

High Quality Random Number Generator using Programmable Cellular Automata



By

Azfar Shakeel Khan
2007-NUST-MS-PhD IT-23

Supervisor

Dr. Nazar Abbas Saqib

Associate Professor
NUST- SECS

A thesis submitted in partial fulfillment of the requirements for the degree
of Masters of Science in Information Technology (MS IT)

in

School of Electrical Engineering and Computer Science,
National University of Sciences and Technology (NUST),
Islamabad, Pakistan.

(April 2011)

APPROVAL

It is certified that the contents and form of thesis entitled "High Quality Random Number Generator using Programmable Cellular Automata" submitted by Mr. Azfar Shakeel Khan have been found satisfactory for the requirement of the degree.

Advisor: Dr. Nazar Abbas Saqib

Signature: _____

Date: _____

Committee Member1: Dr. Fauzan Mirza

Signature

Date:

Committee Member2: Dr. Zahid Anwar

Signature

Date:

Committee Member3: Mr. Ammar Karim

Signature

Abstract

Random Numbers have applications in a number of computer sciences domains specifically in computer security algorithms they play a vital role. One of the techniques to generate a random number is to follow an algorithm which tries to generate random numbers as close as a true random number generator source, this algorithm based random number generation is known as Pseudo random Number Generation (PRNG). In this research we studied existing techniques of Cellular Automata (CA) based Pseudo Random generator (PRNG) and proposed 5 different variations in the existing algorithm, modified the CA rules as per requirement for the proposed techniques i.e. 3D Neuman and Moore neighborhood, 2D to 3D jumping Neuman and 2D Neuman and Moore jumping neighborhoods. In all the cases Neuman neighborhood means non diagonal adjacent neighbors while Moore neighborhood means diagonally adjacent neighbors. CA rules define which neighbors participate in the current cycle for random number generation.

We implemented the simulation code in C++ and tested the outputs in internationally accepted standard for Random Numbers "DIEHARD", the analysis of the results are made as per recommendations of ANSI standards and it is discussed which of the above techniques can be declared to be best in terms of quality of randomness in random numbers generation.

Certificate of Originality

I hereby declare that this submission is my own work and to the best of my knowledge it contains no materials previously published or written by another person nor material which to a substantial extent has been accepted for the award of any degree or diploma at SEecs or at any other educational institute, except where due acknowledgement has been made in the thesis.

Any contribution made to the research by others, with whom I have worked at SEecs or elsewhere, is explicitly acknowledged in the thesis.

I also declare that the intellectual content of this thesis is the product of my own work, except for the assistance from others in the project's design and conception or in style, presentation and linguistics which has been acknowledged.

Author Name: Azfar Shakeel Khan

Signature:

Acknowledgments

First and foremost, I am immensely thankful to Almighty Allah for letting me pursue and fulfill my dreams. Nothing could have been possible without His blessings.

I am deeply indebted to my supervisor, Dr Nazar Abbas Saqib, for providing me the supervision, motivation and encouragement throughout the span of my work. His insight, breadth of knowledge, and enthusiasm has been invaluable. Without his care and able guidance, I would not have been able to complete my thesis.

I am also grateful to my committee members, Dr Fauzan Mirza, Dr Zahid Anwar and Mr. Ammar Karim for their valuable suggestions and guidance in my research. They were abundantly helpful, and I owe a lot to them.

I am thankful to National University of Science and Technology (NUST), Pakistan for providing me scholarship due to which I carried out my research activities with good mind set.

Azfar Shakeel Khan

Table of Contents

Chapter 1	Introduction	i
1.1	Random Number	ii
1.2	Motivation	ii
1.3	Problem Statement.....	iii
1.4	Goals and Objectives of the Research.....	iii
1.5	Scope of the Research	iv
1.6	Thesis outline	iv
Chapter 2	Background	vi
2.1	Types of random Numbers	vii
2.2	DIEHARD Test and its Details.....	xii
2.2.1	DIEHARD Tests Applied	xiii
2.3	Simulation Environment.....	xvii
Chapter 3	Literature Review/ Related work	xviii
3.1	Literature Review	xix
3.2	Related Work.....	xxi
Chapter 4	Methodology.....	xxv
4.1	Literature Survey Phase	xxvi
4.2	Idea Generation Phase	xxvi
4.3	Proposed Solution Implementation Phase.....	xxvi
4.4	Testing Phase	xxvii
4.5	Analysis of Results Phase.....	xxvii
4.6	Assumptions	xxvii
Chapter 5	Proposed Solutions	xxix
5.1	Proposed Methods.....	xxx

5.2 Algorithm used	xxxix
5.2.1 2D Neumann Neighbors.....	xxxix
5.2.2 2D Moore + Neumann Neighbors.....	xxxix
5.2.3 2D Neumann Moore jumping Neighbors	xxxix
5.2.4 3D Neumann Neighbors.....	xxxix
5.2.5 3D to 2D Jumping Neumann Neighbors.....	xli
5.2.6 3D Moore + Neumann Neighbors.....	xli
Chapter 6 Testing of Proposed Solutions.....	xlvi
6.1 DIEHARD Tests Applied	xlvi
6.2 Testing Process	xlix
Chapter 7 Analysis of Results	I
7.1 Criteria 1	lii
7.2 Criteria 2.....	lii
7.3 Comparative Analysis of Results.....	lvi
7.4 Run Time Analysis of the Proposed Methods	lvi
7.4.1 Main Memory Requirements.....	lvii
7.4.2 Number of operations per iteration	lviii
Chapter 8 Conclusion and Future Work	lxi
8.1 Conclusion	lxii
8.2 Future Work.....	lxiii
References	lxiv
Appendix A	lxvi
Appendix B	lxxi
Appendix C	lxxxv
Appendix D.....	lxxxvi
Appendix E	lxxxvii

