Note

Matrix Generator of Pseudorandom Numbers

In the work [1] it was proposed to use the Komogorov's K-systems as the pseudorandom numbers generators for the Monte Carlo simulation of physical processes. The main idea of this approach is as follows: The pseudorandom series $\{P_N\}$ is represented by the trajectory of a particular unstable dynamical system, and the phase space is a Π^d -hypercube in the space of dimension d. The system must be maximally unstable in order for the trajectory to fill this hypercube Π^d uniformly. As it is well known, these are the Kolmogorov's K-systems.

In [1] it was proposed to use for that the automorphisms of compact commutative groups, which are defined by the integer matrix $A = ||a_{ij}||$

$$P_N = AP_{N-1} \mod 1,\tag{1}$$

where P is the d-dimensional vector, belonging to Π^d , $P_i = (X_1^{(i)}, X_2^{(i)}, ..., X_d^{(i)})$, and

$$\det A = \pm 1 \tag{2}$$

in order to preserve phase space volume Π^d .

To ensure the automorphism (1) to be a K-system, all its eigenvalues must satisfy the condition:

$$|\lambda_K| \neq 1, \qquad K = 1, ..., d \tag{3}$$

Thus, the problem of the pseudorandom sequence $\{P_N\}$ construction reduces to the construction of the matrix A which satisfies the conditions (2) and (3).

Here we propose a simple recurrent way for construction of such a matrix A of arbitrary order d. Suppose we have a matrix of the dimension $||d-1 \otimes d-1||$ with det A = 1. Construct now a new matrix the type $||d \otimes d||$:

$$A = \begin{vmatrix} 1 & 0 & 0 & \cdots & 0 \\ 0 & & & \\ \vdots & & \\ & \{ \|d - 1 \otimes d - 1 \| \} \\ 0 & \end{vmatrix}$$
(4)

It is evident that det A is also equal to 1.

Since the calculation of det A is invariant with respect to the addition of the lines and columns, one can get nontrivial matrices which have different elements in the first line and in the first column, without changing the value of det A. This procedure allows us to get a large set of matrices, which are suitable for the automorphism (1). One needs, however, to check whether all $|\lambda_k|$ are not equal to one. We propose for the matrices A the form:

$$A_{d} = \begin{vmatrix} 2 & 3 & \cdots & d & & 1 \\ 1 & 2 & \cdots & d - 1 & & 1 \\ 1 & 1 & \cdots & d - 2 & & 1 \\ \vdots & \cdots & \cdots & & \cdots & \\ 1 & 1 & \cdots & 2 & 2 & 1 \\ 1 & 1 & \cdots & 1 & 2 & 1 \\ 1 & 1 & \cdots & 1 & 1 & 1 \end{vmatrix};$$
(5)

if d = 2, 3, 4, we have

$$A_{2} = \left\| \begin{array}{c} 2 & 1 \\ 1 & 1 \end{array} \right\|, \qquad A_{3} = \left\| \begin{array}{c} 2 & 2 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 1 \end{array} \right\|, \qquad A_{4} = \left\| \begin{array}{c} 2 & 3 & 4 & 1 \\ 1 & 2 & 2 & 1 \\ 1 & 1 & 2 & 1 \\ 1 & 1 & 2 & 1 \\ 1 & 1 & 1 & 1 \end{array} \right|.$$
(6)

We have checked condition (3) for the matrix A_d in the wide range of d = 2, ..., 170). All cases checked were fulfilled to the condition (3). An example of such calculation for d = 170 is given in the Table I. The calculation shows that all $|\lambda_k|$ are in the range $\sim 0.2 \le |\lambda_k| \le 1539.9$ and that the Kolmogorov's entropy

$$h(A) = \sum_{|\lambda_k| > 1} \ln |\lambda_k|$$

increases with d:

$$h(A_{75}) \approx 47.5;$$
 $h(A_{110}) \approx 70.1;$ $h(A_{170}) \approx 108.9.$

From that we obtain the characteristic time for the decay of correlations [1],

$$\tau \leqslant 1/h(A_d),$$

which decreases with d.

In order to realize the transformation (1) and to construct the concrete trajectory, one must define the initial vector,

