Random and pseudorandom numbers

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Agenda

1. Applications of randomness
2. What does it mean «random numers»?
   - Problem formalization
   - Definition 1. Frequency stability
   - Definition 2. Unpredictability
3. Pseudorandom numbers
   - Random and pseudorandom numbers
   - Random numbers sources
   - Pseudorandom numbers sources
   - The Linear Congruential Method
   - Randomness criterias

Monte Carlo

Figure: Monte Carlo city

Stochastic systems modeling and analysis

1. Queueing theory
2. Mathematical finance

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4. Physics

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Decision theory (matrix games)

		Defender		
		(2,0)	(1,1)	(0,2)
Attacker	(2,0)	0	2/3	2/3
	(1,1)	1/3	0	2/3
	(0,2)	1/3	1/3	0

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Strategy $= (p_1, \quad , p_2, \quad p_3)$

Aesthetics

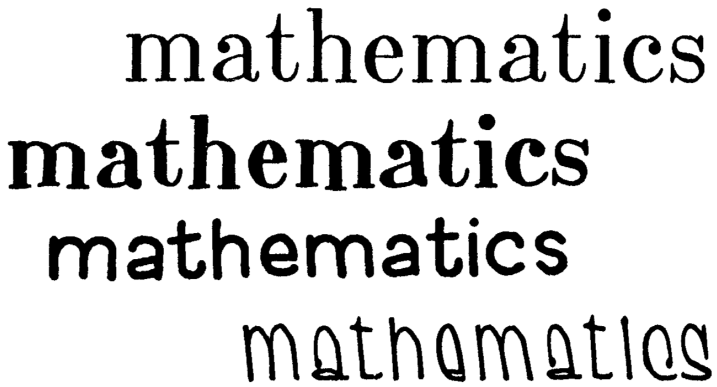

FIGURE 22.
A bit of randomness introduced into various styles of type.

"I think it can be said that the letters in this final example have a warmth and charm which makes it hard to believe that they were really generated by a computer following strict mathematical rules."

D.E. Knuth, Mathematical typography — Bull. Amer. Math. Soc. 1 (1979), 369

Numerical analysis

Denote $f(y) = \sum_{i=1}^{n} y_i^2$. Want to estimate

$$I = \int_0^1 dy_1 \dots \int_0^1 f(y) dy_n$$

with estimators ($N = M^n$):

$$\tilde{I}_1 = \frac{1}{N} \sum_{j=1}^{N} f(\overline{x}_j),$$

where $\{\overline{x}_j\}_{1 \leqslant j \leqslant N}$ form uniform grid in $[0;1]^n$ cube and

$$\tilde{I}_2 = \frac{1}{N} \sum_{j=1}^{N} f(\overline{x}'_j),$$

where $\overline{x}'_j$ is uniform random point in j-th subcube.

Numerical analysis

Table: n=1

N	$\lvert I - \tilde{I}_1 \rvert$	$\lvert I - \tilde{I}_2 \rvert$
10^2	$5 * 10^{-3}$	$7 * 10^{-6}$
10^3	$5 * 10^{-4}$	$4 * 10^{-7}$
10^4	$5 * 10^{-5}$	$5 * 10^{-9}$
10^5	$5 * 10^{-6}$	$4 * 10^{-10}$
10^6	$5 * 10^{-7}$	$1 * 10^{-11}$
10^7	$5 * 10^{-8}$	$1 * 10^{-12}$

Table: n=2

N	$\lvert I - \tilde{I}_1 \rvert$	$\lvert I - \tilde{I}_2 \rvert$
10^2	10^{-1}	$6 * 10^{-5}$
10^4	10^{-2}	10^{-6}
10^6	10^{-3}	$5 * 10^{-7}$
10^8	10^{-4}	$3 * 10^{-9}$

Table: n=3

N	$\lvert I - \tilde{I}_1 \rvert$	$\lvert I - \tilde{I}_2 \rvert$
10^3	0.145	$2 * 10^{-3}$
10^6	0.0145	$6 * 10^{-6}$

