

# Generating Random and Pseudorandom Numbers

Some slides from CS 15-853: Algorithms in the Real World,  
Carnegie Mellon University

# Random Numbers in the Real World



<https://fitforrandomness.files.wordpress.com/2010/11/dilbert-does-randomness.jpg>

```
int getRandomNumber()  
{  
    return 4; // chosen by fair dice roll.  
             // guaranteed to be random.  
}
```

<https://xkcd.com/221/>

# Random number sequence definitions

Each element is chosen independently from a probability distribution [Knuth].

Randomness of a sequence is the Kolmogorov complexity of the sequence (size of smallest Turing machine that generates the sequence) - infinite sequence should require infinite size Turing machine.

# Environmental Sources of Randomness

Radioactive decay <http://www.fourmilab.ch/hotbits/>

Radio frequency noise <http://www.random.org>

Noise generated by a resistor or diode.

- Canada <http://www.tundra.com/> (find the data encryption section, then look under RBG1210. My device is an NM810 which is 228? RBG1210s on a PC card)
- Colorado <http://www.comscire.com/>
- Holland <http://valley.interact.nl/av/com/orion/home.html>
- Sweden <http://www.protego.se>

Inter-keyboard timings (watch out for buffering)

Inter-interrupt timings (for some interrupts)



# Combining Sources of Randomness

Suppose  $r_1, r_2, \dots, r_k$  are random numbers from different sources. E.g.,

$r_1$  = from JPEG file

$r_2$  = sample of hip-hop music on radio

$r_3$  = clock on computer

$$b = r_1 \oplus r_2 \oplus \dots \oplus r_k$$

If any one of  $r_1, r_2, \dots, r_k$  is truly random, then so is  $b$ .

# Skew Correction

Von Neumann's algorithm - converts biased random bits to unbiased random bits:

Collect two random bits.

Discard if they are identical.

Otherwise, use first bit.

Efficiency?

# Chi Square Test

Experiment with  $k$  outcomes, performed  $n$  times.

$p_1, \dots, p_k$  denote probability of each outcome

$Y_1, \dots, Y_k$  denote number of times each outcome occurred

$$\chi^2 = \sum_{1 \leq s \leq k} \frac{(Y_s - np_s)^2}{np_s}$$

Large  $\chi^2$  indicates deviance from random chance

# Analysis of random.org numbers

## John Walker's Ent program

Entropy = 7.999805 bits per character.

Optimum compression would reduce the size of this 1048576 character file by 0 percent.

Chi square distribution for 1048576 samples is 283.61, and randomly would exceed this value 25.00 percent of the times.

Arithmetic mean value of data bytes is 127.46 (127.5 = random).

Monte Carlo value for PI is 3.138961792 (error 0.08 percent).

Serial correlation coefficient is 0.000417 (totally uncorrelated = 0.0)

# Analysis of JPEG file

Entropy = 7.980627 bits per character.

Optimum compression would reduce the size of this 51768 character file by 0 percent.

Chi square distribution for 51768 samples is 1542.26, and randomly would exceed this value 0.01 percent of the times.

Arithmetic mean value of data bytes is 125.93 (127.5 = random).

Monte Carlo value for Pi is 3.169834647 (error 0.90 percent).

Serial correlation coefficient is 0.004249 (totally uncorrelated = 0.0).

# Pseudorandom Number Generators

- A pseudorandom number generator (PRNG) is an algorithm for generating a sequence of numbers whose properties approximate the properties of sequences of random numbers.
- The PRNG-generated sequence is not truly random, because it is completely determined by an initial value, called the PRNG's seed (which may include truly random values).
- Although sequences that are closer to truly random can be generated using hardware random number generators, pseudorandom number generators are important in practice for their speed and reproducibility.

# Pseudorandom Number Generators

- PRNGs are central in applications such as simulations (e.g. for the Monte Carlo method), electronic games (e.g. for procedural generation), and cryptography.
- Cryptographic applications require the output not to be predictable from earlier outputs.

