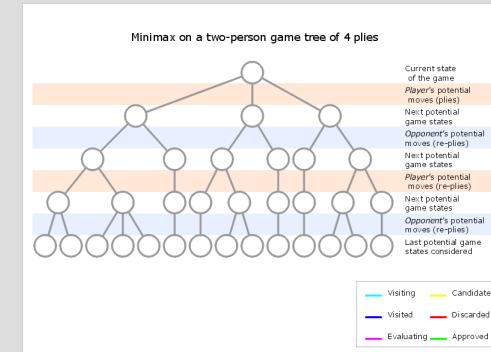
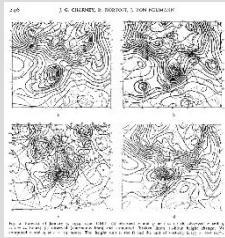


replicators

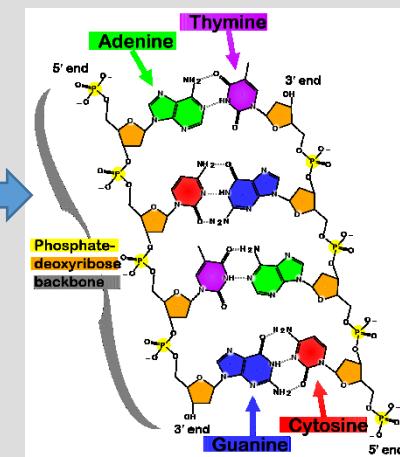


He applied Monte Carlo Methods and

John von Neumann



replicators



msbl
ast.exe I just w
ant to say LOVE
YOU SAN!! billy
gates why do you
make this possi
ble ? Stop makin
g money and fix
your software!!
♦ δ♦ H △
0 00 00-6D 73 62 6C
0 6A 75-73 74 20 77
9 20 4C-4F 56 45 20
0 62 69-6C 6C 79 20
0 64 6F-20 79 6F 75
3 20 70-6F 73 73 69
0 20 6D-61 6B 69 6E
E 64 20-66 69 78 20
7 61 72-65 21 21 00
0 00 00-7F 00 00 00
0 00 00-01 00 01 00
0 00 00-00 00 00 46
á@ C C9 11-9F E8 08 00
0 00 03-10 00 00 00
3 00 00-01 00 04 00
pv o ðv o ðv o ♦

In 1957

Fortran (Formula Translating System)

Multiplicative Congruential Generator (MCG)

$$X_n = (aX_{n-1}) \bmod m$$

```
RANDOM = MOD(A*SEED,M)
PRINT*, RANDOM
SEED = RANDOM
```

Linear congruential generator

$$X_n = (aX_{n-1} + b) \bmod m$$

In 1972

C

```
static unsigned long int next = 1;

int rand(void) // RAND_MAX assumed to be 32767
{
    static const unsigned long int a = 1103515245;
    static const unsigned short b = 12345;
    next = next * a + b;
    return (unsigned int)(next/65536) % 32768;
}

void srand(unsigned int seed)
{
    next = seed;
}
```

In 1972

C

```
static unsigned long int next = 1;

int rand(void) // RAND_MAX assumed to be 32767
{
    static const unsigned long int a = 1103515245;
    static const unsigned short b = 12345;
    next = next * a + b;
    return (unsigned int)(next/65536) % 32768;
}

void srand(unsigned int seed)
{
    next = seed;
}
```

```

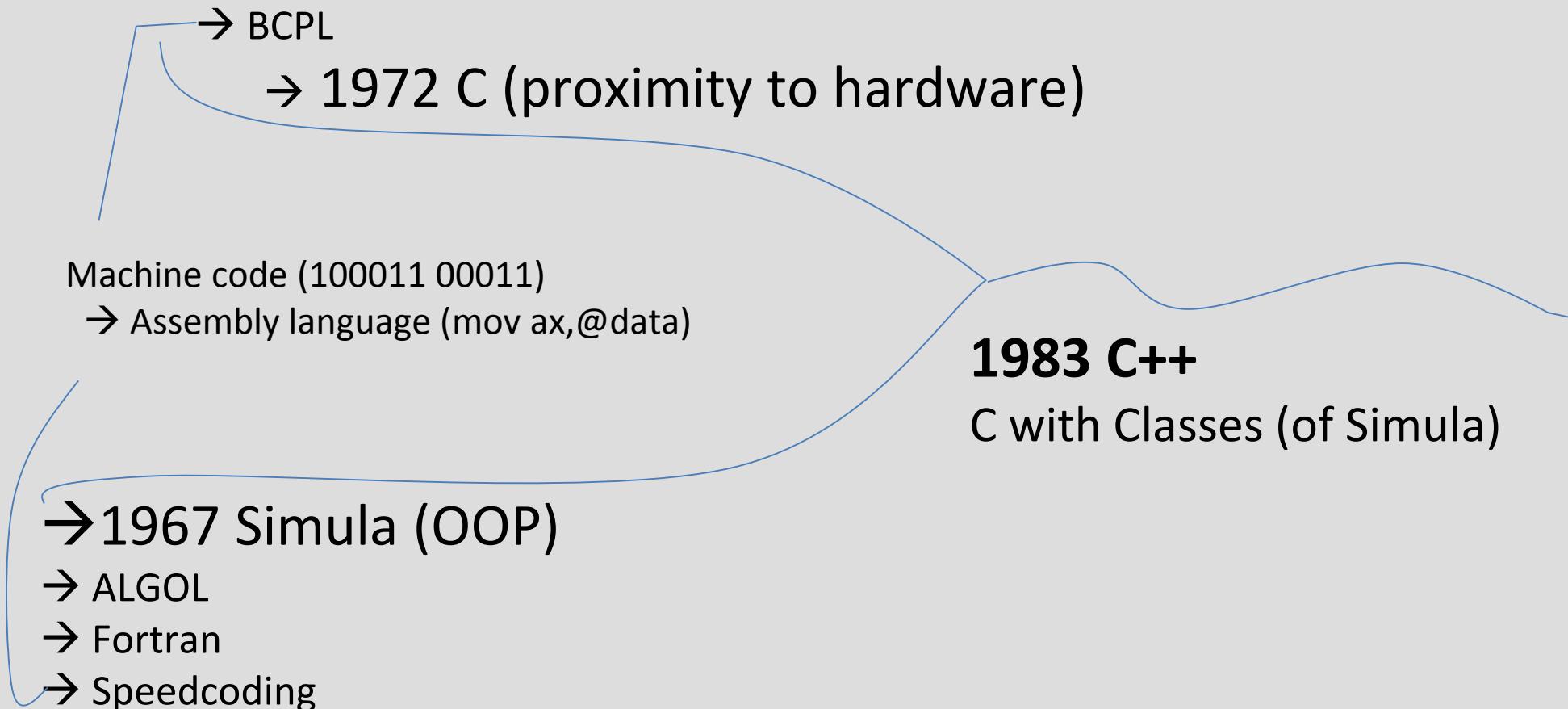
#include <cstdlib>
#include <cstdio>
#include <ctime>

int main()
{
    srand(time(NULL));
    printf("Random numbers:\n");
    float random = 0;
    for (int i = 0; i < 10; ++i) {
        random = (float) rand() / RAND_MAX;
        printf("%f\n", random);
    }
}

```

Random numbers:

0.003143
0.569353
0.033021
0.545427
0.627430
0.035096
0.817377
0.055635
0.282266
0.187628



In 1983

C++

```
int rand (void);
```

```
void srand (unsigned int seed);
```

```
#define RAND_MAX 32767
```

```
std::random_shuffle (data.begin(), data.end());
```

Initial data is {1, 2, 3, 4, 5, 6, 7, 8, 9, 10}

Shuffled data is {6, 10, 5, 2, 3, 7, 9, 8, 1, 4}

YouTube GB random number generator



0:13 / 11:38

Random Numbers - Numberphile

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Random Numbers - Numberphile

<http://www.youtube.com/watch?v=SxP30euw3-0>



Improvements to TR1's Facility for Random Number Generation

Document #: WG21/N1933 = J16/06-0003
Date: 2006-02-23
Revises: None
Project: Programming Language C++
Reference: ISO/IEC IS 14882:2003(E)
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If the numbers are not random, they are at least higgledy-piggledy.

— GEORGE MARSAGLIA

I cannot do it without computers.

— WILLIAM SHAKESPEARE

In 2011

C++ 11

<random> (26.5 Random number generation N3797)

Distribution(Engine)

In 2011

C++ 11

<random> (26.5 Random number generation N3797)

Random = Distribution(Engine)

Engines:

`linear_congruential_engine`
`subtract_with_carry_engine`
`mersenne_twister_engine`

Engines Adaptors:

`discard_block_engine`
`independent_bits_engine`
`shuffle_order_engine`

True random number generator:

`random_device`

Distributions:

`Uniform distributions`
`Normal distributions`
`Bernoulli distributions`
`Rate-based distributions`
`Piecewise distributions`
`Canonical numbers`

linear_congruential_engine

$$X_n = (aX_{n-1} + b) \bmod m$$

minstd_rand0: $a = 16807$; $b = 0$; $m = 2, 147, 483, 647$

minstd_rand: $a = 48271$; $b = 0$; $m = 2, 147, 483, 647$

linear_congruential_engine

$$X_n = (aX_{n-1} + b) \bmod m$$

`minstd_rand0`: $a = 16807$; $b = 0$; $m = 2, 147, 483, 647$

`minstd_rand`: $a = 48271$; $b = 0$; $m = 2, 147, 483, 647$

subtract_with_carry_engine

1991 George Marsaglia and Arif Zaman

Linear Congruential algorithm => $X_n = (aX_{n-1} + b) \text{ mod } m = f(X_{n-1})$

Lagged Fibonacci algorithm => $f(X_{n-1}, X_{n-2}) \Rightarrow f(X_{n-S}, X_{n-R}); \text{ where, } S < R < 0$

$$X_n = (X_{n-S} - X_{n-R} - cy_{n-1}) \text{ mod } m$$

$$\text{where, } cy_n = (X_{n-S} - X_{n-R} - cy_{n-1} < 0) ? 1 : 0;$$

ranlux24_base: 24-bit number S = 10; R = 24;

ranlux48_base: 48-bit number S = 5; R = 12;

mersenne_twister_engine

1997 Makoto Matsumoto and Takuji Nishimura
“Twisted Generalized Feedback Shift Register”

Period length = Mersenne prime = $2^{19937}-1$

MT19937 uses a 32-bit word length

MT19937-64 uses a 64-bit word length

random_device

Generates random numbers from hardware where available

```
CRTIMP2_PURE unsigned int __CLRCALL_PURE_OR_CDECL _Random_device();

unsigned int operator()()
{
    return (_Random_device());
}

std::random_device rd;

std::default_random_engine e1(rd());
```