List of Figures

Figure 1: Random Number Generator	xi
Figure 2: CA Rules Implemented for 2D-PCA	xxi
Figure 3: 2D-PCA Structure	xxii
Figure 4: Results for Different Logical Operations with CA Rules	xxiii
Figure 5: DIEHARD Results Compared using Safe Window Range	xxiv
Figure 6: 2D Neumann Neighbors Illustration	xxxii
Figure 7: 2D Neumann Cell Structure	xxxiii
Figure 8: 2D Moore + Neumann Neighbors Illustration	xxxiv
Figure 9: 2D Moore + Neumann Neighbors Cell Structure	xxxv
Figure 10: 2D Neumann Moore jumping Neighbors Illustration	xxxvi
Figure 11: 2D Neumann Moore jumping Neighbors Cell Structure	xxxvii
Figure 12: 3D Lattice of cells used for all the 3D PCA Algorithms	xxxix
Figure 13: 3D Neumann Neighbors Illustration	xxxix
Figure 14: 3D Neumann Neighbors Cell Structure	xl
Figure 15: 3D-2D jumping Neumann Neighbors Illustration	xlii
Figure 16: 3D to 2D Jumping Neumann Neighbors Cell Structure	xliii
Figure 17: 3D Moore + Neumann Neighbors Illustration	xliv
Figure 18: 3D Moore + Neumann Neighbor Cell Structure	xl v
Figure 19: Testing Phases block diagram	xlix
Figure 20: Snap shot of results	li
Figure 21: Courtesy: en.wikipedia.org/wiki/Standard_deviation	lii
Figure 22: Courtesy: IJCSNS International Journal of Computer Science and Network Security, VOL.10 No.4, April 2010	lii
Figure 23: Results and graph showing values in the range $\mu \pm 3\sigma$	liiii
Figure 24: Results and graph showing values in the range $\mu \pm 2\sigma$	liv
Figure 25: Values falling in the Safe Range (values ≥ 0.25 AND values ≤ 0.75)	lv

Chapter 1 Introduction

1.1 Random Number

Random numbers have applications in many areas from simulation, game-playing, cryptography, statistical sampling, and evaluation of multiple integrals, particle transport calculations, and computations in statistical physics, to name a few.

Random numbers are numbers that occur in a sequence such that two conditions are met, first the values are uniformly distributed over a defined interval or set, and secondly it is impossible to predict future values based on past or present ones. Random numbers are important in statistical analysis and probability theory.

Although it may look simple at first sight to give a definition of what a random number is, it proves to be quite difficult in practice.

A random number is a number generated by a process, whose outcome is unpredictable, and which cannot be sub sequentially reliably reproduced. This definition works fine provided that one has some kind of a black box, such a black box is usually called a random number generator that fulfills this task.

However, if one were to be given a number, it is simply impossible to verify whether it was produced by a random number generator or not. In order to study the randomness of the output of such a generator, it is hence absolutely essential to consider sequences of numbers.

It is quite straightforward to define whether a sequence of infinite length is random or not. This sequence is random if the quantity of information it contains in the sense of Shannon's information theory is also infinite.

1.2 Motivation

As the discussions made in the above sections, there are lot of applications in the domain of computer sciences, IT and other fields of research where random numbers play a vital role, studying existing solutions to generate a random sequence and what possibilities are there to improve the quality of randomness and efficiency of random

number generators is a vast and interesting topic, its very motivating to find a new or better random number generator or discovering better technique for implementing the existing solutions as this addition in the knowledge domain has practical applications and can be useful for many other applications.

1.3 Problem Statement

After doing a comprehensive literature survey, understanding of what random numbers are, their properties, testing quality of randomness we have selected the topic of Cellular Automata to work on, there is vast scope in the field of cellular automata specifically with respect to the two and three dimensional structures as very less people have worked on this area and comprehensive results specifically with respect to a three dimensional cellular automata are not found in the literature, So we have decided to narrow down the area of interest, following statement can be treated as our problem statement.

“High Quality PRNG exploring 3D PCA Structure and possibilities for 2D to 3D jumping PCA”

1.4 Goals and Objectives of the Research

The objectives of the research is to add new ideas with respect to neighbor selection criteria in programmable cellular automata and their impact on the quality of randomness, one of the goals is finding a better random generator than the existing solutions, this opens a new track of research for the future as well as all possible neighborhood combinations can not be covered in a single research.

Our specific goals to be achieved are

- Implementation and Results for 2-D PCA [both for von Neumann and Moore neighborhood].
- 3-D PCA Implementation with All Neighbors.

- 2-D to 3-D jumping PCA where the Z- Neighbors are selected depending on the z-index control register, this area is never explored before.

1.5 Scope of the Research

The scope of this research is limited to programmable cellular automata for 3D lattice of cells and its variations for 3D to 2D jumping, both the Neumann and Moore neighbors are considered, the work studies choosing different neighbors and their effect on the quality of random number sequence generated, this work does not include FPGA implementation details and their hardware complexities due to limited amount of time.

1.6 Thesis outline

The document has the following chapters.

Chapter 2: Background and Experimental Setup

Discusses in general about the types of random numbers, various commonly used generators and about Cellular Automata generators, it also describes about the background needed for testing and simulation environment we used to accomplish the proposed solutions.

Chapter 3: Literature Review/ Related work

Discusses the most relevant research papers regarding the programmable cellular automata, the authors approach and the results they found.

Chapter 4: Methodology

Discuss the work flow and steps taken to accomplish this research work.

Chapter 5: Proposed Solutions

Detail discussion about the solutions proposed, the algorithm, mechanism of working for picking different neighbors and the programmable cellular automata rules applied.

Chapter 6: Testing of Proposed Solutions

In this chapter testing strategy is discussed, the tests we applied and the process of testing of proposed random number generators is explained.