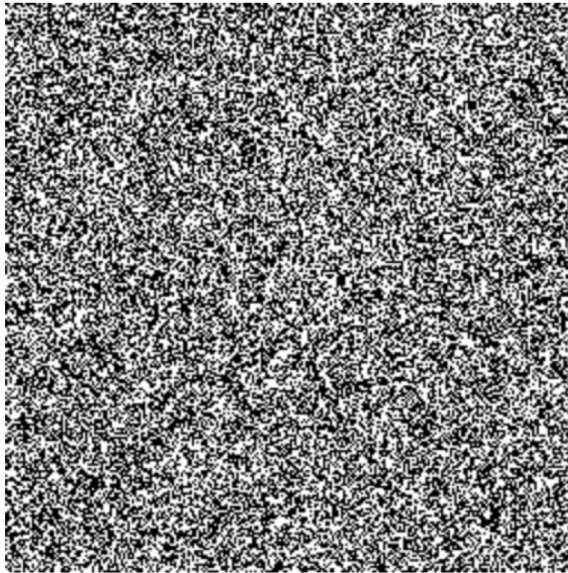
*"Anyone who considers arithmetical methods of producing random digits is, of course, in a state of sin."*

*- John Von Neumann, 1951*

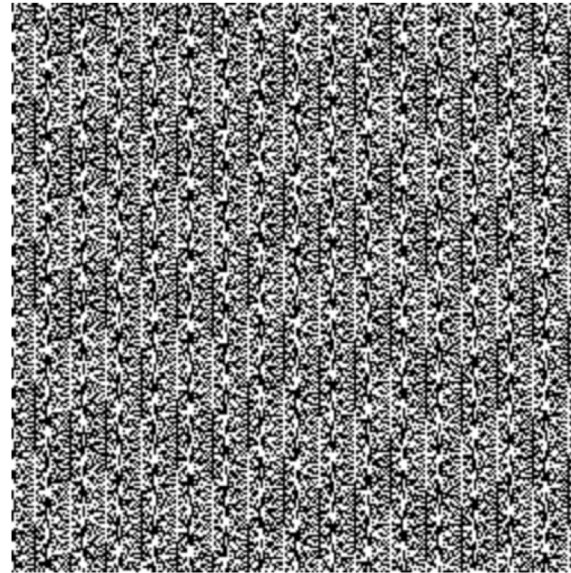


# Simple Visual Test

- Create a visualization of the consecutive tuples of numbers it produces.
- Humans are really good at spotting patterns.



**RANDOM.ORG**



**PHP rand() on Microsoft Windows**



# Linear Congruential Generator (LCG)

$$x_0 = \text{given}, x_{n+1} = P_1 x_n + P_2 \pmod{N} \quad n = 0, 1, 2, \dots \quad (*)$$

$$x_0 = 79, N = 100, P_1 = 263, \text{ and } P_2 = 71$$

$$x_1 = 79 * 263 + 71 \pmod{100} = 20848 \pmod{100} = 48,$$

$$x_2 = 48 * 263 + 71 \pmod{100} = 12695 \pmod{100} = 95,$$

$$x_3 = 95 * 263 + 71 \pmod{100} = 25056 \pmod{100} = 56,$$

$$x_4 = 56 * 263 + 71 \pmod{100} = 14799 \pmod{100} = 99,$$

Sequence: 79, 48, 95, 56, 99, 8, 75, 96, 68, 36, 39, 28, 35, 76, 59, 88,  
15, 16, 79, 48, 95