Numerical analysis

Assume $f(\cdot)$ and $\frac{\partial f}{\partial x_k}$ are continuous and $\left|\frac{\partial f}{\partial x_k}\right| \leqslant L$, $1 \leqslant k \leqslant n$.
Then

Assertion 1

$$P\left(\left|I - \tilde{I}_2\right| \leqslant (nL/\varepsilon)N^{-1/2-1/n}\right) \geqslant 1 - \varepsilon^2, \quad 0 < \varepsilon < 1$$

and

Assertion 2

$\sup_f \left|I - \tilde{I}_1\right| \leqslant nLN^{-1/n}$

Quick sort

Consider array

$$(a[1], a[2], \ldots, a[N]) .$$

We want to sort it in direct (increasing) order.

```
// begin with QuickSort(a, 1, N)
QuickSort(a, begin, end)
if begin < end
  then q = Partition (a,begin, end);
  QuickSort(a, begin, q-1);
  QuickSort(a, q+1, end);
```

```
Partition(a,begin,end)
 x = a[end];
 i = begin - 1;
 for j = begin to (end - 1)
   if a[j] <= x
     then i = i+1;
     exchange a[i] and a[j]
 exchange a[i+1] and a[end];
 return i + 1;
```

Quick sort

$T(N) = ?$

```
// begin with QuickSort(a, 1, N)
QuickSort(a, begin, end)
if begin < end
  then q = Partition (a,begin, end);  // (end - begin) <= (N - 1) iterations
  QuickSort(a, begin, q-1);           // T(q -  begin) <= T(N - 1) iterations
  QuickSort(a, q+1, end);             // T(end - q) <= T(N - 1) iterations
```

In the worst case

$$T(N) = (N-1) + (N-2) + \ldots + 2 + 1 = \frac{(N-2) * (N-1)}{2} = O\left(N^2\right)$$

iterations when

$$a[1] < a[2] < \ldots < a[N].$$

Quick sort

But! If

$$P\left(a[i_1] < a[i_2] < \ldots < a[i_N]\right) = \frac{1}{N!}$$

for any permutation $(i_1, i_2, \ldots, i_N)$, then

$$\overline{T}(N) = E\left[T(N)\right] = O\left(N \log N\right)!$$

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Lets use pseudorandom numbers!

$$\overline{T}(N) = O\left(N \log N\right).$$

Matrix equations

A, B and $C \in R^{N \times N}$.

$$A \times B = C? \tag{1}$$

Complexity (determined algorithms): $O\left(N^3\right)$ or $O\left(N^{2.3}\right)$.

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$x \in \{0;1\}^N$ — uniform random vector. Instead of (1) check

$$A \times B \times x = C \times x? \tag{2}$$

Decision rule: if (2) is true (false) then (1) is true (false).

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Decision rule: if (2) is true (false) then (1) is true (false).
Complexity: $O\left(N^2\right)$. Error probability:

$$P\left(\text{"false"} | A \times B = C\right) = 0, \quad P\left(\text{"true"} | A \times B \neq C\right) \leqslant \frac{1}{2}$$

Repeat it 20 times and get error probability $< 10^{-6}$.

Integer equations

Let $x_1,\ x_2,\ \ldots,\ x_N \in Z_+$ and $Q\left(x_1, x_2, \ldots, x_N\right)$ be k-degree polinom.

$$Q\left(x_1, x_2, \ldots, x_N\right) \equiv 0? \tag{3}$$

No polinomial algorith.

Integer equations

Shwartz Lemm

If $\{x_i\}_{1 \leqslant i \leqslant N}$ are independent uniform random numbers from $\{0, 1, \ldots, n\}$ then

$$P\left(Q\left(x_1, x_2, \ldots, x_N\right) = 0 \middle| Q\left(\cdot, \cdot, \ldots, \cdot\right) \not\equiv 0\right) < \frac{kN}{n}.$$

Let $n = 2kN + 1$ then $P\left(\text{Error}\right) < 1/2$.
Check 100 times $\Rightarrow$ $P\left(\text{Error}\right) < 1/2^{100}$

What does it mean «random numbers»?