$$P_0 = (X_1^{(0)}, ..., X_d^{(0)}), \tag{7}$$

TABLE I

IABLE I								
		Relk	lmλ _K	Relk	lm λ _κ			
N=	170	-0.58970	0.92975	0.14809	-0.32080			
Reλ _κ	Imλ _K	-0.58970	-0.02075	0.14809	- 0.320 80			
1539.88525	0.00000	-0.51913	0.88842	0.15639	0.30852			
-177.21851	636.29004	-0.51913	-0.88842	0.15639	-0.30852			
-177.21851	-636.29004	-0.45855	0.85218	0.16517	0.29533			
-167.06267	125.76009	-0.45855	-0.85128	0.16517	-0.29533			
-167.06267	-125.76009	-0.39966	0.81641	0.17201	0.28229			
-83.64687	39.28154	-0.39966	-0.81641	0.17201	-0.28229			
-83.64687	-39.28154	-0.34836	0.78305	0.17826	0.26946			
-47.59344	18.16008	-0.34836	-0.78305	0.17826	-0.26946			
-47.59344	-18.16008	-0.30112	0.74994	0.18384	0.25821			
-30.22507	10.69967	-0.30112	-0.74994	0.18384	-0.25821			
-30.22507	-10.69967	-0.25926	0.71939	0.18947	0.24681			
-20.73701	7.32249	-0.25926	-0.71937	0.18947	-0.24681			
-20.73701	-7.32249	-0.22253	0.69273	0.19654	0.23554			
-15.03566	5.50904	-0.22253	-0.69273	0.19654	-0.23554			
15.03566	-5.50904	-0.18580	0.66651	0.20162	0.22382			
-11.35427	4.40811	-0.18580	-0.66651	0.20162	-0.22382			
-11.35427	-4.40811	-0.15308	0.64134	0.20611	0.21215			
-8.84335	3.67899	-0.15308	-0.64164	0.20611	-0.21215			
-8.84335	-3.67899	-0.12206	0.61628	0.20975	0.20167			
-7.05865	3.16208	-0.12206	-0.61628	0.20975	-0.20167			
-7.05865	-3.16208	-0.09630	0.59228	0.21512	0.19130			
-5.74103	2.77609	-0.09630	-0.59228	0.21512	- 0 .19130			
-5.74103	-2.77609	-0.07119	0.57082	0.21869	0.19130			
-4.74529				1				
	2.47590	-0.07119	-0.57082	0.21869	-0.18045			
- 4.7452 9 -3.97216	-2.47590	-0.04841	0.55041	0.22288	0.16913			
-	2.47590	-0.04841	-0.55041	0.22288	-0.16913			
-3.97216	-2.23576	-0.02511	0.52987	0.22491	0.15862			
-3.36257	2.04017	-0.02511	-0.52987	0.22491	-0.15862			
-3.36257	-2.04017	-0.00614	0.50993	0.22840	0.14869			
-2.86968	1.87670	-0.00614	-0.50993	0.22840	-0.14869			
-2.86968	-1.87670	0.01222	0.49053	0.23084	0.13927			
-2.46762	1.73665	0.01222	-0.49053	0.23084	-0.13927			
-2.46762	-1.73665	0.02869	0.47272	0.23412	0.12944			
-2.13296	1.61385	0.02869	-0.47272	0.23412	-0.12944			
-2.13296	-1.61385	0.04480	0.45668	0.23747	0.12944			
-1.85761	1.50791	0.04480	-0.45668	0.23747	- 0 .11905			
-1.85761	-1.50791	0.06077	0.43943	0.23908	0.10917			
-1.62230	1.14107	0.06077	-0.43943	0.23908	-0.10917			
-1.62230	-1.14107	0.07471	0.42273	0.24025	0.09898			
-1.42336	1.33389	0.07471	-0.42273	0.24025	-0.09898			
-1.42336	-1.33389	0.56972	0.00000	0.24951	0.00410			
-1.25055	1.25791	0.08758	0.40587	0.24951	-0.00410			
-1.25055	-1.25791	0.08758	-0.40587	0.25096	0.01317			
-1.10046	1.18935	0.09808	0.39055	0.25096	-0.01317			
-1.10046	-1.18935	0.09808	-0.39055	0.25051	0.02301			
-0.97255	1.12860	0.11000	0.37730	0.25051	-0.02301			
-0.97255	-1.12860	0.11000	-0.37730	0.27938	0.03312			
-0.85875	1.07222	0.12128	0.36257	0.24938	-0.03312			
-0.85875	-1.07222	0.12128	-0.36257	0.24912	0.04222			
-0.75967	1.02177	0.13192	0.34737	0.24912	-0.04222			
-0.75967	-1.02177	0.13192	-0.34737	0.24512	0.05101			
-0.66924	0.97531	0.13932	0.33334	0.24711	-0.05101			
-0.66924	-0.97531	0.13932	-0.33334	0.24773	0.06057			

and act on it by matrix A_d , so that the result will be

$$P_{1} = \begin{pmatrix} X_{1}^{(1)} \\ \vdots \\ X_{d}^{(1)} \end{pmatrix} = \begin{pmatrix} a_{11} \cdots a_{1d} \\ \cdots \\ a_{d1} \cdots \\ a_{dd} \end{pmatrix} \begin{pmatrix} X_{1}^{(0)} \\ \vdots \\ X_{d}^{(0)} \end{pmatrix} = AP_{0} \pmod{1}.$$
(8)

Then, operating on P_1 by A_d again, we find vector P_2 ,

$$P_{2} = \begin{pmatrix} X_{1}^{(2)} \\ \vdots \\ X_{d}^{(2)} \end{pmatrix} = \begin{pmatrix} a_{11} \cdots a_{1d} \\ \cdots \\ a_{d1} \cdots \\ a_{dd} \end{pmatrix} \begin{pmatrix} X_{1}^{(1)} \\ \vdots \\ X_{d}^{(1)} \end{pmatrix} = AP_{1} \pmod{1},$$
(9)

and so on:

$$P_{N} = \{A\{A \cdots \{AP_{0}\}\} \cdots \}.$$
(10)

The points P_0 , P_1 , ..., P_N form the very trajectory in the hypercube Π_d . In order to check the statistical features of the sequence (10) we have to calculate χ_s^2 and the discrepancy D_N [1], which determines the convergence of the Monte Carlo sums.