Chapter 7: Analysis of Results

This chapter describes the analysis of all the results we got as a result of testing; the criteria of analysis, memory and run time cost comparisons are also discussed.

Chapter 8: Conclusion and Future Work

This chapter gives the concluding remarks, which random number generator is the best among the proposed methods and also what future work can be done if someone wants to extend this piece of work.

At the end of report the Appendices discuss important code and data structures, DIEHARD dumps and detail results of each implemented solution.

Chapter 2 Background

In this chapter first we will discuss some background about types of random numbers and commonly used random number generators and then the tools and technologies required to accomplish the implementation of our proposed solutions, for simulation of each proposed generator we used standard C++ compiler, for testing the DIEHARD tool is used and for comparison of results, calculating means and standard deviations and other analysis Microsoft Excel is used.

2.1 Types of random Numbers

In this section we briefly discuss the major categories of Random Number generators.

2.1.1 True Random Numbers or Physical Random-Number Generators

Games of chance are the classic examples of random processes, and the first choice would be to use traditional gambling devices as random-number generators.

Unfortunately, these devices are rather slow, especially since the typical computer application may require millions of numbers per second. Also, the numbers obtained from such devices are not always truly random e.g. cards may be imperfectly shuffled, dice and wheels may not be balanced, the random sequence obtained through the physical sources are thus known as physical random numbers or true random numbers, other physical phenomenon can also be considered in this regard, like the little variations in somebody's mouse movements or in the amount of time between keystrokes, a really good physical phenomenon to use is a radioactive source. The points in time at which a radioactive source decays are completely unpredictable, and they can quite easily be detected and fed into a computer. Another suitable physical phenomenon is atmospheric noise, which is quite easy to pick up with a normal radio.

Similarly background noise from an office or laboratory or even from a car engine can also be used as source.

One important shortcoming of any physical generator is the lack of reproducibility.

2.1.2 Pseudo Random Number Generators or Arithmetical Random Generators

The most common method of generating pseudo-random numbers on the computer uses a recursive technique called the linear-congruential, or Lehmer, generator. The sequence is defined on the set of integers by the recursion formula

$$X_{n+1} = AX_n + C \pmod{m}$$

One method to improve the pseudo-random number generators is to use a combination of two or more unrelated generators.

Some other well known PRNG techniques are discussed below very briefly.

Linear Feed back Shift Registers (LFSR)

A feedback shift register is made up of two parts: a shift register and a feedback function. The shift register is a sequence of bits. (The length of a shift register is figured in bits; if it is n bits long, it is called an n -bit shift register.) Each time a bit is needed, all of the bits in the shift register are shifted 1 bit to the right. The new left-most bit is computed as a function of the other bits in the register. The output of the shift register is 1 bit, often the least significant bit. The period of a shift register is the length of the output sequence before it starts repeating.

The simplest kind of feedback shift register is a linear feedback shift register, or LFSR. The feedback function is simply the XOR of certain bits in the register; the list of these bits is called a tap sequence. Sometimes this is called a Fibonacci configuration.

Geffe Generator

This key stream generator uses three LFSRs, combined in a nonlinear manner

Two of the LFSRs are inputs into a multiplexer, and the third LFSR controls the output of the multiplexer. If a_1 , a_2 , and a_3 are the outputs of the three LFSRs, the output of the Geffe generator can be described by

$$b = (a_1 \wedge a_2) \cdot ((\neg a_1) \wedge a_3)$$

a variation is known as Generalized Geffe Generator, Instead of choosing between two LFSRs, this scheme chooses between k LFSRs, as long as k is a power of 2. There are $k + 1$ LFSRs total. LFSR-1 must be clocked $\log_2 k$ times faster than the other k LFSRs.

Similar other variations of using LFSR in different combinations are, Jennings Generator, Beth-Piper Stop-and-Go Generator, Alternating Stop-and-Go Generator and Bilateral Stop-and-Go Generator.

Self-Decimated Generators

Self-decimated generators are generators that control their own clock. When the output of the LFSR is 0, the LFSR is clocked d times. When the output of the LFSR is 1, the LFSR is clocked k times.

Shrinking Generator

The shrinking generator uses a different form of clock control than the previous generators. Take two LFSRs: LFSR-1 and LFSR-2. Clock both of them. If the output of LFSR-1 is 1, then the output of the generator is LFSR-2. If the output of LFSR-1 is 0, discard the two bits, clock both LFSRs, and try again.

A5 is the stream cipher used to encrypt GSM (Group Special Mobile). A5 consists of three LFSRs; the register lengths are 19, 22, and 23; all the feedback polynomials are sparse. The output is the XOR of the three LFSRs. A5 uses variable clock control. Each register is clocked based on its own middle bit, XORed with the inverse threshold function of the middle bits of all three registers. Usually, two of the LFSRs clock in each round.

Additive Generators

Additive generators (sometimes called lagged Fibonacci generators) are extremely efficient because they produce random words instead of random bits, basically these are the combinations of techniques discussed above, Fish, Pike and Mush are some of the examples of additive generators.

2.1.3 Properties of a Good Random Number Generator

If we go through the literature about random number generators and their qualities we get the following points for a good random generator to have:

- Random pattern : Passes statistical tests of randomness
- Long period : Goes as long as possible before repeating
- Efficiency : Executes rapidly and requires little storage
- Repeatability : Produces same sequence if started with same initializations
- Portability : Capable of producing same results on different computers

2.1.4 The Cellular Automata and its types

The concept of Cellular Automata (CA) was initiated in the early 1950's by J. Von Neumann and Stan Ulam, Cellular automata can be viewed as a simple model of a

spatially extended decentralized system made up of a number of individual components (cells). The communication between constituent cells is limited to local interaction.

Each individual cell is in a specific state which changes over time depending on the states of its local neighbors.

The reason behind the popularity of cellular automata is their simplicity in understanding and implementation, different structural variations of CA have been proposed in their designs and analysis of the CA to make it versatile for modeling purposes.