Required behaviour: The 10000th consecutive invocation of a default-constructed object of type mt19937 shall produce the value 4123659995.

```
void my_random_generator::check_mt19937()
{
    mt19937 engine;
    static long long random_number = 0;

    for(int i =0; i != 9999; ++i) engine();

    random_number = engine();

    // If the implementation is correct then the 10000th consecutive
    // invocation of a default-constructed object of type mt19937
    // shall produce the value 4,123,659,995, ref. C++11 ISO statement.

    if(random_number == 4123659995)
        cout << "\n Note:\n\t The pseudorandom number generator\n\t "
            << "Mersenne twister: MT19937 has been tested\n\t and "
            << "it shows it is implemented properly.\n";
    else
        cout << "\nWarning:\n\t "
            << "The pseudorandom number generator\n\t "
            << "Mersenne twister: MT19937 has been "
            << "tested\n\t and it shows it is NOT "
            << "implemented properly.\n";
}
```

```
// A Uniform Distribution based on default_random_engine:  
// Using Bind function:  
  
auto dist = bind(uniform_real_distribution<double>{0.0, 1.0}, default_random_engine{});  
  
Random = dist();
```

```
class uniform_dist {  
  
public:  
  
    double operator()() { return uniform(engine); }  
  
    uniform_dist() :uniform{ 0.0, 1.0 } {}  
  
    void discard(unsigned long long z) { engine.discard(z); }  
  
    void discard_distribution(unsigned long long z)  
    {  
        for (auto i = z; i != 0; --i)  
            uniform(engine);  
    }  
  
    uniform_dist(double low, double high) : uniform(low, high) {}  
  
private:  
  
    default_random_engine engine;  
  
    uniform_real_distribution <double> uniform;  
};
```

```
// A Normal distribution based on
// Mersenne Twister engine:

class normal_dist {
public:
    double operator()() { return normal(engine); }

    normal_dist() :normal {0.0, 1.0} {}

    void discard(unsigned long long z) { engine.discard(z); }

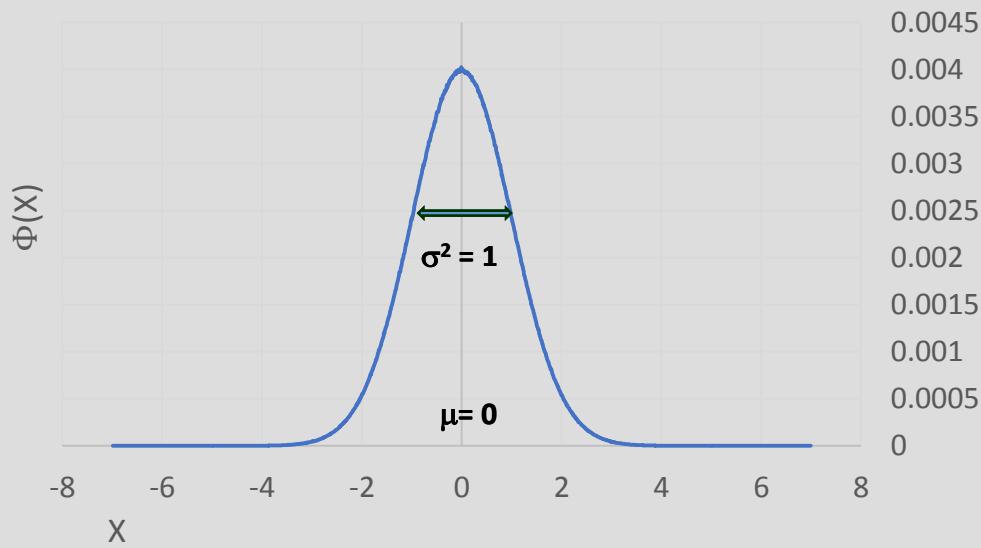
    void discard_distribution(unsigned long long z)
    {
        for (auto i = z; i != 0; --i)
            normal(engine);
    }

    normal_dist(double mean, double std_dev) : normal( mean, std_dev ) {}

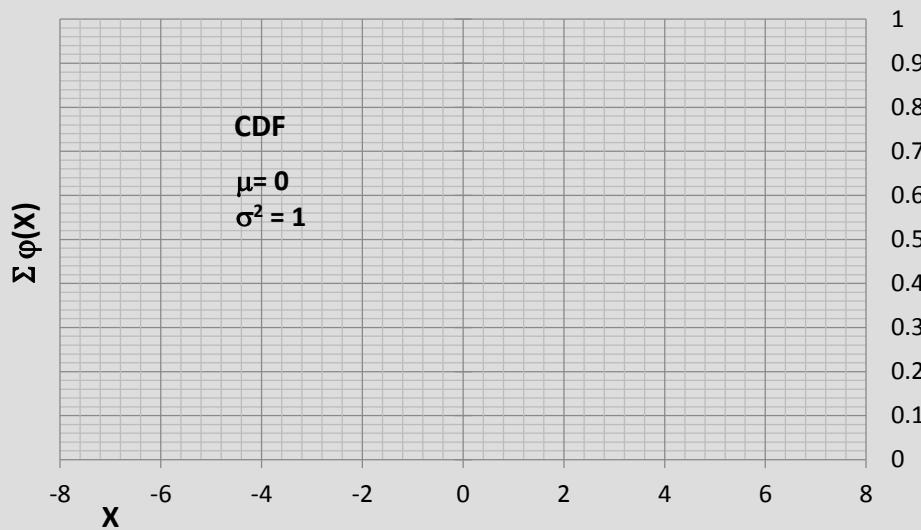
private:
    mt19937 engine;

    normal_distribution<double> normal;
};
```