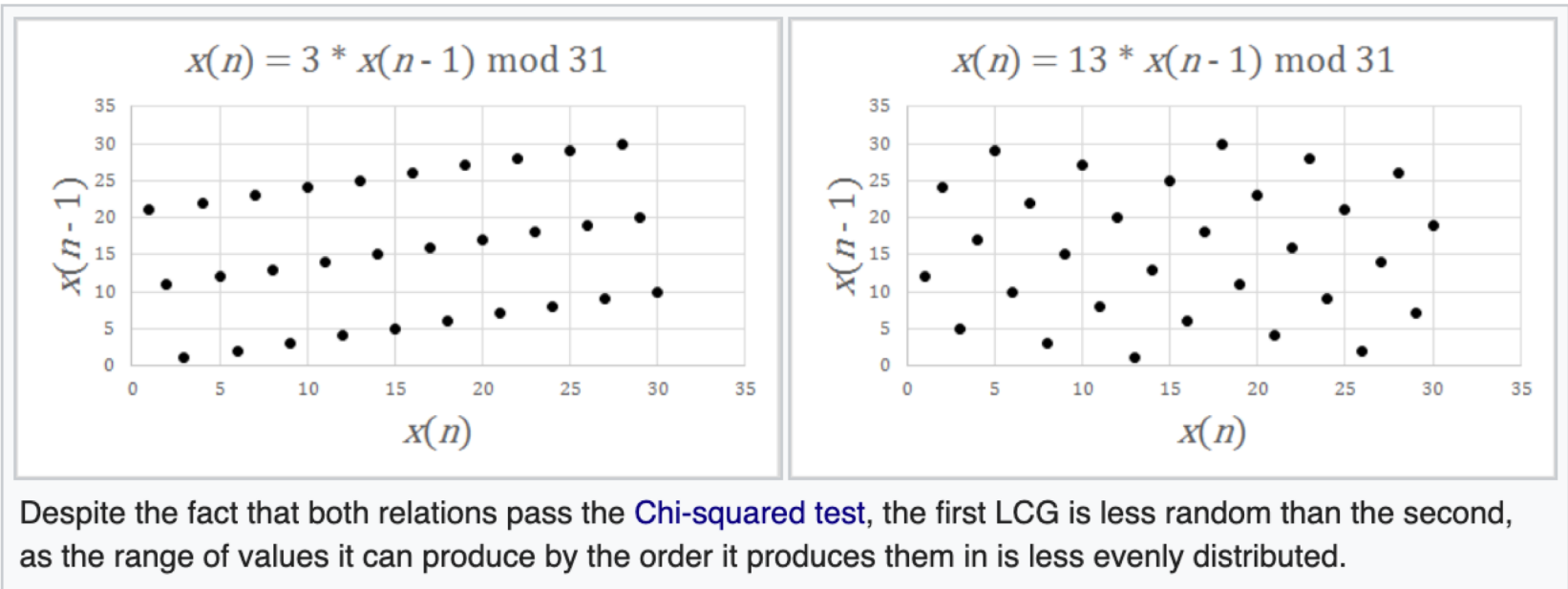
Park and Miller:

$$P_1 = 16807, P_2 = 0, N = 2^{31} - 1 = 2147483647, x_0 = 1.$$

ANSI C rand():

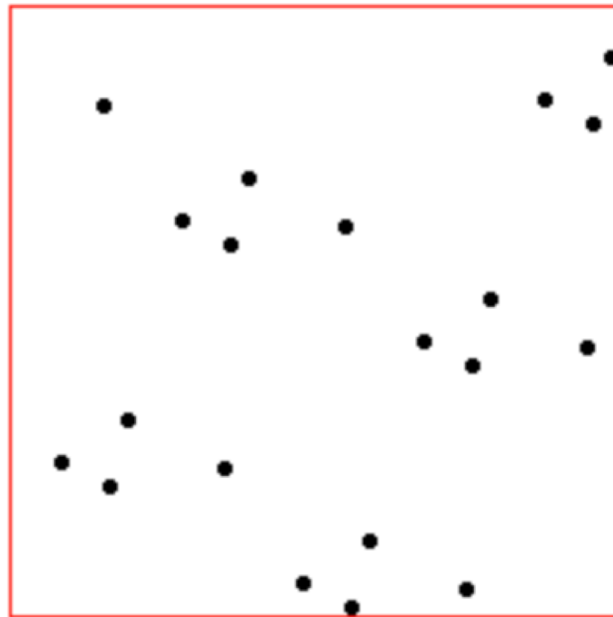
$$P_1 = 1103515245, P_2 = 12345, N = 2^{31}, x_0 = 12345$$