The two variations of neighborhood are termed as Von Neumann and Moore neighborhood, Neumann means adjacent Non Diagonal and Moore means adjacent Diagonal neighbors.

Cellular automata on multi-dimensional grids have also been proposed. The grids have either null or periodic boundary. In null boundary configurations the boundary cells are assumed to have 'null' dependency. A periodic boundary is one in which the grid is considered to be folded.

The nature of next state functions also varies significantly; researchers have defined the rule set according to the design requirements of the applications. Also there are some standard rule sets, in all cases, the output of next state depends upon the output of the previous state.

In general a random number generator can be treated as a block box to which we give initial seed and it is assumed to give random number sequence as output for a desired length.

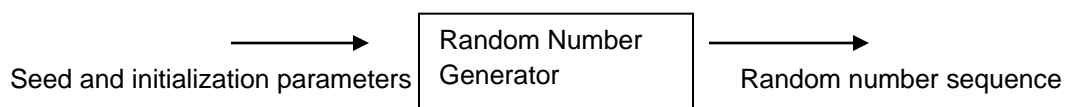


Figure 1: Random Number Generator

The multidimensional grid of a cellular automata can be found in various forms like 1D cellular automata, where the cells are placed adjacent to each other in linear fashion, 2D cellular automata where the lattice of cells are placed in a matrix structure that can be viewed in rows and columns, similarly in 3D cellular automata the structure can be viewed in the form of planes, rows and columns.

In each configuration each cell has certain neighbors, the number of neighbors considered is known as the radius, thus a radius of 1 means adjacent left and right neighbors in case of 1D cellular automata and four neighbors i.e. left, top, right, bottom in case of 2D cellular structure.

When we talk about a cellular automaton greater than one dimension there is also concept of Neumann and Moore neighbors in cellular automata, the Non Diagonally adjacent neighbors are known as Neumann while the diagonally adjacent neighbors are known as Moore neighbors.

2.2 DIEHARD Test and its Details

DIEHARD is Internationally Accepted for Randomness Testing and treated as de facto standard for testing RNGs.

DIEHARD is suit of tests, it basically comprises of 18 rigorous statistical tests, each test checks the randomness of a series of numbers on a specific parameter defined.

Most of the tests in DIEHARD return a p-value, which should be uniform on $[0,1)$ if the input file contains truly independent random bits. Those p-values are obtained by $p=F(X)$, where F is the assumed distribution of the sample random variable X , which is often normal.

2.2.1 DIEHARD Tests Applied

We applied the following tests of DIEHARD on all the proposed solutions.

- BIRTHDAY SPACINGS TEST
- THE OVERLAPPING 5-PERMUTATION TEST
- BINARY RANK TEST
- THE BITSTREAM TEST
- OPSO(Overlapping-Pairs-Sparse-Occupancy) TEST
- OQSO(Overlapping-Quadruples-Sparse-Occupancy) TEST
- DNA TEST
- COUNT-THE-1's TEST
- PARKING LOT TEST
- THE MINIMUM DISTANCE TEST
- THE 3DSPHERES TEST
- SQUEEZE TEST
- OVERLAPPING SUMS TEST
- RUNS TEST
- CRAPS TEST

Here we briefly discuss the internal workings of each test applied.

- BIRTHDAY SPACINGS TEST

Choose m birthdays in a year of n days. List the spacing's between the birthdays. If j is the number of values that occur more than once in that list, then j is asymptotically Poisson distributed with mean $m^2/(4n)$.

- THE OVERLAPPING 5-PERMUTATION TEST

This tests looks at a sequence of one million 32-bit random integers. Each set of five consecutive integers can be in one of 120 states, for the $5!$ possible orderings of five numbers. Thus the 5th, 6th, 7th, numbers each provide a state. As many thousands of state transitions are observed, cumulative counts are made of the number of occurrences of each state.

- BINARY RANK TEST

This is the BINARY RANK TEST for 31x31 matrices. The leftmost 31 bits of 31 random integers from the test sequence are used to form a 31x31 binary matrix over the field $\{0,1\}$. The rank is determined. Ranks are found for 40,000 such random matrices and a chi-square test is performed on counts for ranks 31,30,29 and ≤ 28 .

- THE BITSTREAM TEST

The file under test is viewed as a stream of bits. Call them b_1, b_2, \dots . Consider an alphabet with two "letters", 0 and 1 and think of the stream of bits as a succession of 20-letter words, overlapping. Thus the first word is $b_1 b_2 \dots b_{20}$, the second is $b_2 b_3 \dots b_{21}$, and so on. The bit stream test counts the number of missing 20-letter (20-bit) words in a string of 2^{21} overlapping 20-letter words.

- OPSO(Overlapping-Pairs-Sparse-Occupancy) TEST

OPSO means Overlapping-Pairs-Sparse-Occupancy, The OPSO test considers 2-letter words from an alphabet of 1024 letters. Each letter is determined by a specified ten bits from a 32-bit integer in the sequence to be tested. OPSO generates 2^{21} (overlapping) 2-letter words, The OPSO test takes 32 bits at a time from the test file and uses a designated set of ten consecutive bits. It then restarts the file for the next designated 10 bits, and so on.

- OQSO(Overlapping-Quadruples-Sparse-Occupancy) TEST

The test OQSO is similar, except that it considers 4-letter words from an alphabet of 32 letters, each letter determined by a designated string of 5 consecutive bits from the test file, elements of which are assumed 32-bit random integers.

- COUNT-THE-1's TEST

It considers the file under test as a stream of bytes (four per 32 bit integer). Each byte can contain from 0 to 8 1's, with probabilities 1, 8,28,56,70,56,28,8,1 over 256. Now let the stream of bytes provide a string of overlapping 5-letter words, each "letter" taking values A,B,C,D,E.