We arrange the components of $P_0, P_1, ...$ in the series

$$X_{1}^{(0)}, ..., X_{d}^{(0)}, X_{1}^{(1)}, ..., X_{d}^{(1)}, ..., X_{1}^{(2)}, ..., X_{d}^{(2)},$$
(12)

As in the case of the multiplicative generator one now can compose *the words* of length K and represent them as the points of the Π^{K} hypercube. Varying K from 1 to d and beyond, we obtain points that fill the cube of the varying dimension K. Using the χ_{s}^{2} criterion one can estimate the degree of uniformity of filling the cube Π^{K} for different K.

We have considered χ_s^2 when K=1, 2, 3, 4 for the matrices A_d with different d. Each side of Π^K was divided into ten parts, so we have

$$\chi_{s-1}^{2} = \chi_{9}^{2} \qquad \text{for} \quad \Pi^{1},$$

$$\chi_{s-1}^{2} = \chi_{99}^{2} \qquad \text{for} \quad \Pi^{2},$$

$$\chi_{s-1}^{2} = \chi_{999}^{2} \qquad \text{for} \quad \Pi^{3},$$

$$\chi_{s-1}^{2} = \chi_{9999}^{2} \qquad \text{for} \quad \Pi^{4},$$

(13)

where s-1 is the number of degrees of freedom: $s = 10^{K}$.

Let us write down values of χ^2_{s-1} for different matrices (5):

A₂,
$$\chi_9^2 = 7.9$$
 $X_1^{(0)} = \pi/7$
 $\chi_{99}^2 = 106.6$ $X_2^{(0)} = 1/\sqrt{2}$
 $\chi_{9999}^2 = 34,046$
 $\chi_{9999}^2 = 5,769,697$

In all those examples N/s = 24, where N is the full amount of points and s - 1 is the number of degrees of freedom. For comparison we give below the values of χ^2_{s-1} for the standard multiplicative generator RNDM,

$$\chi_{9}^{2} = 8.9$$

$$\chi_{99}^{2} = 97.8$$

$$\chi_{999}^{2} = 1007.5$$

$$\chi_{9999}^{2} = 10,301.1.$$

It is seen that the statistical properties become better with the growth of the dimension of matrix A_d . Also we have calculated the dependence of χ^2_{s-1} on the total number of generated points for the matrix A_{12} .

N/s	48	72	96	144
χ ² χ ² ₉₉ χ ² ₉₉₉	7.9	10.1	6.3	9.83
χ^2_{99}	82	105.3	91.8	95
χ^{2}_{999}	1021	968.5	1006.9	999.5
χ ² 9999	9955.3	9976.4	10,069	9846
		RNDM		
N/s	48	RNDM 72	96	144
	48		96	144
		72		
$\frac{N/s}{\chi_{9}^{2}}$	5.5	72	6.3	12.2

A₁₂

One can also calculate mean value $\bar{\chi}_{s-1}^2$,

$$\bar{\chi}_{s-1}^{2} = \frac{\sum \chi^{2}(i) \cdot N_{i}/s - 1}{\sum_{i} N_{i}/s - 1}$$

for A_{12} and RNDM,

<i>A</i> ₁₂	RNDM				
9.41	9.1				
90.9	104.8				
1001.4	1035.2				
9935.6	10321.5				

As we mentioned before, the convergence of the Monte Carlo sums is connected with the behavior of the discrepance D_N . Below we give the RNDM and matrix generator A_{100} when K=4:

				RN	DM					
$N \cdot 10^5$	1	2	3	4	5	6	7	8	9	10
$\begin{array}{c} D_N \sqrt{N} \\ \chi^2_{9999} \end{array}$	1.714 10016	1.930 10081	1.944 10455	2.071 10202	2.355 10280	2.610 10197	2.937 10139	3.164 10199	3.204 10269	3.302 10226
				A	100					
$N \cdot 10^5$	1	2	3	4	5	6	7	8	9	10
$D_N \sqrt{N} \\ \chi^2_{9999}$	1.233 10049	1.350 10086	1.287 9870	1.141 9751	1.264 9748	1.587 9858	1.470 9857	1.410 9901	1.228 9864	1.102 9952

All this shows that those matrix generators (we call them "mixmax") have good statistical properties and are usable for Monte Carlo simulation. The great advantage of the matrix generator is the possibility to change widely the components of the initial vector $P_0 = X_1^{(0)}$, ..., $X_d^{(0)}$ and matrix A_d in order to have the dependence of the Monte Carlo calculations on these parameters, which, naturally, must be weak. The proposed matrix generator can be successfully realized on computers with a matrix processor.

Reference

RECEIVED: November 16, 1989; REVISED: June 11, 1990

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