# Example Comparison



# Plot ( $x_i, x_{i+1}$ )

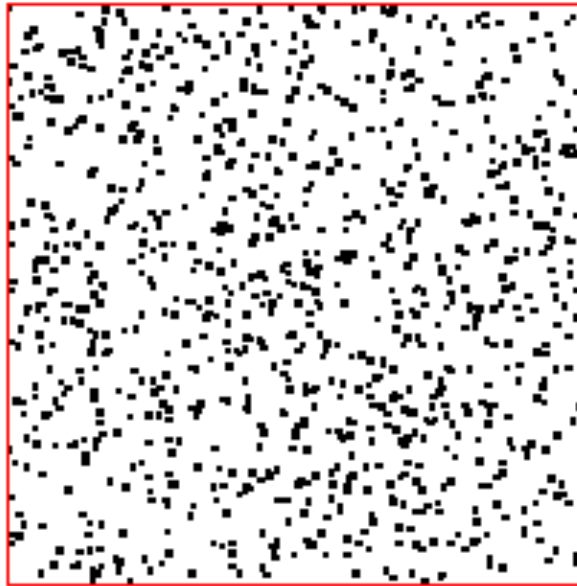
P1 = 263, P2 = 71, N = 100



100 dots drawn, seed = 79

# Plot $(x_i, x_{i+1})$

P1 = 16807, P2 = 0, N = 2147483647

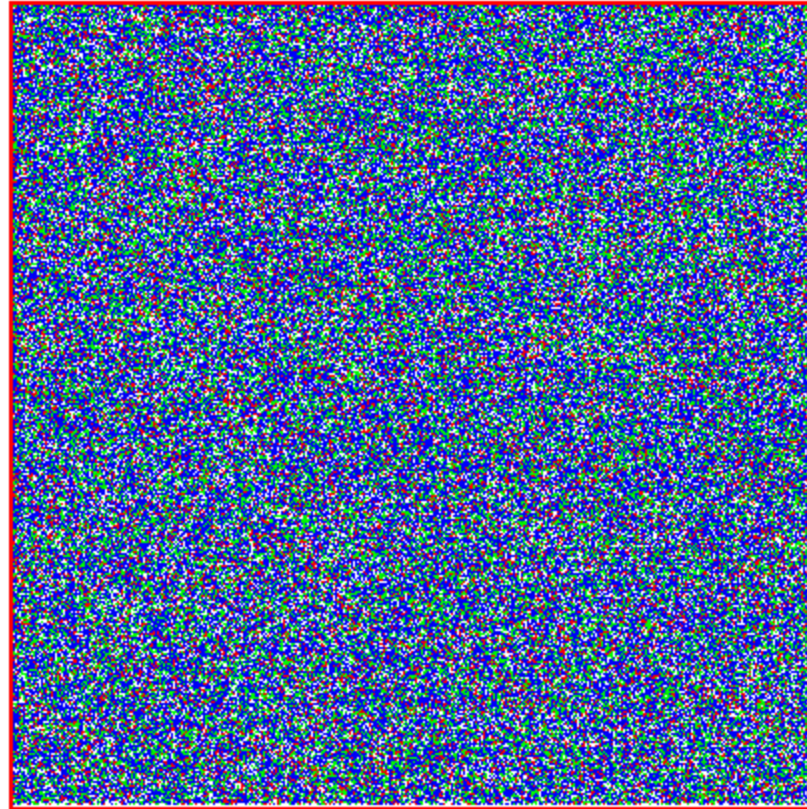


1000 dots drawn, seed = 1

Park and Miller

$(x_i, x_{i+1}), (x_i, x_{i+2}), (x_i, x_{i+2})$

P1 = 16807, P2 = 0, N = 2147483647

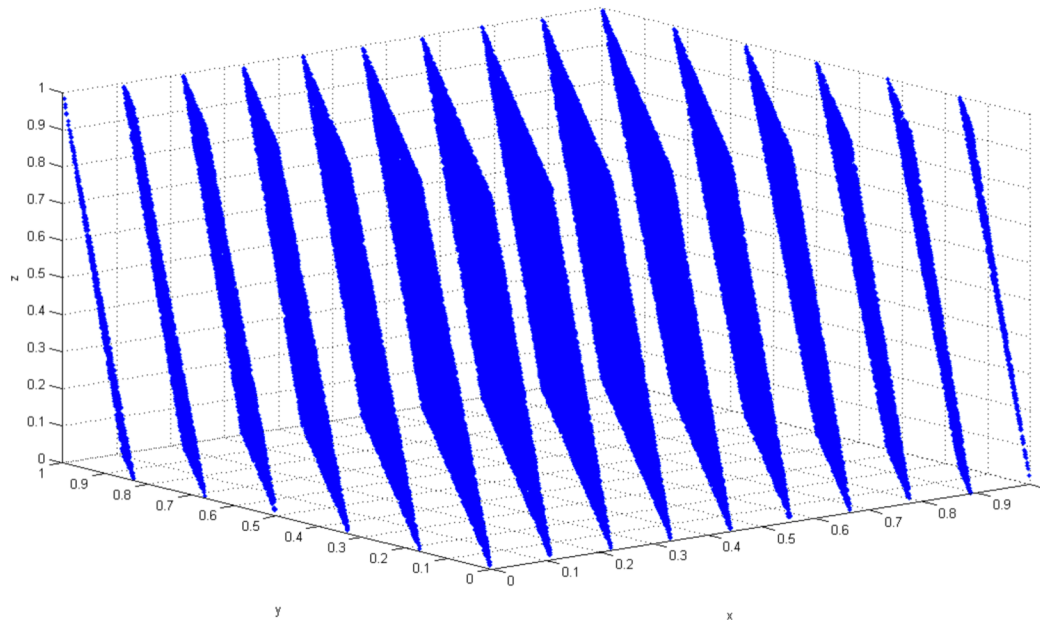


100000 dots drawn, seed = 1

<http://www.math.utah.edu/~alfeld/Random/Random.html>

# Visual Test in 3D

- Three-dimensional plot of 100,000 values generated with IBM RANDU routine. Each point represents 3 consecutive pseudorandom values.
- It is clearly seen that the points fall in 15 two-dimensional planes.