Psychological test

$$10001011101111010000 \quad (4)$$
$$01111011001101110001$$

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$$10001011101111010000 \quad (4)$$
$$01111011001101110001$$

$$00000000000000000000 \quad (5)$$
$$11111111111111111111$$

Remarks:

- $P(\xi_i = 1) = P(\xi_i = 0) = \frac{1}{2}$
- Consider only $(\xi_1, \xi_2, \ldots, \xi_N)$, where N is large $(N \to \infty)$

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Problem

Denote

$$\Omega = \{0, 1\}^{\infty}.$$

Want to find

$$R \subset \Omega \text{ — set of random sequences.}$$

General approach

Let $L\colon \Omega \to \{0,1\}$ be a characteristic property indicator of sequence $(a_1, a_2, \ldots)$.

$(a_1, a_2, \ldots)$ — random sequence $\Leftrightarrow L(a_1, a_2, \ldots) = 1$

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What L functions can you imagine?

Intuitive assumption

Consider sequence

$$(a_1, \quad a_2, \quad \ldots, \quad a_{N-1}, \quad a_N, \quad a_{N+1}, \ldots).$$

Obviously we want

$$\lim_{N\to\infty} \frac{\nu_N("1")}{N} = \frac{1}{2}, \tag{6}$$

where $\nu_N("1") = \sum_{i=1}^{N} Indicator\{a_i = 1\}$.

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What about sequence

0 1 0 1 0 1 0 1 ... ?

Definition 1

$(a_1, a_2, \ldots)$ is random if for any computable functions $F()$ and $G()$:

1. $n(1) = F(\Lambda), n(2) = F(a_{n(1)})$

$$\left(a_{n(1)}, a_{n(2)}, \ldots\right) \text{ – mixed}$$

2. Get $a_{n(k)} \Leftrightarrow G(a_{n(1)}, a_{n(2)}, \ldots, a_{n(k-1)}) = 1$

$$(\tilde{a}_1, \tilde{a}_2, \ldots) \text{ — result sequense}$$

true following

$$\lim_{N \to \infty} \frac{\nu_N("1")}{N} = \frac{1}{2}$$

Intuitive assumption

Let

$$(a_1, a_2, \ldots)$$

be a casino tool, that is gamer tries to predict these numbers.

1. Gamer knows nothing and he just predicts $(n(1), i(1), v(1))$; casino checks whether $i(1) = a_{n(1)}$; $V(1) = V(0) + / - v(1)$;
2. Gamer knows $a_{n(1)}$ and tries to predict the next number; he says $(n(2), i(2), v(2))$; $V(2) = V(1) + / - v(2)$;
3. ...

Every time gamer applies some decision function $\Xi()$, named "strategy".

Definition 2

Suppose gamer has $V(0) = 1$ money initially.

Definition

$(a_1, a_2, \ldots)$ is random if there is no computable strategy $\Xi()$, that implies

$$V(k) \xrightarrow[k \to \infty]{} \infty.$$

Theorem

Let U be class of «unpredictable» sequences and S be class of frequency stable sequences. Then

$$U \subset S.$$

Random and pseudorandom numbers

Figure: Random

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

Figure: Pseudorandom

Examples

- Flipping a coin or dices, dragging balls from an urn etc;
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Advantages

They are truly random! Nobody can argue with it.

Application: as a seed for pseudorandom numbers.

The first pseudorandom numbers generator

John von Neumann.

Suppose we have X_n. Calculate X_n^2 and get, for example, 10 digits from the middle as X_{n+1}.

Example:

$$X_n = 5772156649 \to X_n^2 = 33317792380594909201$$

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Disadvantage: a few numbers and hard to calculate.

Assertion

20-digits binary number generates at maximum 142 different numbers.

"Random numbers should not be generated with a method chosen at random"

D. E. Knuth "The Art of Computer Programming", 2, page 6

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Some theory should be used.

What numbers do we need?

