

ORIE 4580/5580: Simulation Modeling and Analysis

ORIE 5581: Monte Carlo Simulation

unit 4: generating random numbers

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generating uniformly distributed random variables

random number: a sample from $U[0, 1]$

the 'fundamental theorem' of simulation

can 'transform' a stream of $U[0, 1]$ to any other random variable

- arbitrary probability distribution
- arbitrary correlations
- complex processes

where can we find random numbers?

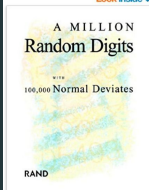
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A Million Random Digits with 100,000 Normal Deviates 0th Edition

by The RAND Corporation (Author)

★★★★☆ 682 customer reviews

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★★★★☆ **Weirdest sudoku book ever**

By [John Peter O'Connor](#) on October 6, 2012

Format: Paperback

This has got to be the most useless set of sudoku puzzles ever.

In my copy of the book, all of the puzzles were already filled in which I find really annoying and what is worse, most of them have been filled in wrongly.

I have been through the whole book really carefully and only found seven puzzles that had been filled out correctly! Yes, just seven.

Well, making the best of a bad job, I am now going through the book trying to correct all of the faulty puzzles and I will then submit my corrections.

Perhaps a second edition will be more useful.

I did find last week's winning lottery numbers on page 18 though.

• [Comment](#) 135 people found this helpful. Was this review helpful to you? [Report abuse](#)

★★★★☆ **Too unpredictable**

By [pontifex](#) on January 24, 2011

Format: Paperback

The book is too hard to follow, the author randomly shifts from one number to another without any prior warning.

• [1 comment](#) | 408 people found this helpful. Was this review helpful to you? [Report abuse](#)

★★★★☆ **Not really random**

By [TDB](#) on September 26, 2012

Format: Paperback

physical Methods

manual methods: coin toss, dice throw, drawing from an urn

objects that “appear” random: computer clock

physical devices: circuit noise, gamma-ray detectors

advantage

“true” random numbers (critical for cryptographic applications)

- for example, check out [Radiolab podcast](#) on launching a cryptocurrency

drawbacks

- **slow** (if generated as needed)
expensive (if precomputed and stored in memory)
- **bias** may exist in the device
for example, see [Persi Diaconis on coin-tossing](#)
- **hard to replicate** the random input sequence

'pseudo-random' generators

- mid-square method (von Neumann, 1949)

$$8234 \times 8234 = 67(7987)56$$

$$7987 \times 7987 = 63(7921)69$$

$$7921 \times 7921 = 62(7422)41$$

$$7422 \times 7422 = 55(0860)84 \dots$$

- **objection:** random numbers are not random at all!
 - this criticism applies to all **pseudo-random number generators**
 - need tests to determine if algorithm produces “valid” outputs

linear congruential generators

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

- (fixed) parameters: modulus m , multiplier a , increment c
- seed: X_0 (the first input) is typically supplied by the user
- each X_n is an integer in the set $\{0, 1, \dots, m-1\}$.
- to get pseudorandom number $U_n \in (0, 1)$, set:

LCG: example

$m = 32$, $a = 11$, $c = 0$, different seeds.

n	X_n	X_n	X_n
0	1	2	4
1	11	22	12
2	25	18	4
3	19	6	12
4	17	2	4
5	27	22	12
6	9	18	4
7	3	6	12
8	1	2	4
9	11	22	12
\vdots	\vdots	\vdots	\vdots

LCG: properties

- an LCG produces **periodic output**. $\text{period} \leq m$
 1. if $\text{period} = m$ with a seed X_0 , then $\text{period} = m$ for any seed
 2. if $\text{period} < m$, then it may depend on the seed
- **full period is desirable**:
 1. one should never use the whole period of a LCG, otherwise dependencies between the random numbers will occur.
 2. not have full period \implies gaps in the output sequence.
- **full period \implies granularity $= 1/m$** .
not a problem when m is large.

LCG: theory

the period of LCGs is well understood

(the following results are for demonstration; you do not need to know them)

Theorem: LCGs with full period

an LCG(m, a, c) has full period if **all of the following** are true

1. m and c are relatively prime.
2. if q is a prime number that divides m , then q divides $a - 1$.
3. if 4 divides m , then 4 divides $a - 1$.

Corollary:

an LCG with $m = 2^b$ has full period if **c is odd and 4 divides $a - 1$** .

example – $m = 8, \quad a = 5, \quad c = 3$

multiplicative generators

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

- $X_n = 0 \implies X_n = X_{n+1} = X_{n+2} = \dots = 0$
- for an MG, full period $\implies \{1, \dots, m-1\}$.

Theorem: MGs with full period

an MG has full period if **all of the following** are true

1. m is prime
2. $a^{m-1} - 1$ is divisible by m .
3. there is no $j < m-1$ such that $a^j - 1$ is divisible by m .