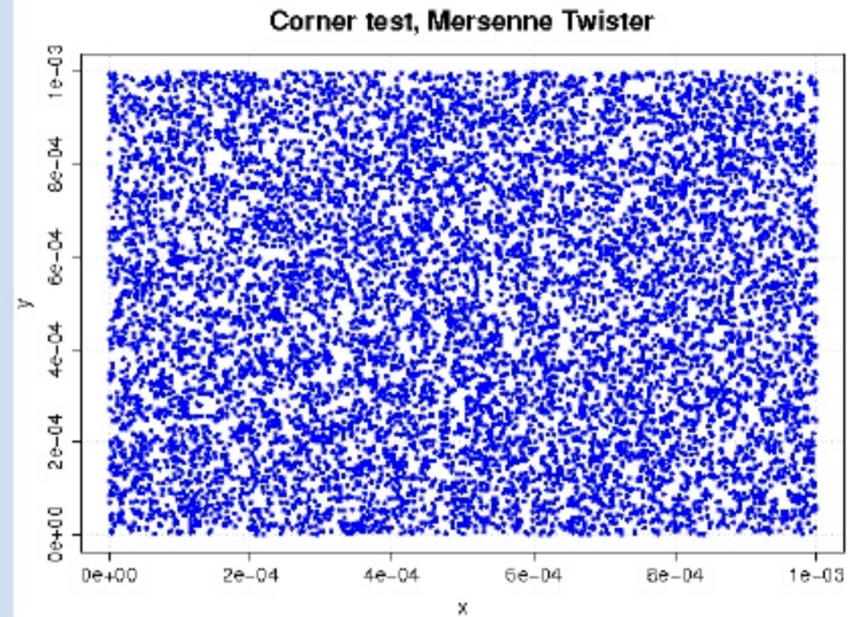
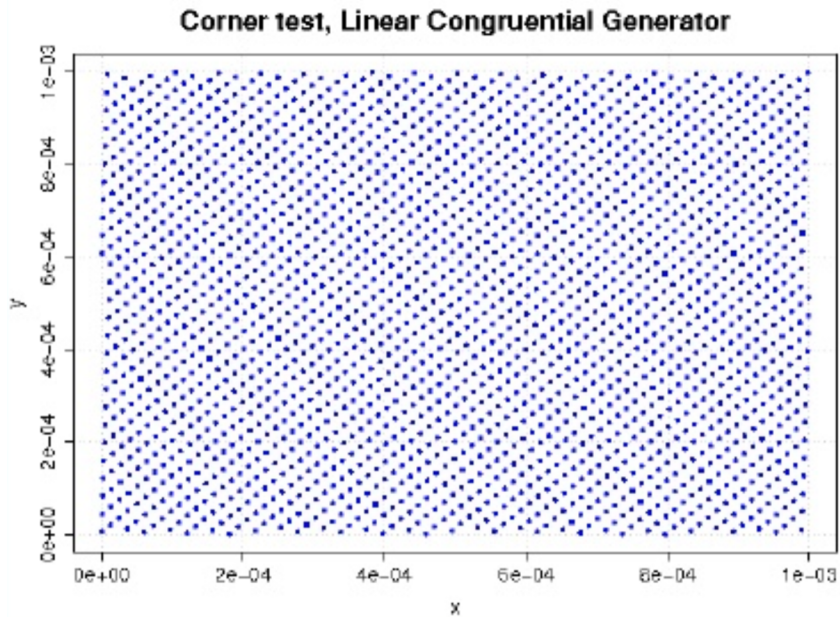


# Matsumoto's Marsenne Twister

Considered one of the best linear congruential generators.