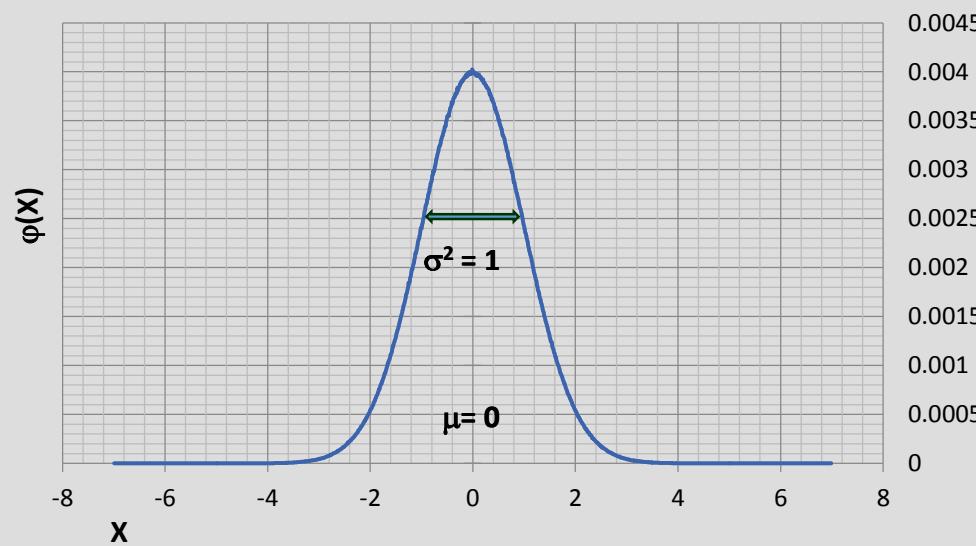
Normal distribution



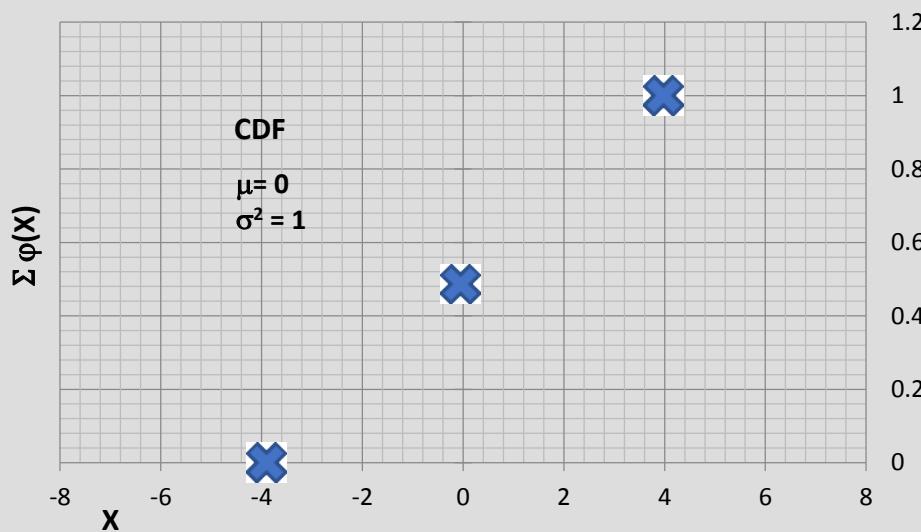
Cumulative normal distribution



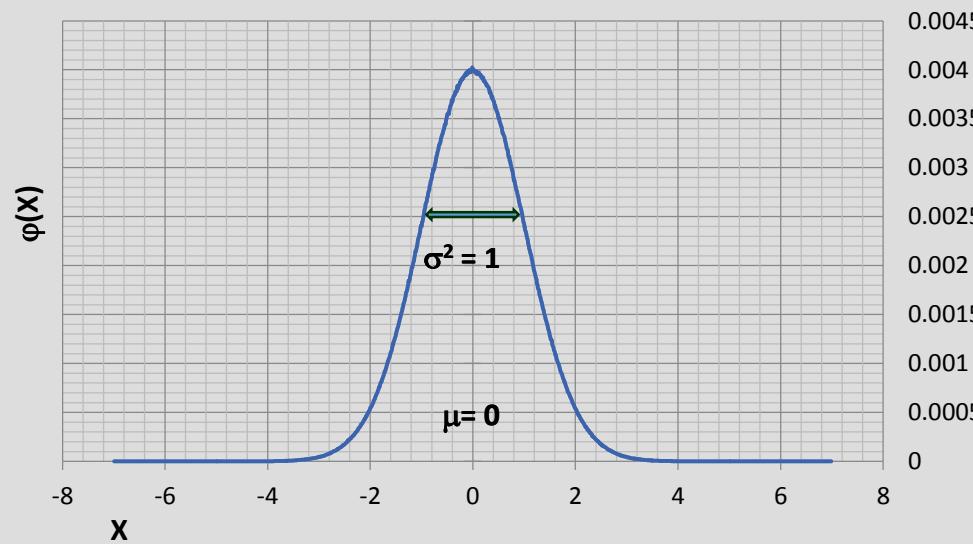
Probability normal distribution



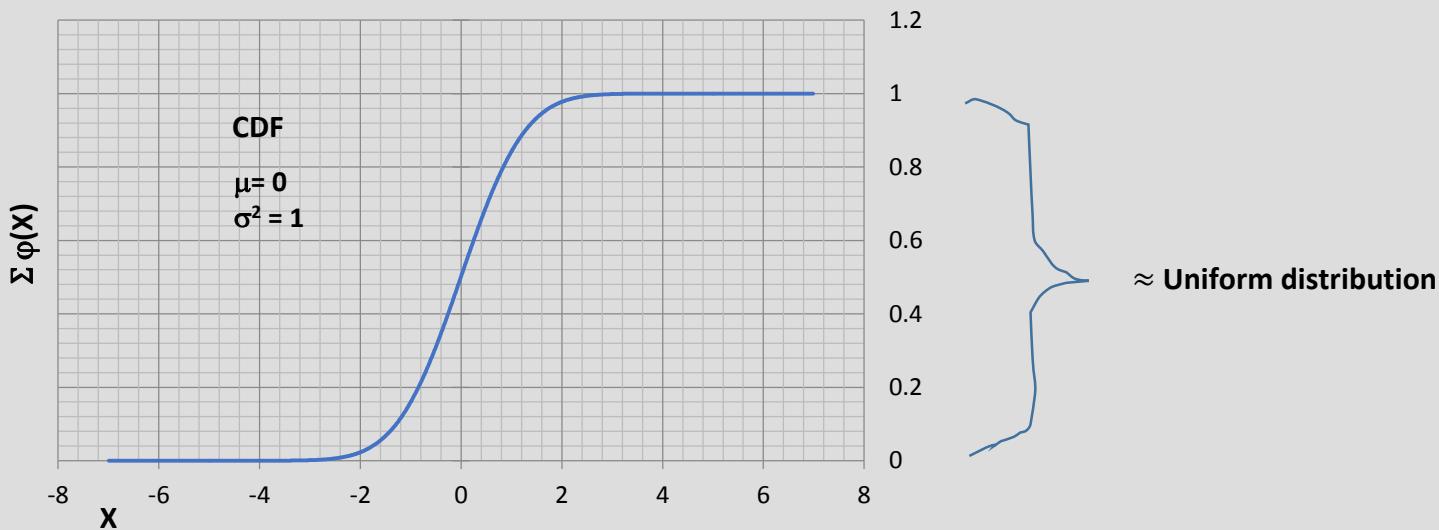
Cumulative normal distribution



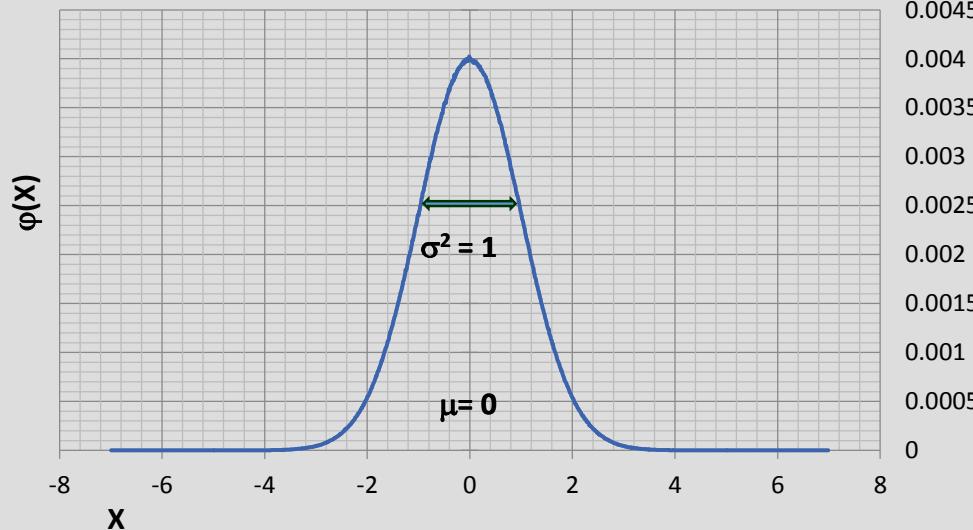
Probability normal distribution



Cumulative normal distribution



Probability normal distribution



Pseudo-code algorithm for rational approximation

The algorithm below assumes p is the input and x is the output.

Coefficients in rational approximations.

```
a(1) <- -3.969683028665376e+01
a(2) <- 2.209460984245205e+02
a(3) <- -2.759285104469687e+02
a(4) <- 1.383577518672690e+02
a(5) <- -3.066479806614716e+01
a(6) <- 2.506628277459239e+00

b(1) <- -5.447609879822406e+01
b(2) <- 1.615858368580409e+02
b(3) <- -1.556989798598866e+02
b(4) <- 6.680131188771972e+01
b(5) <- -1.328068155288572e+01

c(1) <- -7.784894002430293e-03
c(2) <- -3.223964580411365e-01
c(3) <- -2.400758277161838e+00
c(4) <- -2.549732539343734e+00
c(5) <- 4.374664141464968e+00
c(6) <- 2.938163982698783e+00

d(1) <- 7.784695709041462e-03
d(2) <- 3.224671290700398e-01
d(3) <- 2.445134137142996e+00
d(4) <- 3.754408661907416e+00
```

Define break-points.

```
p_low <- 0.02425
p_high <- 1 - p_low
```

Rational approximation for lower region.

```
if 0 < p < p_low
    q <- sqrt(-2*log(p))
    x <- (((((c(1)*q+c(2))*q+c(3))*q+c(4))*q+c(5))*q+c(6)) /
        (((d(1)*q+d(2))*q+d(3))*q+d(4))*q+1)
endif
```

Rational approximation for central region.

```
if p_low <= p <= p_high
    q <- p - 0.5
    r <- q*q
    x <- (((((a(1)*r+a(2))*r+a(3))*r+a(4))*r+a(5))*r+a(6))*q /
        (((((b(1)*r+b(2))*r+b(3))*r+b(4))*r+b(5))*r+1)
endif
```

Rational approximation for upper region.

```
if p_high < p < 1
    q <- sqrt(-2*log(1-p))
    x <- -(((c(1)*q+c(2))*q+c(3))*q+c(4))*q+c(5))*q+c(6)) /
        (((d(1)*q+d(2))*q+d(3))*q+d(4))*q+1)
endif
```

Please avoid coding random number generators or distributions!

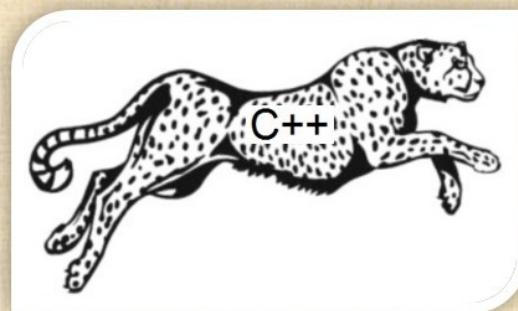
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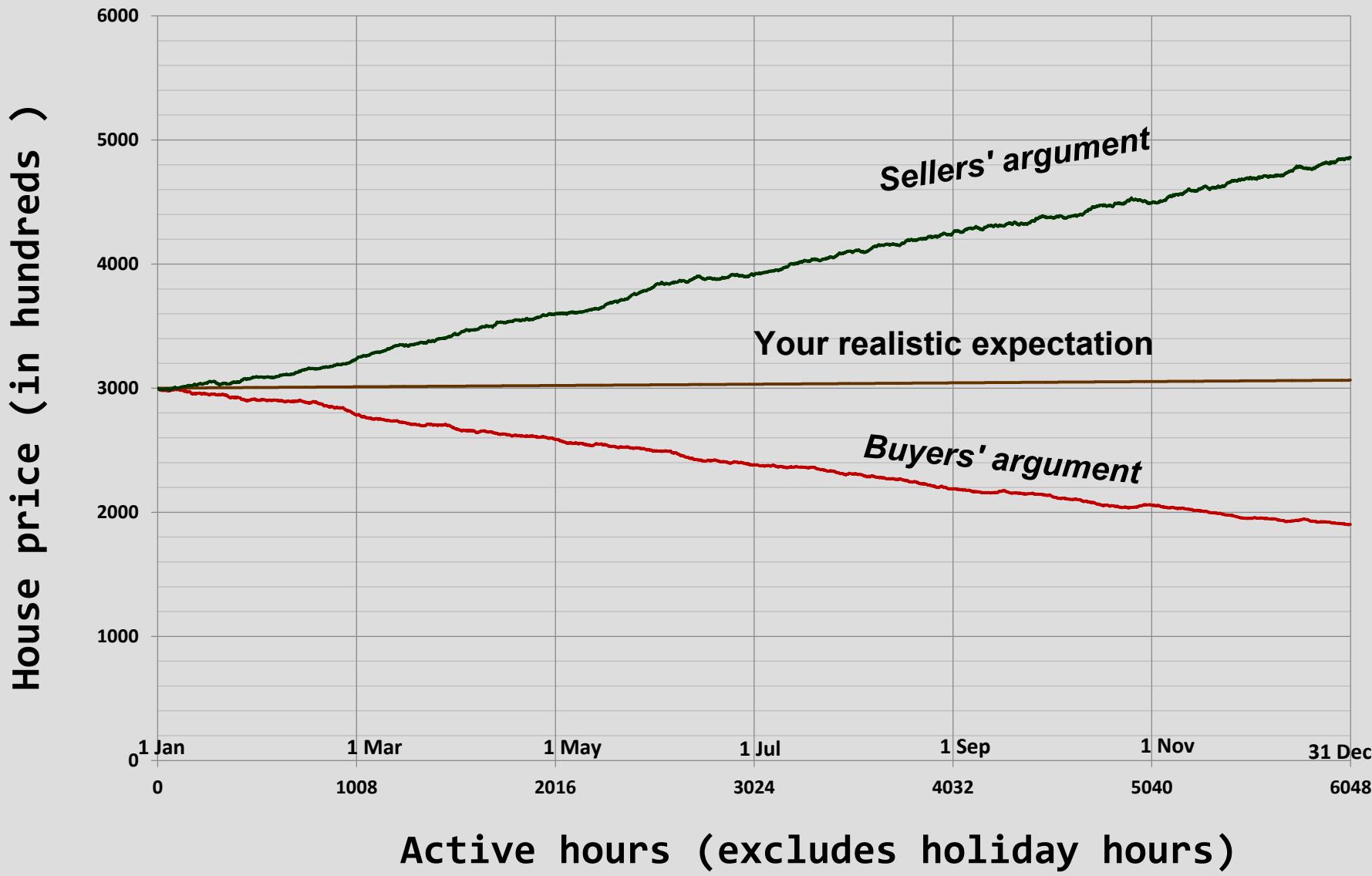