Theorem: Sufficient conditions for good MGs

the largest possible period for a MG with $m = 2^b$ is $m/4$

this is achieved when X_0 is odd and a is of the form:

$$a = 3 + 8k \quad \text{or} \quad a = 5 + 8k,$$

for some positive integer k .

deficiencies of LCGs

- LCG's possess **theoretical deficiencies**
(any deterministic generator must have deficiencies.)
- if U_0, U_1, \dots are iid $U[0, 1]$, then the points

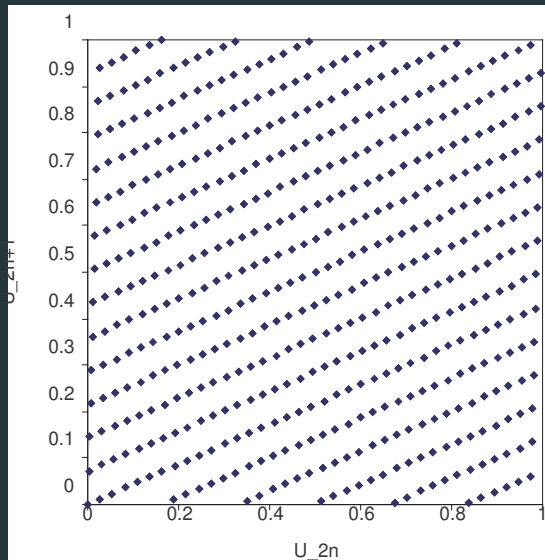
$$(U_0, U_1), (U_2, U_3), (U_4, U_5), \dots$$

should lie uniformly over the square $[0, 1] \times [0, 1]$.

- suppose U_0, U_1, \dots be generated by a LCG:
how do the points $(U_0, U_1), (U_2, U_3), (U_4, U_5), \dots$ behave?

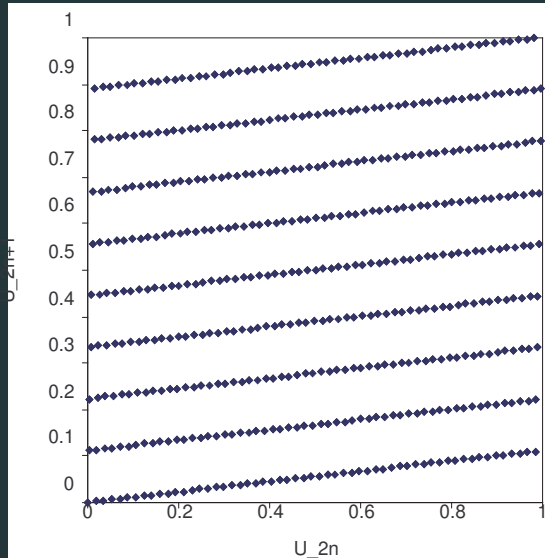
deficiencies of LCGs

$$m = 2^{10}, a = 37, c = 1$$



deficiencies of LCGs

$$m = 2^{10}, a = 57, c = 1$$



deficiencies of LCGs

- the points

$$(U_0, U_1), (U_2, U_3), (U_4, U_5), \dots$$

lie on a relatively small number of parallel lines!

- in general, the points

$$(U_0, U_1, U_2, \dots, U_{d-1}), (U_d, U_{d+1}, U_{d+2}, \dots, U_{2d-1}), \dots$$

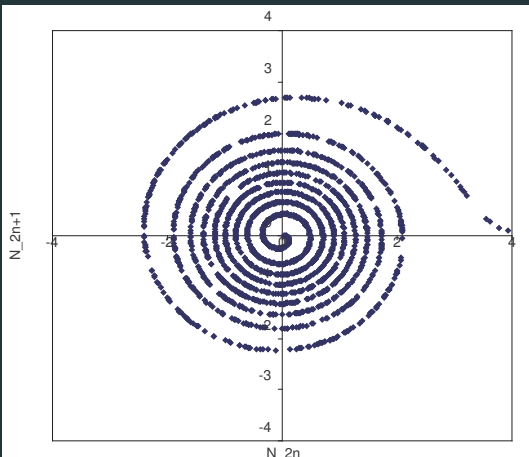
lie on parallel $(d - 1)$ -dimensional hyperplanes!

- problematic in simulations of geometric phenomena.
OK for discrete-event simulations.

deficiencies of LCGs for generating other rvs

let N_0, N_1, N_2, \dots be samples from $\mathcal{N}(0, 1)$ generated using the Box-Muller method using U_0, U_1, U_2, \dots from an LCG

– then the pairs $(N_0, N_1), (N_2, N_3), (N_4, N_5), \dots$ lie on a spiral in two-dimensional space. E.g., $a = 9, m = 2^{21}, c = 1$



combining generators

- $m = 2^{31} - 1$ is popular, but period is only about 2 billion.
- **not sufficient!** e.g. traffic simulators need lots of random numbers. (10s of 000s of vehicles, 1000s of random disturbances, lots of replications).
- shouldn't use anywhere near the full period - maybe $\leq 1\%$
- to generate longer period, take two MG's

$$X_{n+1} = (a_1 X_n) \bmod m_1 \quad , \quad Y_{n+1} = (a_2 Y_n) \bmod m_2$$

and set

$$Z_n = (X_n + Y_n) \bmod m_3.$$

- period can be on the order of $m_1 m_2$; e.g., set $a_1 = 40014$, $a_2 = 40692$, $m_1 = 2147483563$, $m_2 = 2147483399$ and $m_3 = m_1$.
- can combine more than two.

streams and substreams

- useful to divide PRNG output into streams and substreams
- stream = simulation replication
substream = source of randomness
- useful for debugging and for variance-reduction techniques

- hyperplane/spiral problems are well understood (and avoided)
- current generators have been carefully tested, pass lots of statistical tests (but must fail at least one test...)

the last word

modern PRNGs are

- **random enough** for your sim answer to be correct
- **deterministic enough** (by setting seed) for your sim to be repeatable