- PARKING LOT TEST

In a square of side 100, randomly "park" a car---a circle of radius 1. Then try to park a 2nd, a 3rd, and so on, each time parking. That is, if an attempt to park a car causes a crash with one already parked, try again at a new random location.

- THE MINIMUM DISTANCE TEST

It does this 100 times, choose $n=8000$ random points in a square of side 10000. Find d , the minimum distance between the $(n^2-n)/2$ pairs of points. If the points are truly independent uniform, then d^2 , the square of the minimum distance should be (very close to) exponentially distributed with mean .995.

- THE 3DSPHERES TEST

Chooses 4000 random points in a cube of edge 1000. At each point, center a sphere large enough to reach the next closest point. Then the volume of the smallest such sphere is (very close to) exponentially distributed with mean $120\pi/3$.

- SQUEEZE TEST

Random integers are floated to get uniforms on $[0,1)$. Starting with $k=2^{31}=2147483647$, the test finds j , the number of iterations necessary to

reduce k to 1, using the reduction $k = \text{ceiling}(k \cdot U)$, with U provided by floating integers from the file being tested.

- OVERLAPPING SUMS TEST

Integers are floated to get a sequence $U(1), U(2), \dots$ of uniform $[0,1)$ variables. Then overlapping sums, $S(1) = U(1) + \dots + U(100)$, $S(2) = U(2) + \dots + U(101)$, ... are formed. The S 's are virtually normal with a certain covariance matrix. A linear transformation of the S 's converts them to a sequence of independent standard normals.

- RUNS TEST

This is the RUNS test. It counts runs up, and runs down, in a sequence of uniform $[0,1)$ variables, obtained by floating the 32-bit integers in the specified file. Runs are counted for sequences of length 10,000. This is done ten times.

- CRAPS TEST

It plays 200,000 games of craps, finds the number of wins and the number of throws necessary to end each game. The number of wins should be (very close to) a normal with mean $200000p$ and variance $200000p(1-p)$, with $p = 244/495$. Throws necessary to complete the game can vary from 1 to infinity.

DIEHARD requires a binary file of 32 bit integers to be generated; following is the code we used to generate the required binary file.

First the file is opened in binary append mode using the following statement

```
if ( (outfile = fopen("Rnd2DM.rng", "ab+" )) == NULL )
```


to write each character as 8 bits, where each bit in our case represents either 1 or 0 of a random number sequence we used the technique of character wise bit shifting operator, we declared 8 unsigned characters to hold each individual bit.

```
unsigned char mask8, mask7, mask6, mask5, mask4, mask3, mask2, mask1;
```

Each mask corresponds to bit 1 at a specific position,

```
e.g. mask8 = mask8<<7; // 1 0 0 0 0 0 0 0
```

Similarly we adjust each mask at the time of writing, then copy these values to a temp character variable by using bitwise OR operator using `temp = temp | mask8;` at the end this temp variable is written to the binary file using the following statement.

```
fwrite(&temp, sizeof(temp), 1, outfile);
```

the whole code is written in a iterative loop, so at the end we have generated a binary file in the form as required by DIEHARD.

2.3 Simulation Environment

The whole code for the proposed solutions is written in C++ and compiled and tested on both Microsoft Visual C++ and Borland Turbo C++ compilers, no compiler specific routine or function is called to maintain cross compilers compatibility, the written code can be easily ported to any OOP like C# or Java if any one wish to use those programming environments in future.

Chapter 3 Literature Review/ Related work

3.1 Literature Review

Random number generators are used in number of applications like computer simulations, built-in self test, and cryptography to name a few [1]. Most random number generators are deterministic in nature as the numbers are generated using some mathematical formulas or deterministic algorithms, known as Pseudo Random Number Generators or PRNG as opposed to a True random Number Generator where the source of a random number might be some natural source.

Standard Statistical tests are conducted to ensure a PRNG produces numbers that are uniformly distributed, uncorrelated with extreme long period [2]. There are quite a number of tests that can be applied to judge the quality of randomness, failing a single test does not mean that the sequence is not random so normally a set of tests is applied to check the quality of randomness, some times the tests to be applied might be application specific as well.

There are quite a number of PRNGs. Some examples are linear feedback shift register (LFSR), linear congruential generator (LCG), and cellular automata (CA) based PRNGs.

CA-based PRNGs were focus of researchers because its easy to implement in hardware, typically FPGA based Implementations, also computer based simulations in which implementation of cell structures and the corresponding CA rules in a corresponding data structure with each iteration generating the next state can be done easily.

The majority of research on CA-based PRNG has been focused on one-dimensional (1-D) CA with nearest three-cell neighborhood known as elementary CA, different types of cellular Automata and their applications are thoroughly discussed in a survey done on cellular automata [3].

The latest research focuses on increasing complexity of CA cell configurations and increasing CA dimensionality that can lead to better performance [4]. By increasing complexity we mean what logical operations are performed on a single cell per iteration and which neighbors are participating in finding the next state value, to make the

situation further implausible the dimensions of the lattice structure is increased thus making more options for the participating neighbors.

Certain modifications in basic CA are also suggested using extra control registers which transforms the basic CA to Programmable CA (PCA) [5]. These control registers works as decision makers in each iteration for number of participating neighbors, whether the current value of cell itself participates or not and similarly any kind of transformation in value post CA-logical operation is required or not. The control register values are themselves updated after each iteration.

One of the first 2D based CA was proposed by [6] based on 8 x 8 lattice of cells, while [7] worked on Asymmetric neighborhood and lattice structure PRNG for 2D CA.

For testing of Random Numbers the DIEHARD test is used by a number of researchers as it is treated as industry standard for checking the randomness quality of random numbers [8]. DIEHARD comprises of some 18 different statistical tests which are supposed to be passed if the resultant values are in a specific range, DIEAHRD requires a binary file of at least 10MB as an input.