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General approach

$$X_n \in \{0, 1, \ldots, m-1\} \to U_n = X_n/m \to \textit{ arbitrary distribution}$$

Hence we need $X_n \in \{0, 1, \ldots, m-1\}$.

Definition

Choose four numbers:

- m — the modulus; $0 < m$.
- a — the multiplier; $0 \leqslant a < m$.
- c — the increment; $0 \leqslant c < m$.
- X_0 — the starting value; $0 \leqslant X_0 < m$

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$$X_{n+1} = (aX_n + c) \mod m, \quad n \geqslant 0.$$

Example

For $m = 10$, $X_0 = a = c = 7$ we get

$$7,\ 6,\ 9,\ 0,\ 7,\ 6,\ 9,\ 0,\ \ldots$$

Congruential sequences always get into loop. Unfortunately.

Choise of modulus (m). $X_{n+1} = (aX_n + c) \mod m$

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- Period length $< m$.
- Generation speed.
  Using $m = 2^e$ numbers (e is a computer's word size, e.g.) is faster.
- Random properties.
  Suppose $m = 2^e \Longrightarrow$ the right-hand digits of X_n are much less random than the left-hand digits.

Assertion

If d is divisor of m and if $(Y_n = X_n \mod d)$ then

$$Y_{n+1} = (aY_n + c) \mod d.$$

You can use $m = 2^e - 1$ instead.

Choise of multiplier (a) and increment (c).

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

Goal: achieve the longest period (m). Is it possible?

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Theorem

The linear congruential sequence defined by m, a, c and X_0 has period length m if and only if

1. c is relatively prime to m;
2. $a - 1$ is a multiple of p, for every prime p dividing m;
3. $a - 1$ is a multiple of 4, if m is a multiple of 4.

Potency. $X_{n+1} = (aX_n + c) \mod m$

Suppose $\{X_n\}$ has maximum period (that is m).

Definition

The least integer s such that

$$(a-1)^s \equiv 0 \,(\ \textit{modulo}\ m)$$

called «potency».

The larger potency $\Longrightarrow$ the more random sequence.

Potency. $X_{n+1} = (aX_n + c) \mod m$

Let $X_0 = 0$ and $b = a - 1$, then

$$X_n = ((a^n - 1)\, c/b) \mod m,$$

after transformations:

$$X_n = c \left(n + \binom{n}{2} b + \ldots + \binom{n}{s} b^{s-1} \right) \mod m.$$

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In particular:

$s = 1 \Longrightarrow X_n = cn \mod m$;

$s = 2 \Longrightarrow X_n = cn + c\binom{n}{2} b \mod m \Longrightarrow X_{n+1} - X_n \equiv c + cbn$

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denoting $(d = cb \mod m)$ we have $(X_n, X_{n+1}, X_{n+2}) \in$

$$x - 2y + z = d + m, \qquad x - 2y + z = d - m,$$
$$x - 2y + z = d, \qquad x - 2y + z = d - 2m.$$

Potency. $X_{n+1} = (aX_n + c) \mod m$

Examples:

1. Let $m = 2^e$; it is divisible by high powers of prime 2. Choosing $a = 2^k + 1$, have relatevely big potency.

Potency. $X_{n+1} = (aX_n + c) \mod m$

Examples:

1. Let $m = 2^e$; it is divisible by high powers of prime 2. Choosing $a = 2^k + 1$, have relatevely big potency.
2. Let $m = 2^e - 1$; commonly it is not divisible by high powers of primes $\Longrightarrow$. There could not be a big potency.