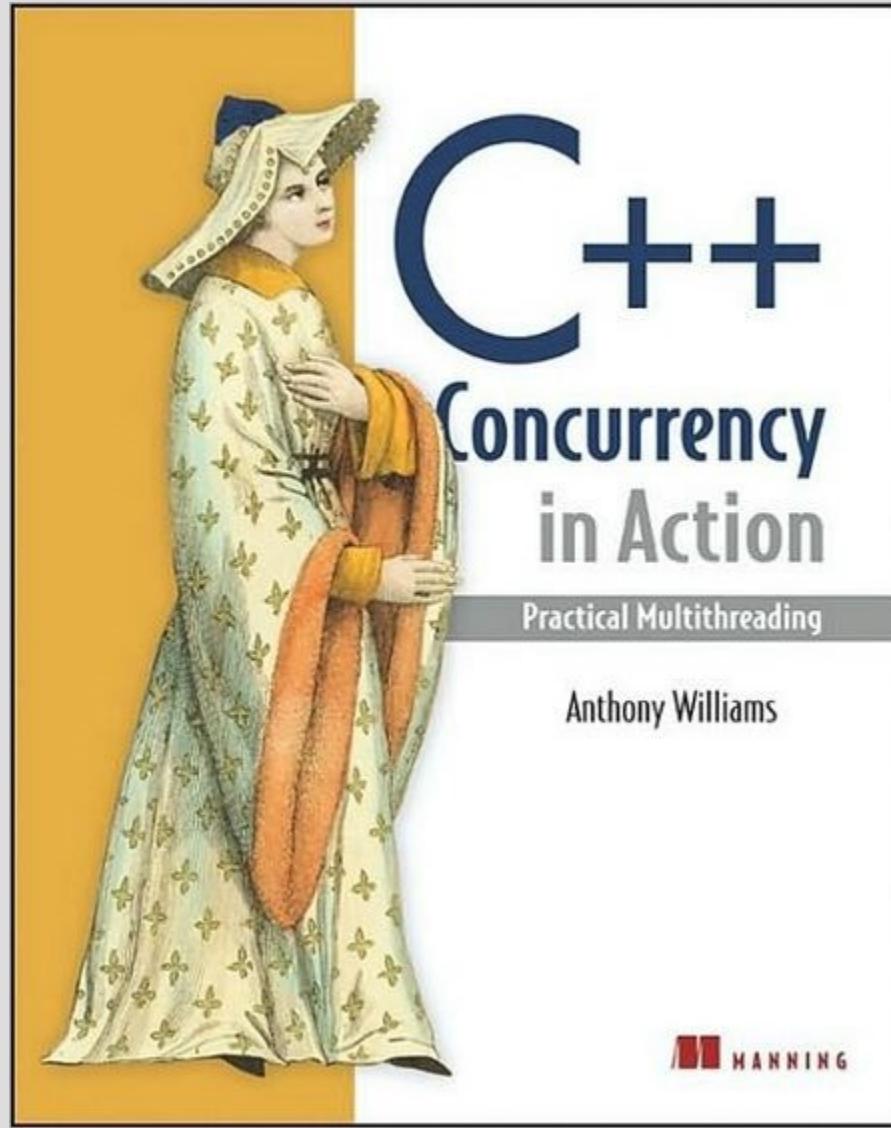
N.S.Pattabi Raman

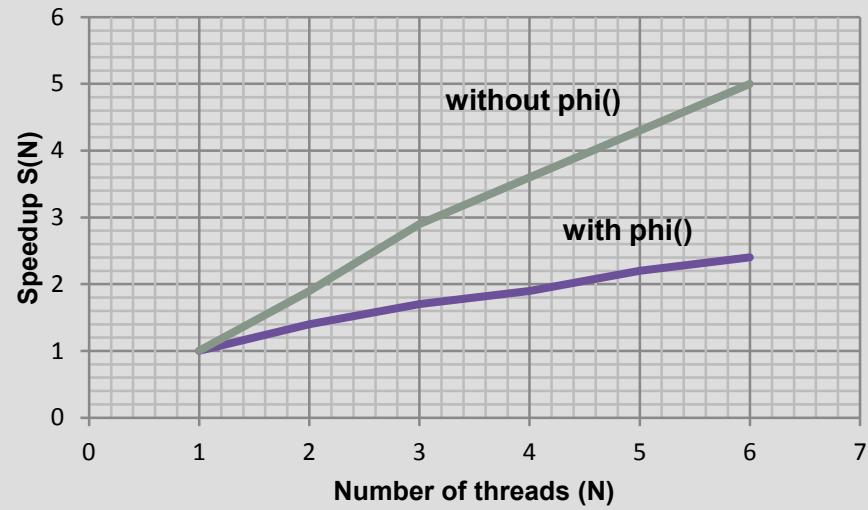
λsl www.numericalsolution.co.uk

$$V_t = V_{t-dt} (1 + r.dt + \sigma.\phi().\sqrt{dt}),$$

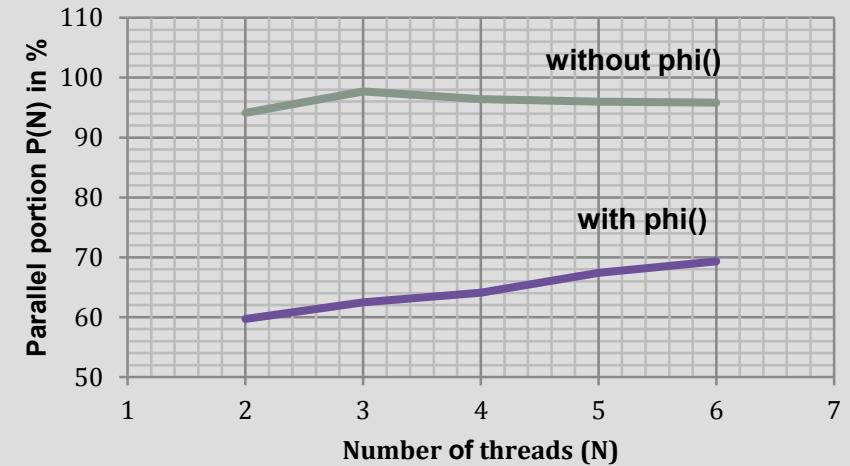
```
for(i = 1; i < T; ++i) {
    dX = phi()*sqrt_dt;
    dS = S*(r_dt + sigma*dX);
    S += dS;
    data[i] = S;
}
```



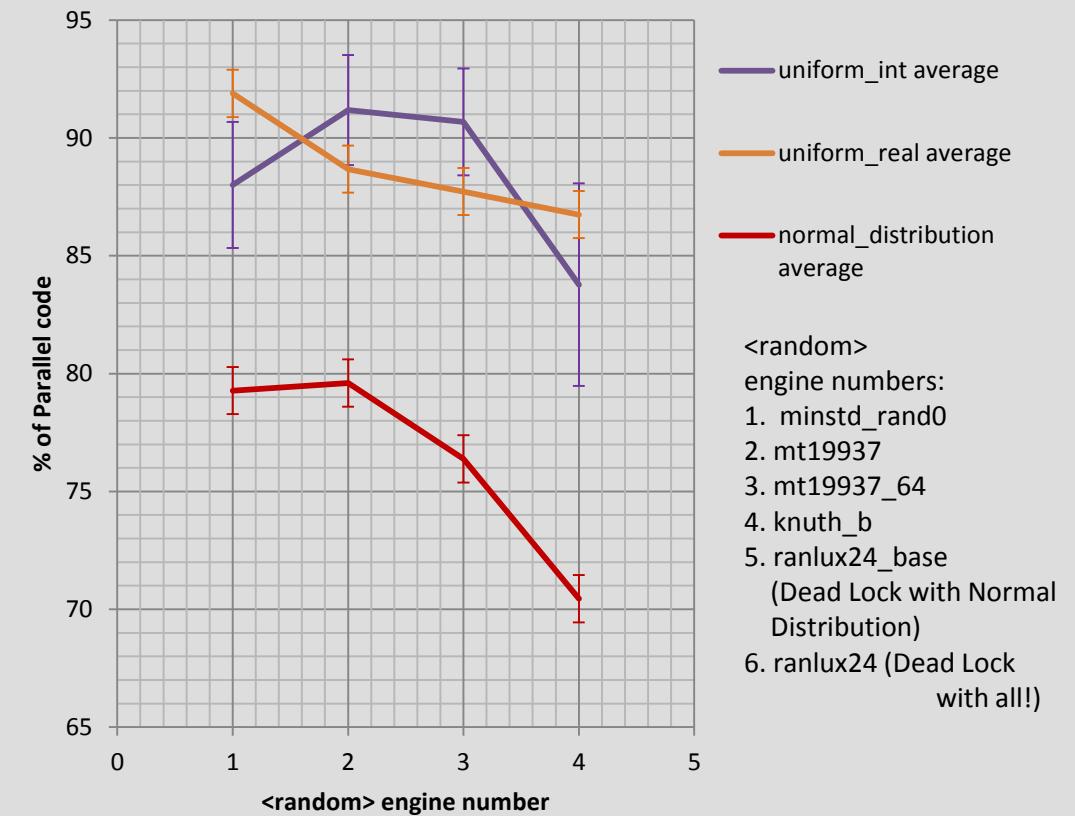




$$\text{Amdahl's law} \Rightarrow P = \frac{(1-S)N}{S(1-N)}$$



<random> engine	uniform_int		uniform_real		normal_distribution	
	average	stdev	average	stdev	average	stdev
1. Linear congruential: minstd_rand0	88.00	2.67	91.89	2.97	79.28	3.60
2. Mersenne twister: mt19937	91.18	2.34	88.67	3.47	79.60	2.61
3. Mersenne twister: mt19937_64	90.68	2.27	87.73	3.10	76.38	3.04
4. Shuffle order: knuth_b	83.77	4.29	86.75	6.12	70.45	5.71
5. Subtract with carry: ranlux24_base	88.09	2.14	83.62	4.74	Dead Lock!	
6. Discard block: ranlux24	Dead Lock!		Dead Lock!		Dead Lock!	



Document no: N3397=12-0087
Date: 2012-08-13
Project: Programming Language C++
Reply to: Roger Orr
rogero@howzatt.demon.co.uk

Spring 2013 JTC1/SC22/WG21 C++ Standards Committee Meeting

Bristol, UK, April 15 – 20, 2013

(note: this is 6 days: Mon - Sat)

The meeting venue and host hotel is the **Marriott Hotel, Bristol City Centre**.
2 Lower Castle Street, Old Market, Bristol, England BS1 3AD
<http://www.marriott.co.uk/hotels/travel/brsdt-bristol-marriott-hotel-city-centre/>

Thanks to:

Hans Boehm, Lawrence Crowl, Mike Giles, Peter Jäckel, Stephan T. Lavavej, Nick McLaren
Alisdair Meredith, Roger Orr, I.M. Sobol', Herb Sutter, Jonathan Wakely, Michael Wong and more

```
template<class UIntType, size_t w, ...>
class mersenne_twister_engine
{
public:
    ....
    ....
    ....
    explicit mersenne_twister_engine(result_type value = default_seed);
    template<class Sseq> explicit mersenne_twister_engine(Sseq& q);
    void seed(result_type value = default_seed);
    template<class Sseq> void seed(Sseq& q);

    // generating functions
    result_type operator()() const; // if const method then it would be safe for concurrency.
    void discard(unsigned long long z);
};
```

Thanks to:

Hans Boehm, Lawrence Crowl, Mike Giles, Peter Jäckel, Stephan T. Lavavej, Nick McLaren
Alisdair Meredith, Roger Orr, I.M. Sobol', Herb Sutter, Jonathan Wakely, Michael Wong and more

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    void seed(result_type value = default_seed);
    template<class Sseq> void seed(Sseq& q);

    // generating functions
    result_type operator()(); // As it is not const method, it is not safe for concurrency!
    void discard(unsigned long long z);
};
```

Thanks to:

Hans Boehm, Lawrence Crowl, Mike Giles, Peter Jäckel, Stephan T. Lavavej, Nick McLaren
Alisdair Meredith, Roger Orr, I.M. Sobol', Herb Sutter, Jonathan Wakely, Michael Wong and more

- Declare random number generator as **thread_local**, then each thread can have independent copy of the random object.
- But, it will provide same set of numbers in all threads, so that is multiplication of same data, therefore, **that cannot add to statistics!**

Thanks to:

Hans Boehm, Lawrence Crowl, Mike Giles, Peter Jäckel, Stephan T. Lavavej, Nick McLaren
Alisdair Meredith, Roger Orr, I.M. Sobol', Herb Sutter, Jonathan Wakely, Michael Wong and more

- **Construct independent engines for each thread and set different range!**