The analysis of results produced by the DIEHARD is done by comparing p-values and finding the safe ranges when we compare them among several algorithms as done by [9]. Similarly others values can be calculated e.g. incomplete gamma count to check the uniformity of probability distribution among the generated p-values [2].

Sheng-Uei Guan etal. [10], describes the concept of self programmable automata (SPACA) and how to generate PRNG using this technique, authors used the concept on linear form of cells where the rule selection values are updated internally along with two external control registers.

The basic concept is to add rule selection neighborhood to the state transition neighborhood of PCA, thus making it SPCA. Authors used rules 90/150/165/105 for their rule selection neighborhood.

The paper discusses the concept for 1D CA structure and more emphasis is given on making FPGA implementation more efficient and simpler, moreover no results

comparison is given in results section using same approach; however authors considered different algorithms with different length blocks for analysis, also suggested out put sampling to increase efficiency of algorithm.

3.2 Related Work

A Two Dimensional Cellular Automata with programmable scheme is used by Byung-Heon Kang etal [11]. The authors applied rules 15, 31, 47 and 63 with two external control registers and compared the results with Guan, authors concluded that their scheme passes all DIEHARD tests with better quality as generated by Guan.

Created Rule Number	Complemented Control Signal	Self Control Signal
15 (001111)	0	0
31 (011111)	0	1
47 (101111)	1	0
63 (111111)	1	1

Figure 2: CA Rules Implemented for 2D-PCA [11]

The two control registers are self control register and sign control register, self control register's value decide whether the current value of cell participates in finding next state and sign control's value is used to logically invert the calculated results for next state's value.

The proposed algorithm is run 100 times and a pass is considered if the p-values from DIEHARD fall in the range $0.025 < p < 0.975$. Authors also used a novel time spacing technique for minimization of auto correlations among block of random numbers.

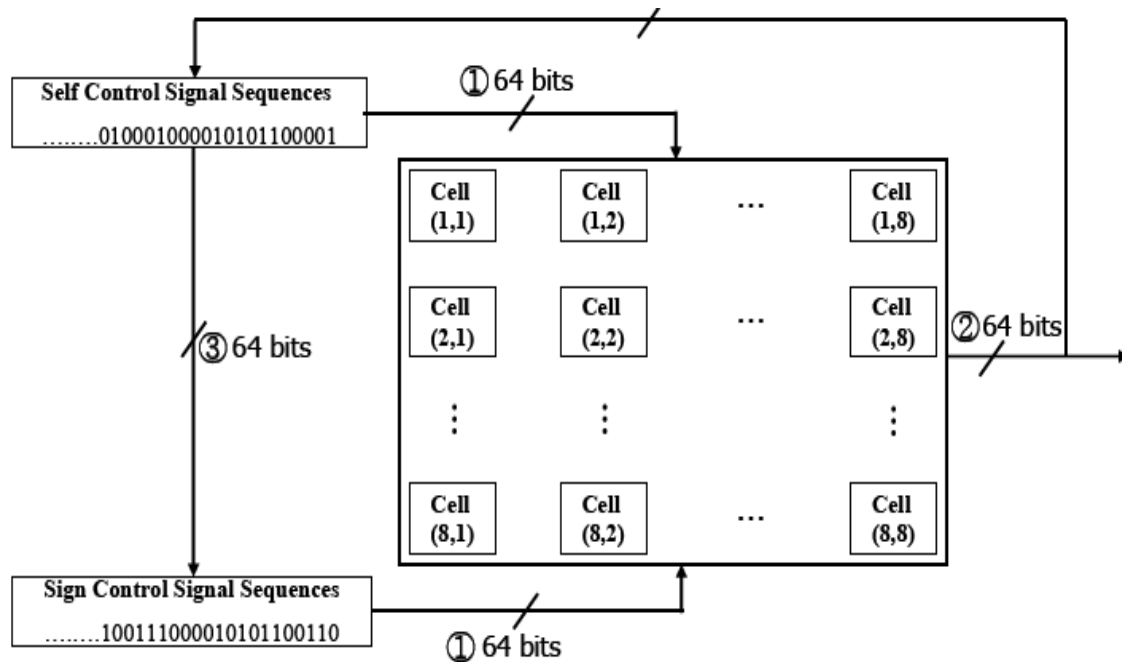


Figure 3: 2D-PCA Structure [11]

The research paper discusses only Von-Neumann neighbors for 2D CA implementation; strategy of how initial values are given to control registers is not discussed.

Sang-Ho Shin et al [12] worked on 3Dimmensional CA but with Moore Neighborhood only i.e. considering the diagonal neighbors, the focus of paper was to reduce the Hardware Implementation cost, the results were given in the form of Passing or Failing a DIEHARD test, what actual p-values are produced are not discussed so the quality of Randomness can not be discussed. No experimental setup or implementation methodology details are discussed, however authors claim that they have executed their scheme 100 times to generate the results.

Sang-Ho Shin, Kee-Young Yoo [13] worked on finding out on the optimal combination of Logical operations and CA rules for 3-Neighbors i.e. r+1 Von Neumann, the authors suggest a number of CA rules LCA, NCAAX, and NCAOX (as per their naming convention) for better results and also suggested that periodic boundary condition is better than the NULL boundary conditions i.e. for extreme cells the structure wraps around. The results discussed are just in the form of pass/ fail rate but the comparison is given between Null boundary and periodic boundary conditions.