"Chi-square" test. Experiment

Throw two "true" dices.

s — sum of the result numbers.

value of $s =$	2	3	4	5	6	7	8	9	10	11	12
probability, $p_s =$	$\frac{1}{36}$	$\frac{1}{18}$	$\frac{1}{12}$	$\frac{1}{9}$	$\frac{5}{36}$	$\frac{1}{6}$	$\frac{5}{36}$	$\frac{1}{9}$	$\frac{1}{12}$	$\frac{1}{18}$	$\frac{1}{36}$

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Repeat $n = 144$ times:

value of $s =$	2	3	4	5	6	7	8	9	10	11	12
observed number, $Y_s =$	2	4	10	12	22	29	21	15	14	9	6
probability, $p_s =$	$\frac{1}{36}$	$\frac{1}{18}$	$\frac{1}{12}$	$\frac{1}{9}$	$\frac{5}{36}$	$\frac{1}{6}$	$\frac{5}{36}$	$\frac{1}{9}$	$\frac{1}{12}$	$\frac{1}{18}$	$\frac{1}{36}$

"Chi-square" test. Statisctic

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Better to analyse:

$$V = \frac{(Y_2 - np_2)^2}{np_2} + \frac{(Y_3 - np_3)^2}{np_3} + \ldots + \frac{(Y_{12} - np_{12})^2}{np_{12}}. \qquad (8)$$

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In our example:

$$V = \frac{(2-4)^2}{4} + \frac{(4-8)^2}{8} + \ldots + \frac{(9-8)^2}{8} + \frac{(6-4)^2}{4} = 7\frac{7}{48}. \quad (9)$$

"Chi-square" test. Test

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

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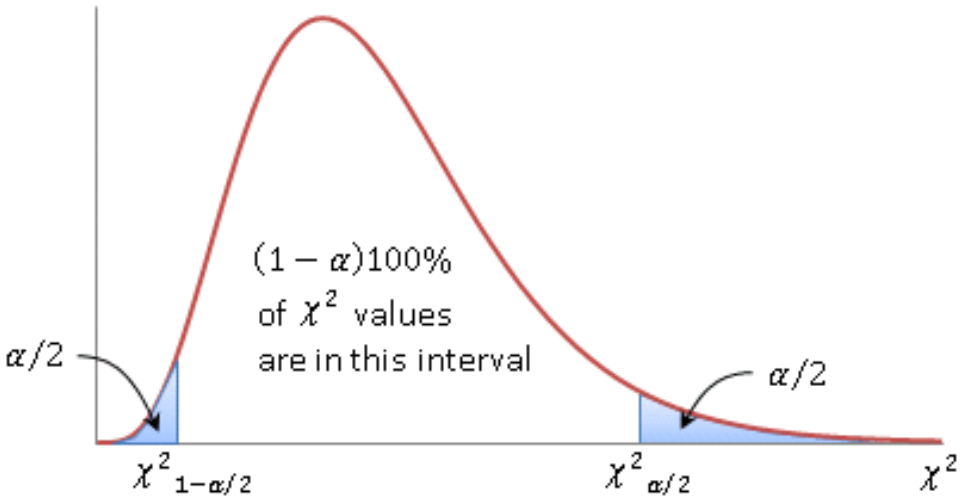

Figure: $\chi^2(k-1)$ density

$$P\Big(\chi^2_{1-\alpha/2}(k-1) \leqslant \leqslant V(Y_1, Y_2, \ldots, Y_k) \leqslant \leqslant \chi^2_{\alpha/2}(k-1) \,|\, V \in \chi^2(k-1)\Big) = = 1-\alpha$$

"Chi-square" test. Pseudorandom numbers generators app

Consider

$$A = \{0, 1, \ldots, m-1\}.$$

"Chi-square" test. Pseudorandom numbers generators app

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$$A = \{0, 1, \ldots, m-1\}.$$

Split A into k subsets (of size $\frac{m}{k}$, e.g.):

$$A = \left\{0, 1, \ldots, \frac{m}{k} - 1\right\} \cup \left\{\frac{m}{k}, \frac{m}{k} + 1, \ldots, 2\frac{m}{k} - 1\right\} \cup \ldots$$
$$\ldots \cup \left\{m - \frac{m}{k}, m - \frac{m}{k} + 1, \ldots, m - 1\right\}.$$

"Chi-square" test. Pseudorandom numbers generators app

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$$np_s = \frac{n}{k} \geqslant 5 \quad \text{better} \quad \geqslant 8 \quad \text{better} \quad \geqslant 10$$