<http://www.math.sci.hiroshima-u.ac.jp/~m-mat/MT/emt.html>

# Example Visual Test



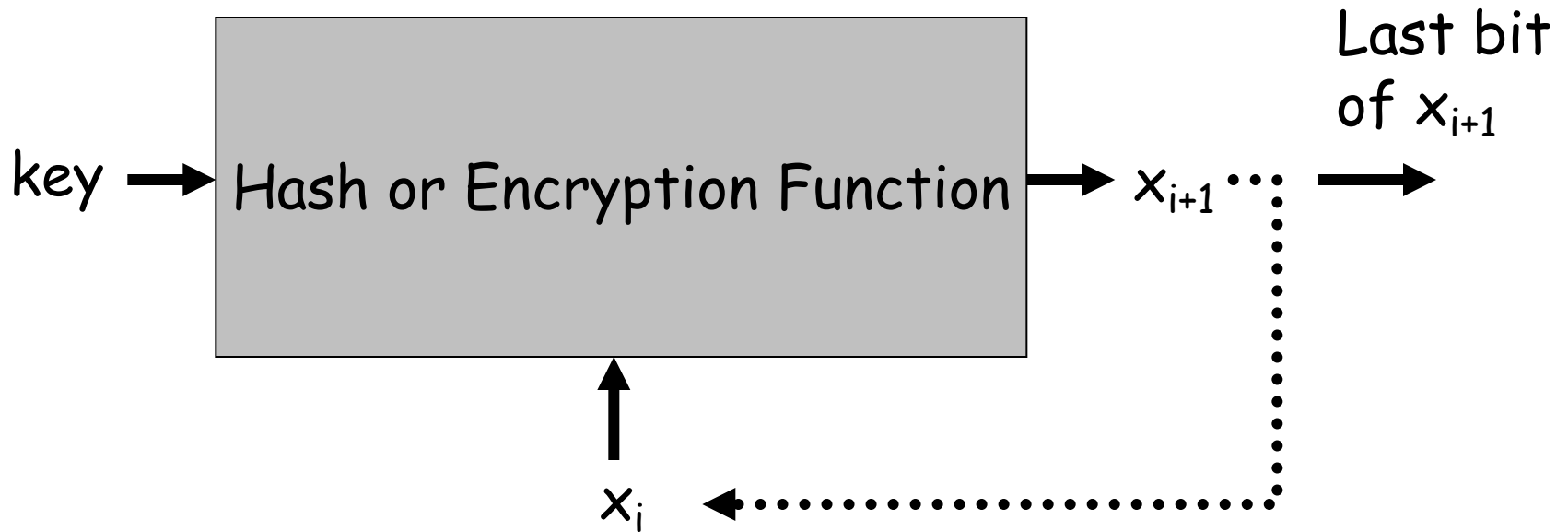


# Cryptographically Strong Pseudorandom Number Generator

Next-bit test: Given a sequence of bits  $x_1, x_2, \dots, x_k$ , there is no polynomial time algorithm to generate  $x_{k+1}$ .

Yao [1982]: A sequence that passes the next-bit test passes all other polynomial-time statistical tests for randomness.

# Hash/Encryption Chains



(need a random seed  $x_0$  or key value)

# Some Encryption and Hash Functions

- SHA-1 Hash function <https://en.wikipedia.org/wiki/SHA-1>
- MD5 Hash function <https://en.wikipedia.org/wiki/MD5>
- Data Encryption Standard (DES)  
[https://en.wikipedia.org/wiki/Data\\_Encryption\\_Standard](https://en.wikipedia.org/wiki/Data_Encryption_Standard)
- Advanced Encryption Standard (AES)  
[https://en.wikipedia.org/wiki/Advanced\\_Encryption\\_Standard](https://en.wikipedia.org/wiki/Advanced_Encryption_Standard)

# BBS "secure" random bits

## BBS (Blum, Blum and Shub, 1984)

- Based on difficulty of factoring, or finding square roots modulo  $n = pq$ .

### Fixed

- $p$  and  $q$  are primes such that  $p = q = 3 \pmod{4}$
- $n = pq$  (is called a Blum integer)

### For a particular bit seq.

- **Seed:** random  $x$  relatively prime to  $n$ .
- **Initial state:**  $x_0 = x^2$
- **$i^{\text{th}}$  state:**  $x_i = (x_{i-1})^2$
- **$i^{\text{th}}$  bit:** lsb of  $x_i$

Note that:  $x_0 = x_i^{-2^i \bmod \phi(n)} \pmod{n}$

Therefore knowing  $p$  and  $q$  allows us to find  $x_0$  from  $x_i$