Test Name	The results of tests (Pass or Fail)									
	Null boundary					Periodic boundary				
	L^*	N^*_{AX}	N^*_{OX}	N^*_{AO}	N^*_{AOX}	L^*	N^*_{AX}	N^*_{OX}	N^*_{AO}	N^*_{AOX}
Birthday Spacing	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass
Overlapping 5-Permutation	Pass	Pass	Pass	Fail	Pass	Pass	Pass	Pass	Fail	Pass
Binary Rank 31×31	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass
Binary Rank 32×32	Pass	Pass	Pass	Pass	Fail	Pass	Pass	Pass	Pass	Pass
Binary Rank 6×8	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass
Bitstream	Fail	Pass	Pass	Fail	Fail	Fail	Pass	Pass	Fail	Fail
OPSO	Fail	Fail	Pass	Pass	Pass	Fail	Fail	Pass	Pass	Pass
OQSO	Pass	Fail	Fail	Fail	Fail	Pass	Fail	Fail	Fail	Fail
DNA	Pass	Fail	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass
Count-The-1's 01	Pass	Pass	Pass	Pass	Fail	Pass	Pass	Pass	Pass	Fail
Count-The-1's 02	Pass	Fail	Pass	Fail	Pass	Pass	Pass	Pass	Fail	Pass
Parking Lot	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass
Minimum Distance	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass	Pass
3DS Spheres	Fail	Pass	Fail	Fail	Pass	Pass	Pass	Pass	Fail	Fail
Squeeze	Pass	Pass	Pass	Pass	Fail	Pass	Pass	Pass	Pass	Fail
Overlapping Sums	Pass	Pass	Pass	Pass	Fail	Pass	Pass	Pass	Pass	Pass
Runs	Pass	Pass	Pass	Fail	Pass	Pass	Pass	Pass	Fail	Pass
Craps	Pass	Pass	Pass	Fail	Fail	Pass	Pass	Pass	Pass	Fail
The number of pass	15	14	16	11	11	16	16	17	12	13

Figure 4: Results for Different Logical Operations with CA Rules [12]

The research paper is very good if some one wants to explore the results of different combinations of logical gates with different CA rules.

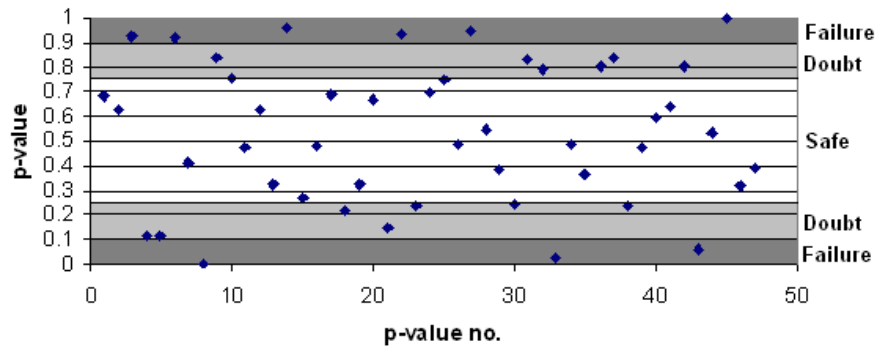


Figure 5: DIEHARD Results Compared using Safe Window Range [9]

Instead of comparing just Pass and Fail results as done by [12] and [13] a better approach is adopted by [9].

Here the author has defined three regions, Failure, Doubtful and Safe values providing a more better and quantitative analysis among the results being compared.

Chapter 4 Methodology

To complete this research work we followed the following sequence of events.

4.1 Literature Survey Phase

First of all a thorough Literature Survey is done to study current techniques applied in generation of Random Numbers specifically with respect to Cellular Automata, types of Random Numbers and their properties etc. this phase of work is already discussed in previous chapters.

4.2 Idea Generation Phase

Then we explored the areas where we can work and finalized the methods for Random Number Generation to be implemented. This idea generation phase lead to the proposed solutions, the methods to be implemented are given below; the details about each are discussed in chapter 5.

1. 2D Neumann Neighbors PCA
2. 2D Moore Neighbors PCA
3. 2D Neumann Moore jumping Neighbors PCA
4. 3D Neumann Neighbors PCA
5. 3D to 2D Jumping Neumann Neighbors PCA
6. 3D Moore + Neumann Neighbors PCA

4.3 Proposed Solution Implementation Phase

We used Computer Simulation to implement the required group of PCA for 3D CA and its variations and generated binary files of Random Numbers; the important piece of code is discussed in Appendix A.

The complete implementation is done and no compiler specific function is used nor any compiler settings has been made for the code to work, this makes the code portable with very little changes to any other object oriented programming language.

4.4 Testing Phase

In the next phase Testing is done on DIEHARD suit of Tests, DIEHARD is Internationally Accepted suit of tests for Randomness Testing and treated as de facto standard for testing RNGs, details of this phase are discussed in chapter 4 and detailed results are shown in Appendix B. 100 Binary files in the specific format as required by the DIEHARD are generated for each proposed solution and the DIEHARD dumps are gathered for the analysis phase.

4.5 Analysis of Results Phase

The Analysis and Comparison of Results comes in the last phase and it concludes which approach is better for Random number generation, this is discussed in chapter 6, analysis phase also describes the criteria we have selected for checking the quality of randomness of each proposed method.

4.6 Assumptions

We have made the following assumptions at the start of this research; these assumptions are further verified by the results produced.

- In general if you increase the complexity of neighbors the quality of Randomness for cellular automata will increase.
- By increasing dimensions the run time efficiency will not be increasing in exponential order.
- Although various other tests are present for randomness testing but DIEHARD is selected as it is considered as toughest to pass, plus DIEHARD suit contains a set of tests.
- The random numbers generated follows the normal distribution for a large data set; the large data set i.e. at least a 10MB file is also requirement of DIEHARD suit.
- The safe window size i.e. values that pass in certain range is taken from a journal paper published [9], the conclusive remarks might change if different safe window size is taken e.g. a broader window might pass those values that were considered to be failed and vice versa.