```
// A Normal distribution based on
// Mersenne Twister engine:

class normal_dist {
public:
    double operator()() { return normal(engine); }

    normal_dist() :normal {0.0, 1.0} {}

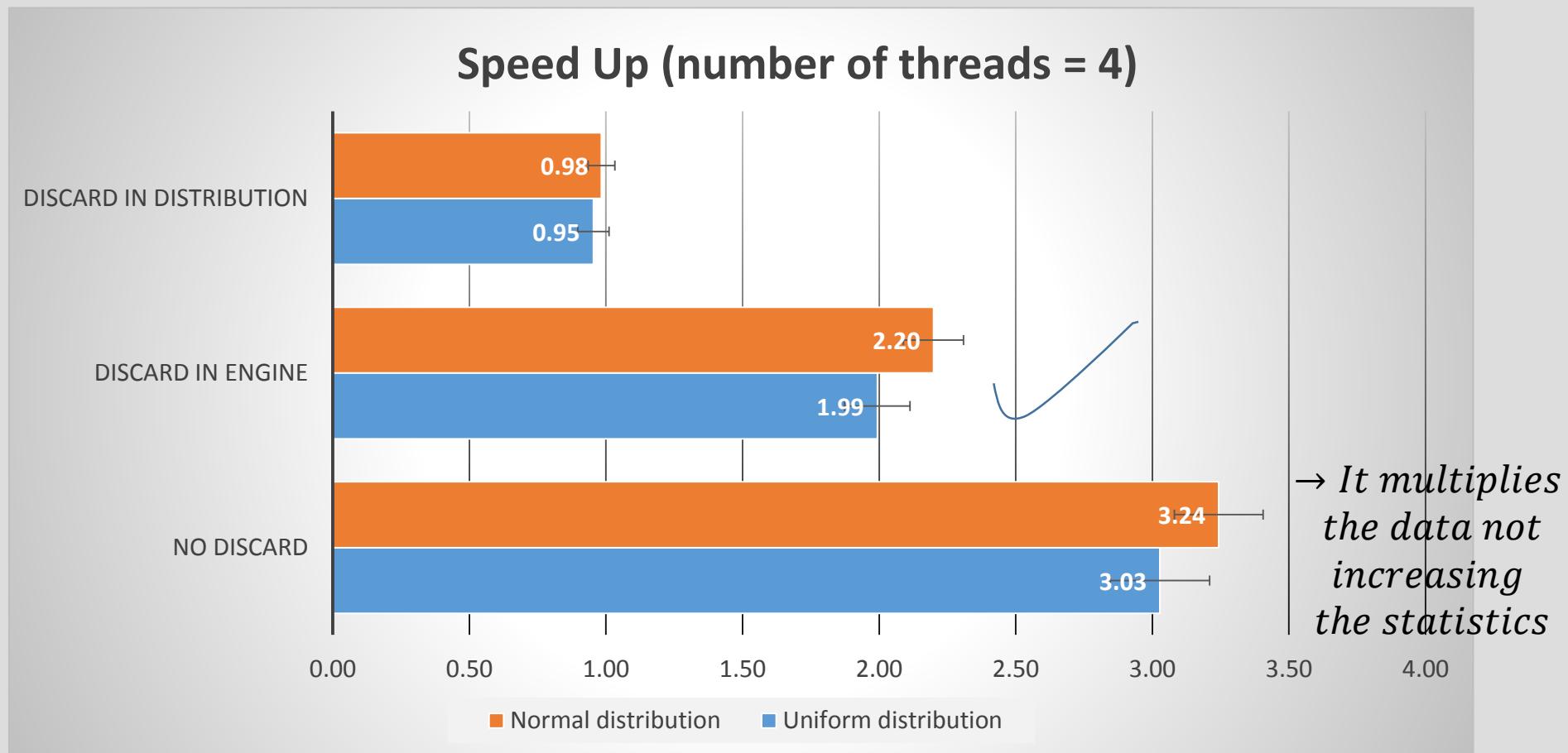
    {
        for (auto i = z; i != 0; --i)
            normal(engine);
    }

    normal_dist(double mean, double std_dev) : normal( mean, std_dev ) {}

private:
    mt19937 engine;

    normal_distribution<double> normal;
};


```

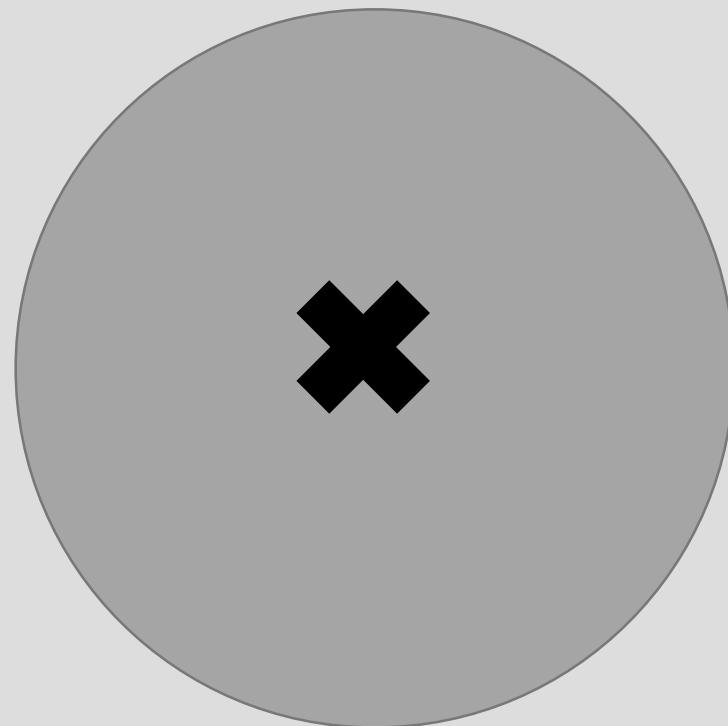


Complexity of 'discard(number)', i.e., performance time should be reduced, so number theoreticians can help!

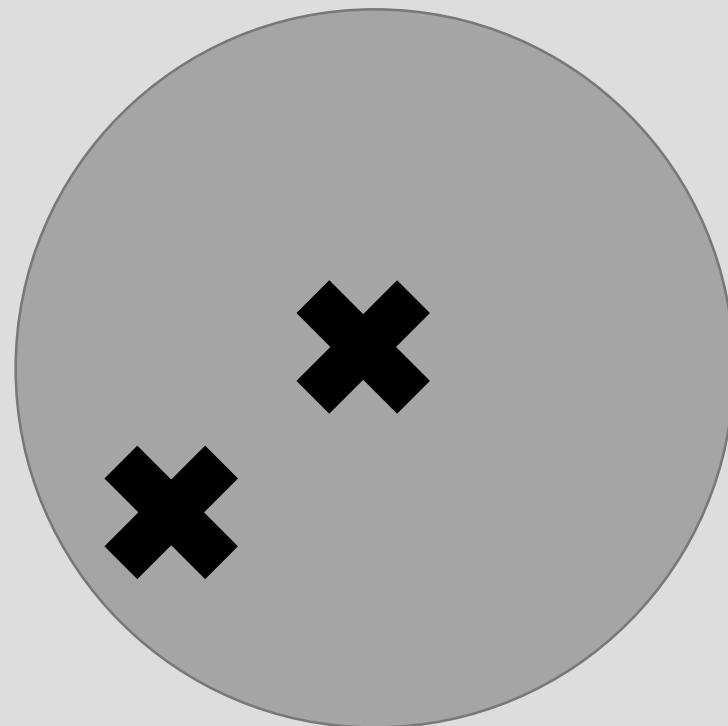
Quasi-Random Numbers

- What are they?
- How do they differ from Pseudo-random numbers?
- Where do they help?

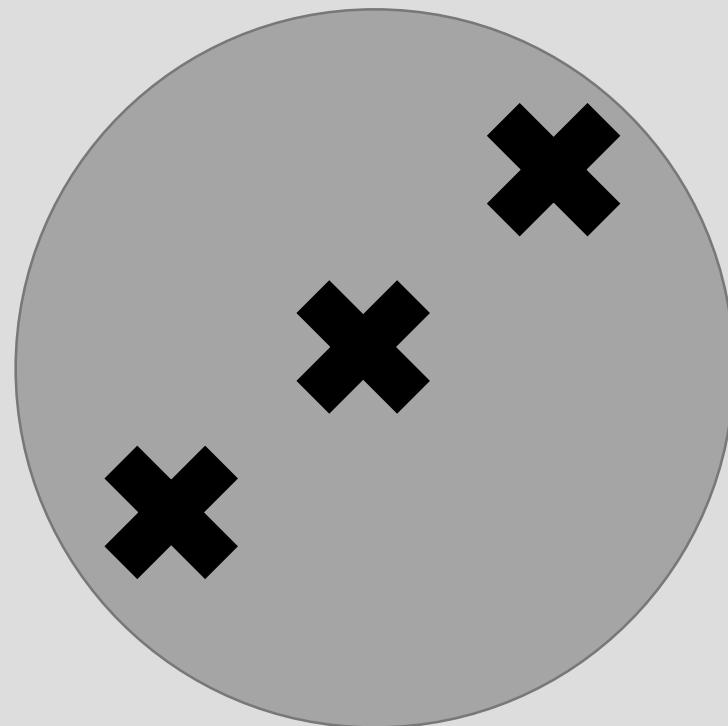
Quasi-Random Numbers



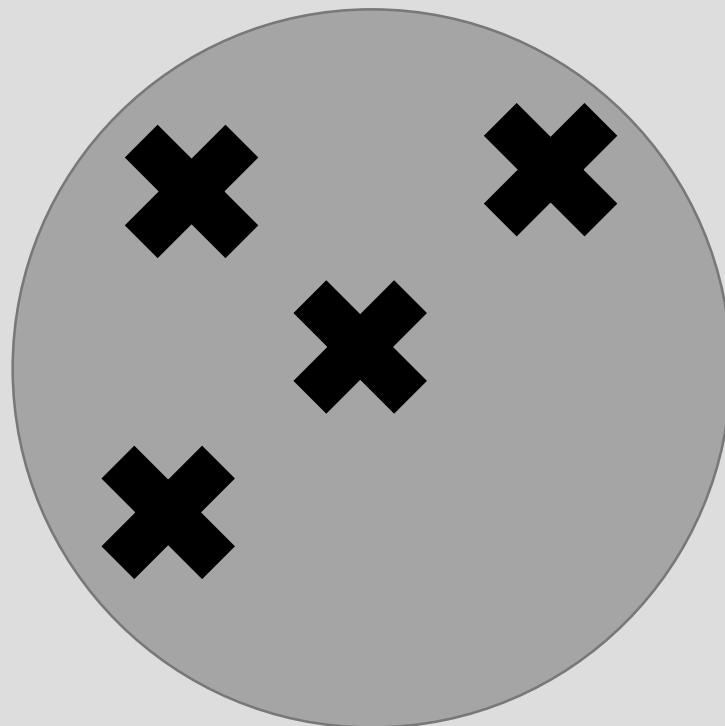
Quasi-Random Numbers



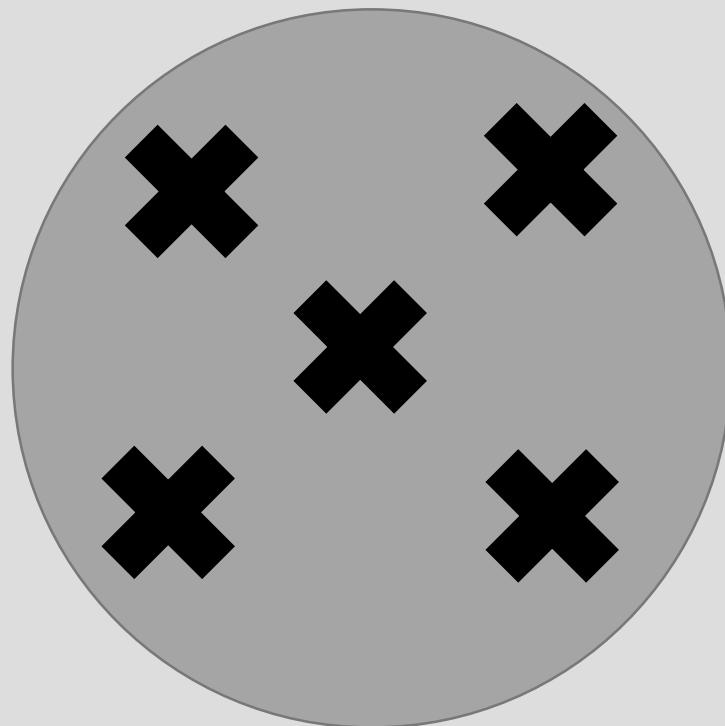
Quasi-Random Numbers



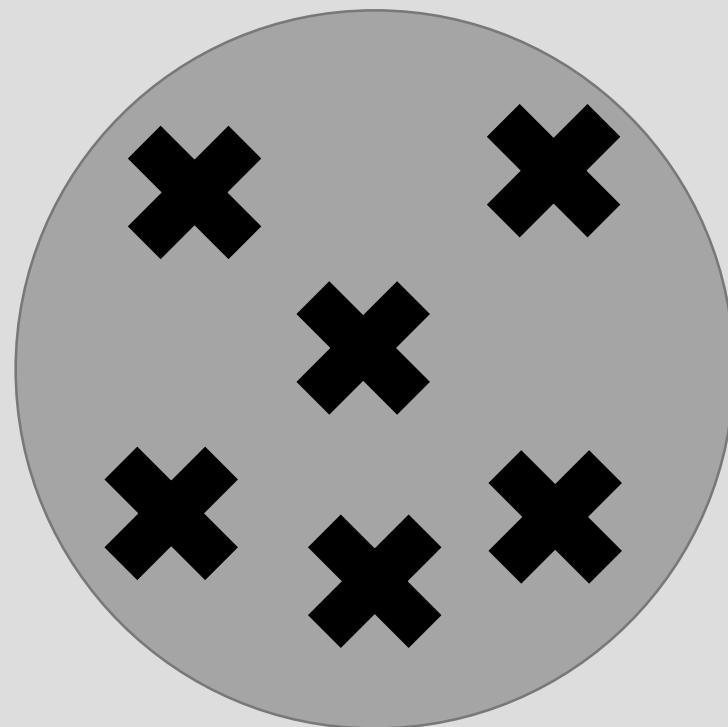
Quasi-Random Numbers



Quasi-Random Numbers



Quasi-Random Numbers



Halton

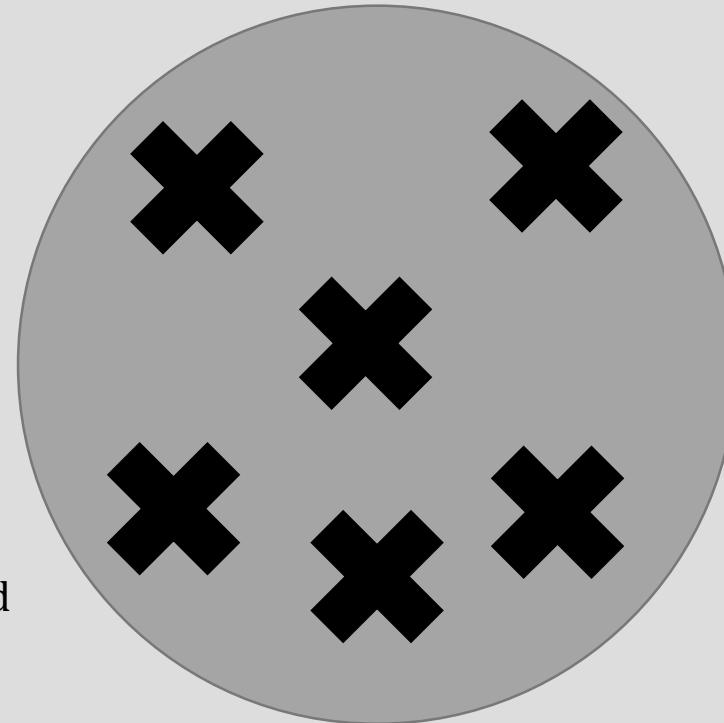
Niederreiter-Xing

Sobol

Quasi-Random Numbers Low-discrepancy sequence

$$\epsilon \propto c(d) \frac{(\ln N)^d}{N}$$

d is dimension, i.e.,
d numbers of random
numbers were estimated
per iteration.

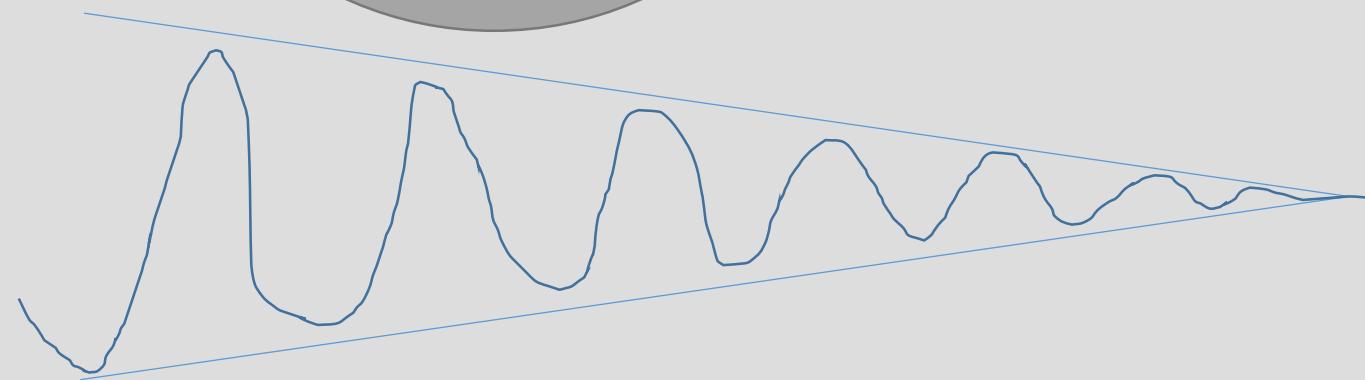


Pseudo-Random Numbers

Linear congruential
Subtract with carry
Mersenne twister

$$\epsilon \propto \frac{1}{\sqrt{N}}$$

N is number of iterations.



Halton

Niederreiter-Xing

Sobol

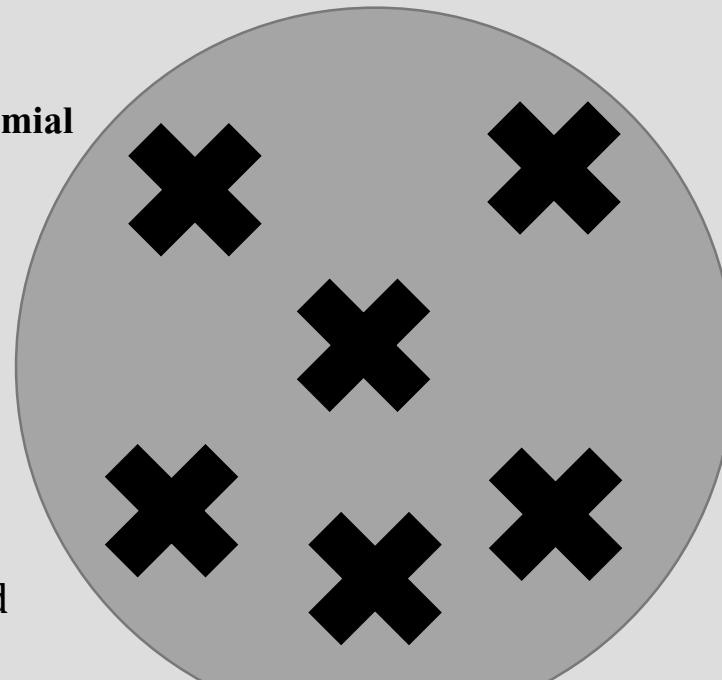
–it uses primitive polynomial
based bitwise arithmetic
without carry - XOR

$$\epsilon \propto c(d) \frac{(\ln N)^d}{N}$$

d is dimension, i.e.,
d numbers of random
numbers were estimated
per iteration.

$y = f(x, \text{random}()) + g(x, \text{random}()); \rightarrow \text{equation 1}$

$P = f(Q, \text{random}()) + g(R, \text{random}()) + h(T, \text{random}()); \rightarrow \text{equation 2}$



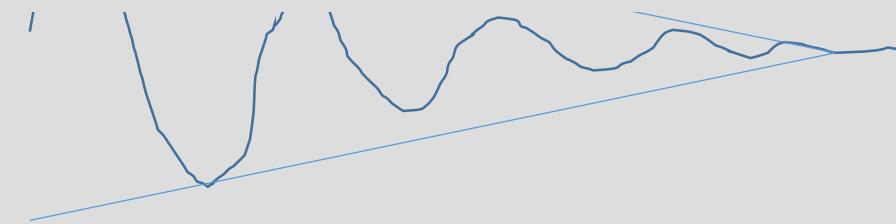
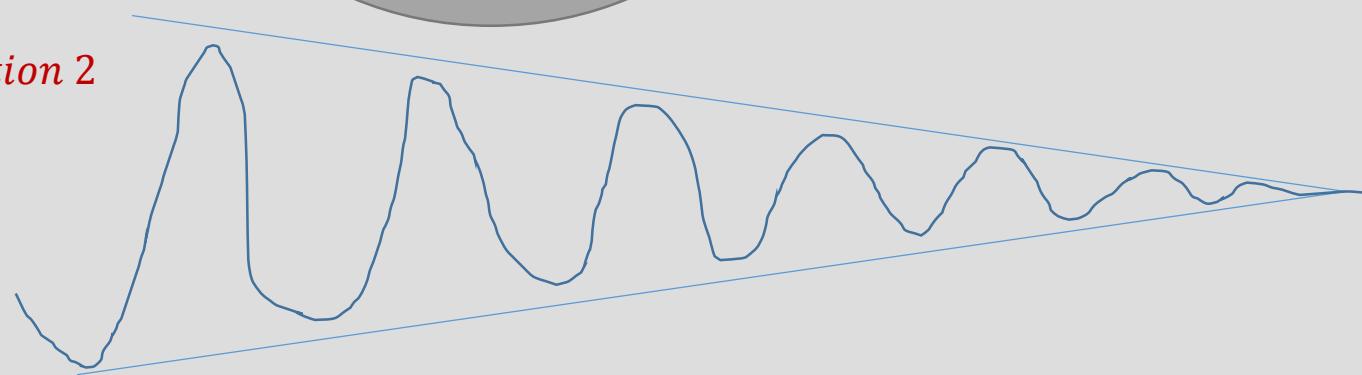
Quasi-Random Numbers Low-discrepancy sequence

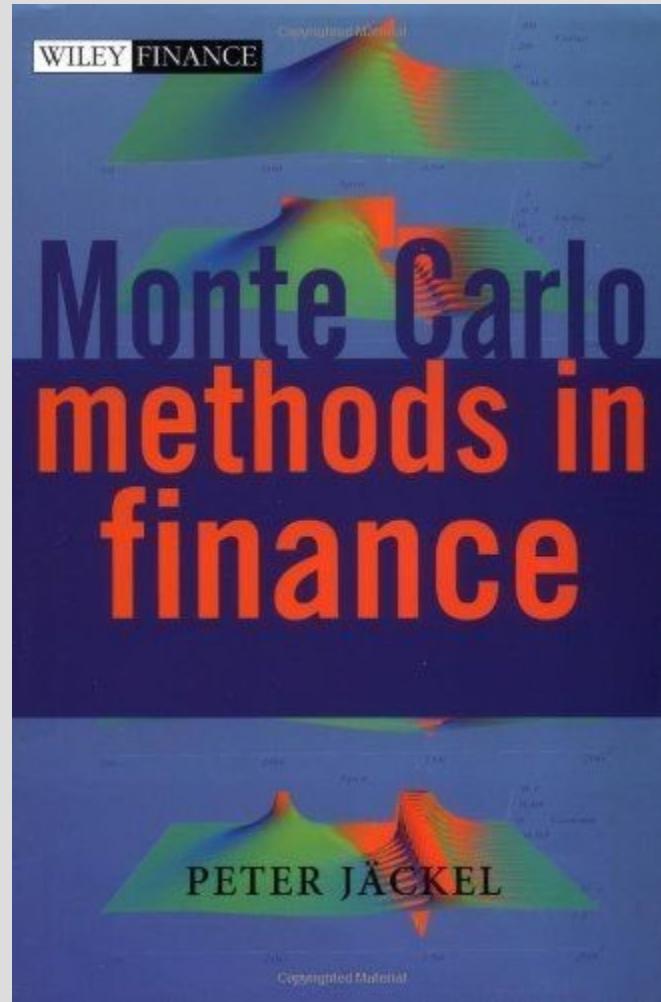
Pseudo-Random Numbers

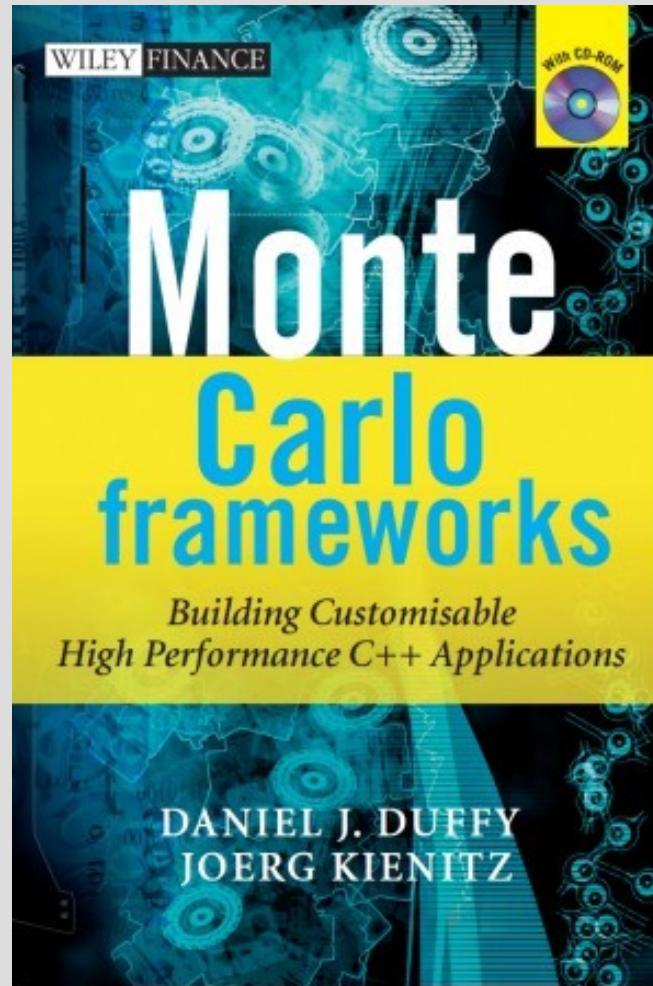
Linear congruential
Subtract with carry
Mersenne twister

$$\epsilon \propto \frac{1}{\sqrt{N}}$$

N is number of iterations.

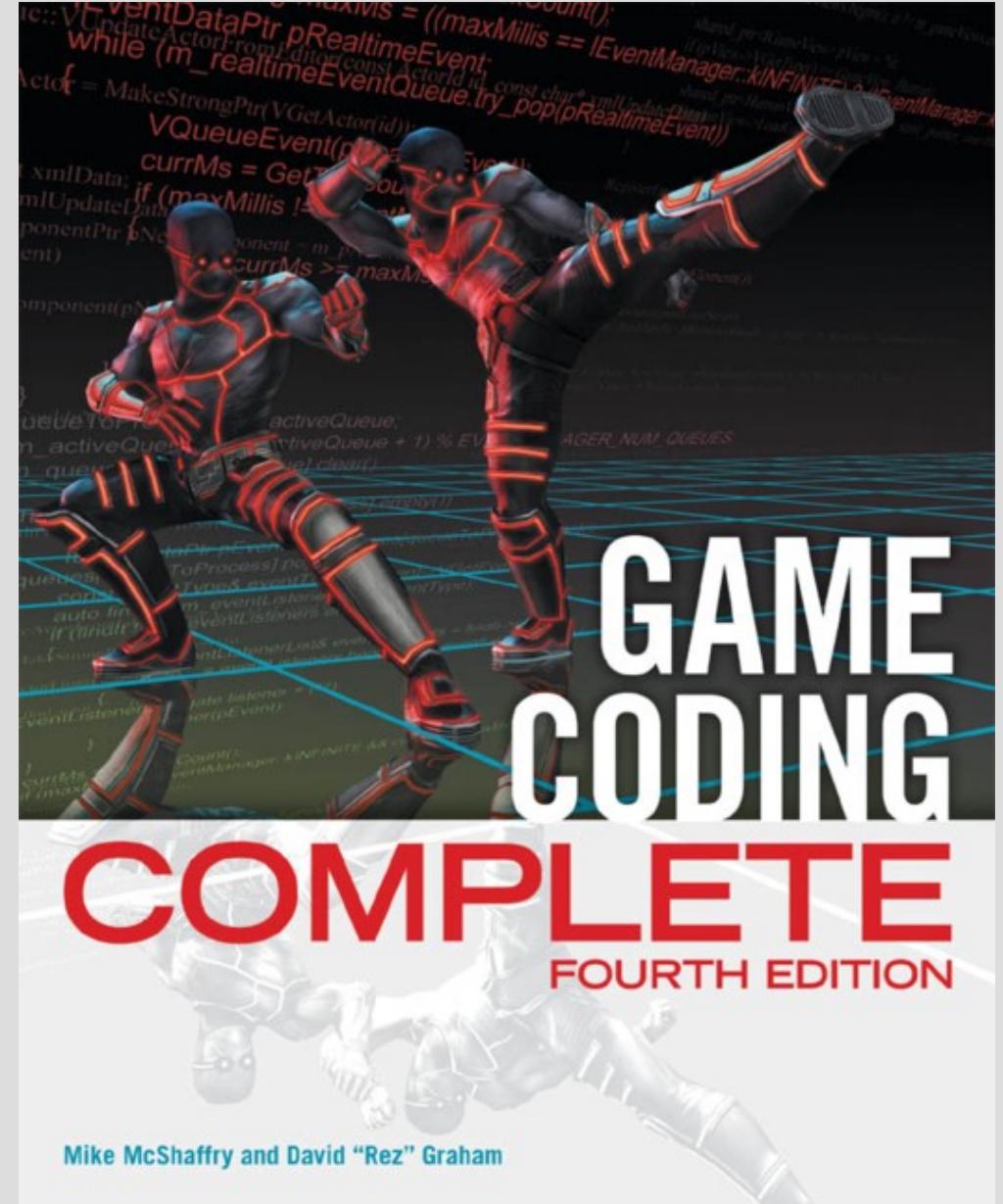






“Most programmers will soon discover that the rand() function is completely inadequate because it can only generate a single stream of random numbers.

Most games need multiple discrete streams of random numbers”.



Potential candidates for C++17:

- Dimensionality to random number generations;
- Low-discrepancy sequence (Sobol numbers);

```
int pick_a_number( int from, int thru )
{
    static std::uniform_int_distribution<> d{};
    using parm_t = decltype(d)::param_type;
    return d( global_urng(), parm_t{from, thru} );
}

double pick_a_number( double from, double upto )
{
    static std::uniform_real_distribution<> d{};
    using parm_t = decltype(d)::param_type;
    return d( global_urng(), parm_t{from, upto} );
}
```

Three <random>-related Proposals, v2

Document #: WG21 N3742

Date: 2013-08-30

Revises: [N3547](#)

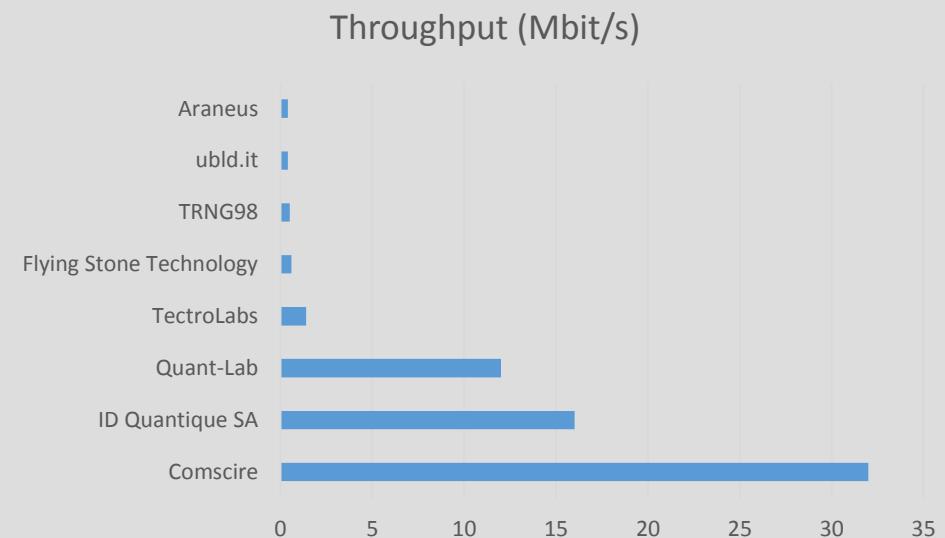
Project: JTC1.22.32 Programming Language C++

Reply to: Walter E. Brown <webrown.cpp@gmail.com>

Hardware Random Number Generator

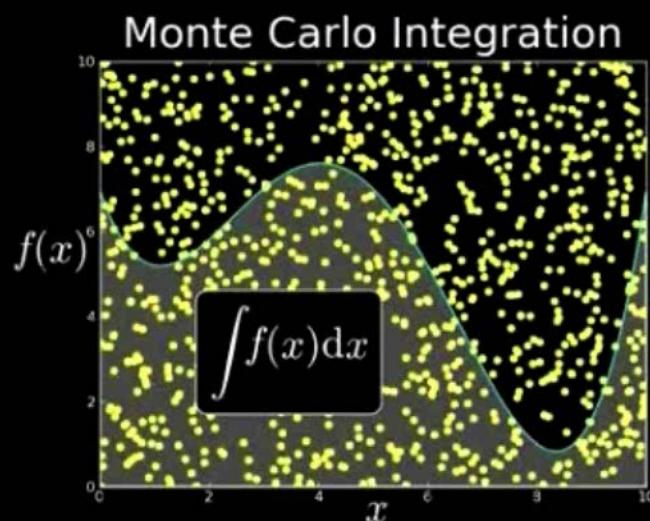


Manufacturer	Throughput (Mbit/s)	Price	Intro Date
Comscire	32	\$1,495	2013
ID Quantique SA	16	2,230.00 €	2006
Quant-Lab	12	2,700.00 €	2005
TectroLabs	1.4	\$329	2013
Flying Stone Tech	0.6	JPY 4,000	2013
TRNG98	0.5	620.00 €	2009
ubld.it	0.4	\$49.95	2014
Araneus	0.4	159 €	2003



cuRAND: Random Number Generation

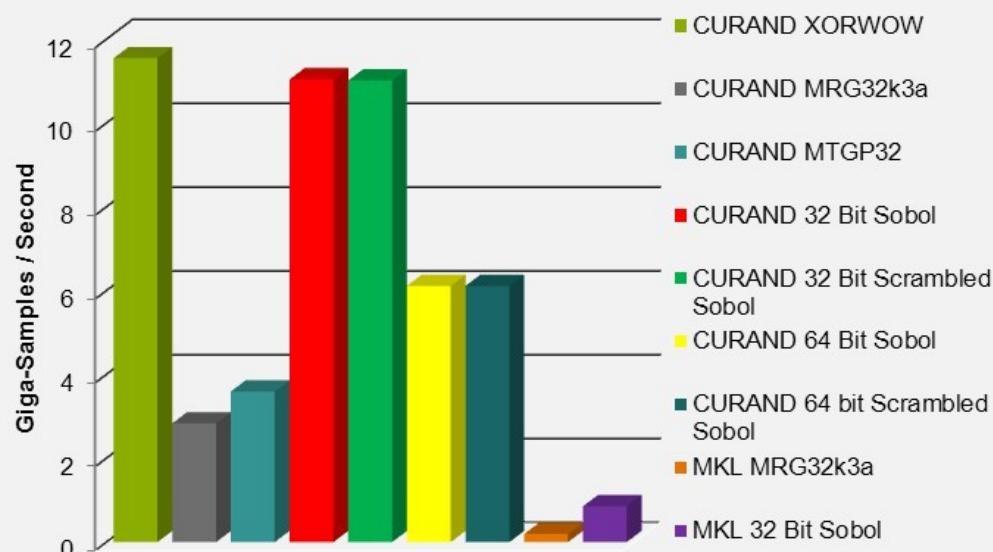
- Generating high quality random numbers in parallel is hard
 - Don't do it yourself, use a library!
- Pseudo- and Quasi-RNGs
- Supports several output distributions
- Statistical test results in documentation
- New in CUDA 5.0: Poisson distribution



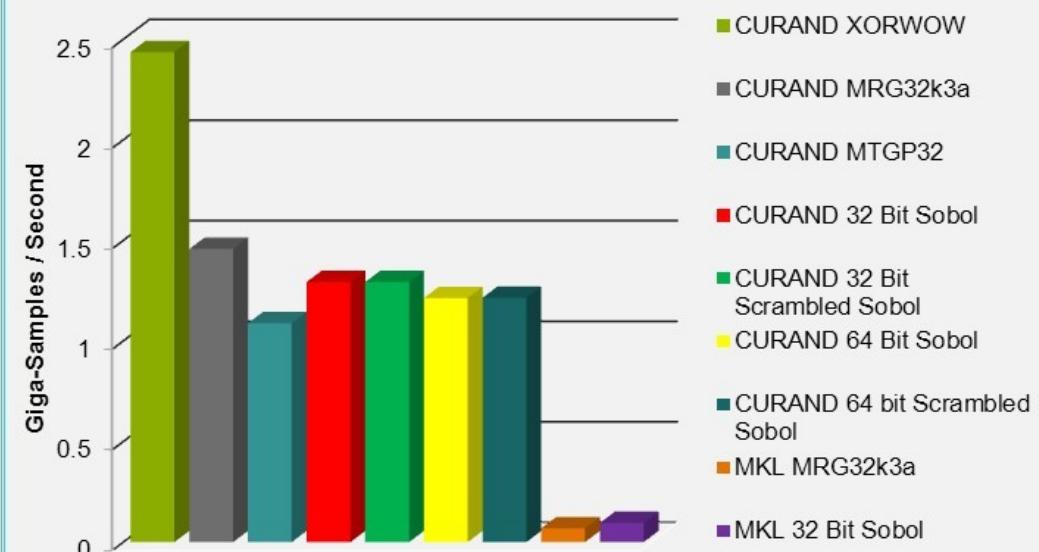
cuRAND Engines

Normal Distributions(cuRAND Engines)

Double Precision Uniform Distribution



Double Precision Normal Distribution



- cuRAND 4.1 on Tesla M2090, ECC on
- MKL 10.2.3, TYAN FT72-B7015 Xeon x5680 Six-Core @ 3.33 GHz

• Performance may vary based on OS ver. and motherboard config.

“Skip ahead or saving state to avoid overlap is the caller’s responsibility”

seed[n] = seed[n-1] + 1000; (.discard(i * iterations);)

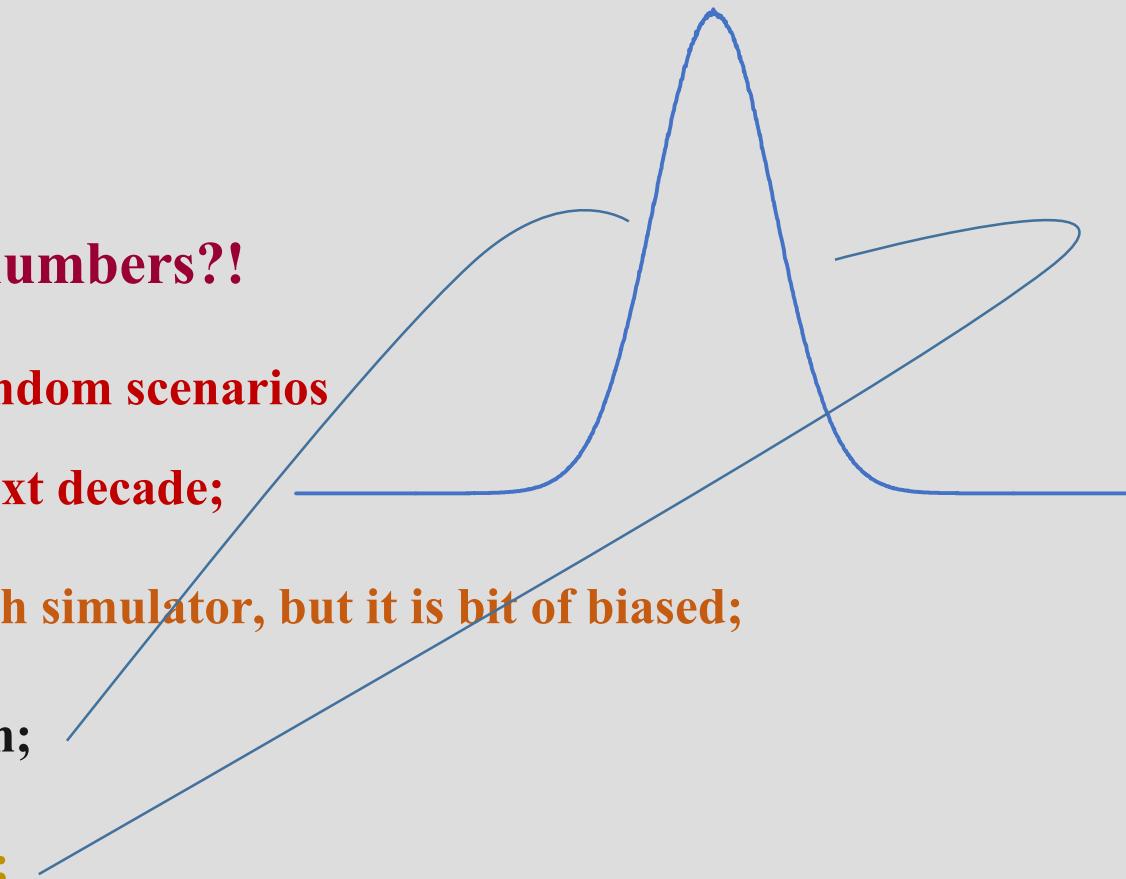
The screenshot shows a web browser displaying the NVIDIA Developer Zone CUDA Toolkit Documentation. The URL in the address bar is docs.nvidia.com/cuda/curand/index.html. The page title is "CURAND". The left sidebar contains a navigation menu with the following structure:

- CUDA Toolkit
- ▼ CURAND
 - Introduction
 - 1. Compatibility and Versioning
 - 2. Host API Overview
 - 2.1. Generator Types
 - 2.2.1. Seed
 - 2.2.2. Offset
 - 2.2.3. Order
 - 2.3. Return Values
 - 2.4. Generation Functions
 - 2.5. Host API Example
 - 2.6. Performance Notes
 - 3. Device API Overview
 - 3.1. Pseudorandom Sequences
 - 3.1.1. Bit Generation with XORWOW and MRG32k3a generators
 - 3.1.2. Bit Generation with the MTGP32 generator
 - 3.1.3. Distributions
 - 3.2. Quasirandom Sequences
 - 3.3. Skip-Ahead
 - 3.4. Device API for discrete distributions
 - 3.5. Performance Notes
 - 3.6. Device API Examples
 - 3.7. Thrust and CURAND Example
 - 3.8. Poisson API Example
 - 4. Testing
 - 5. Modules
 - 5.1. Host API
 - 5.2. Device API
 - A. Bibliography
 - B. Acknowledgements
 - Notices

The "2.2.1. Seed" link under "2.2. Generator Types" is highlighted in red, indicating it is the current section being viewed.

Do you realise, our brain is a source of random numbers?!

- It accesses the past and present data, generates random scenarios and simulates future be it next minute or next decade;
- It is a sophisticated random generator, very smooth simulator, but it is bit of biased;
- Pessimistic brain skewed towards negative random;
- Optimistic brain skewed towards positive random;



Thank You!