Chapter 5 Proposed Solutions

The analysis of literature survey gives us the following points:

1. Most of the work is done on 2D CA based structures, researchers tried different length/ sized structures and their symmetric and asymmetric variations.
2. Finding out which rules or logical gates work best, e.g. instead of using XOR, if we use AND or a combination of AND, OR gates etc, which rules of CA gives better results with which combination of gates.
3. Changing the rule selection mechanism, i.e. after each iteration a specific rule has to be selected from a given set of rules, this kind of PCA is also known as Self Programmable PCA or SPCA.
4. Different control registers added for different purposes e.g. introducing a Sign Control register that decides whether to take logical inverse of the result or the value should be simply saved, this type of operations increases the level of unpredictability and hence increases the level of randomness.
5. Hardware implementations of different CA proposed by the researchers, especially FPGA based implementations and their run time efficiencies.
6. We get only one paper on 3D PCA (only for Moore neighbors) and that also does not discuss results in detail nor are any implementation details given.

So we decided to explore 2D PCA and its variations, 3D PCA and its different variations like jumping between 3D PCA to 2D PCA with greater in-depth analysis of results.

5.1 Proposed Methods

We have proposed the following 3D PCA and variations of 2D PCA i.e. programmable cellular automata.

1. 2D Neumann Neighbors PCA
2. 2D Moore + Neumann Neighbors PCA
3. 2D Neumann Moore jumping Neighbors PCA

4. 3D Neumann Neighbors PCA
5. 3D to 2D Jumping Neumann Neighbors PCA
6. 3D Moore + Neumann Neighbors PCA

All the above ideas are never tried before by any of the researcher except the first one i.e. 2D Neumann Neighbors PCA, we re-implemented the 2D Neumann Neighbors PCA case to use it as a base case for comparison of results. It is to be noted that this study compares the effect of positional changes with respect to neighbors therefore we used the same algorithm for all the methods.

5.2 Algorithm used

We applied the following steps for each and every method.

1. Initialize All the Cell Values
2. Initialize Self Control register
3. Initialize Sign Control register
4. Pick Each Cell, select the Rule to be applied, Apply the Rule and Update Cell value
5. After Complete iteration Update the values of Registers
6. Extract 64bits Random Number (8 x 8 in case of 2D) and Go for Next iteration i.e. Step 4 for 64 * N bits of Random where N is the Number of iterations. For 3D it will be giving 8 x 64 i.e. 8 sets of 64 bit random numbers

0		1	1
		0	0
			1

For all the methods proposed we will apply the Non Null Boundary conditions i.e. the structure wraps around for extreme cells to pick the neighbors.

e.g. if the red boxed cell is the current cell and we want to apply rule 15 on it i.e. current cell will not participate and the neighbors to pick are left, top, right, bottom then the result will be, $\text{xor}(1,1,0,0) = 1$.

Now we will discuss each method one by one.

5.2.1 2D Neumann Neighbors

It is a two dimensional structure of 8 x 8 matrix, where we pick only Neumann neighbors i.e. immediate Non diagonal Neighbors of each cell.

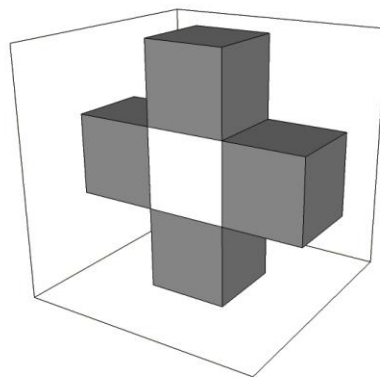


Figure 6: 2D Neumann Neighbors Illustration

In Fig. 6, we have the following participating neighbors,

White Cell: Current cell whose value is to be updated.

Gray Neighbors: Participating Neighbors (Non Diagonal Adjacent)
Left, Top, Right and Bottom.

The cell structure is shown in Fig. 7, there are two control registers

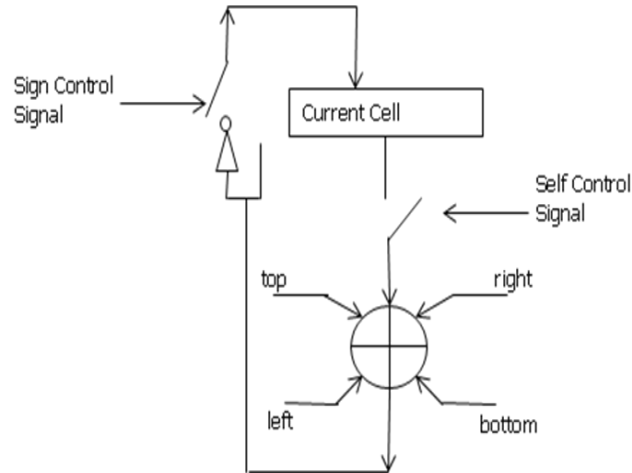


Figure 7: 2D Neumann Cell Structure

Sign Control: Sign control value decides whether we have to do a logical negation of result produced post XOR operation or should we store the value directly to the current cell.

Self Control: Self control value decides whether the current cells value participates in the XOR operation or only adjacent non diagonal neighbors have to participate.

Table 1. Shows the rules used during each iteration for calculating the cells next value.

Rule	CL	SI	Left	Top	Right	Bottom
15	0	0	1	1	1	1
31	0	1	1	1	1	1
47	1	0	1	1	1	1
63	1	1	1	1	1	1

Table1: 2D Neumann CA Rules

e.g. Applying Rule 15 means

$$(001111) \rightarrow (15)_{10} == (001111)_2$$

The first two zeros correspond to values of CL and SI register while the four 1's shows which neighbors are participating.

5.2.2 2D Moore + Neumann Neighbors

It is a two dimensional structure of 8 x 8 matrix, where we pick both the Neumann neighbors and Moore neighbors i.e. immediate Non diagonal Neighbors of each cell and diagonal neighbors as well, this increases the participating neighbors count and thus expected to give a better result in terms of randomness of generated number.

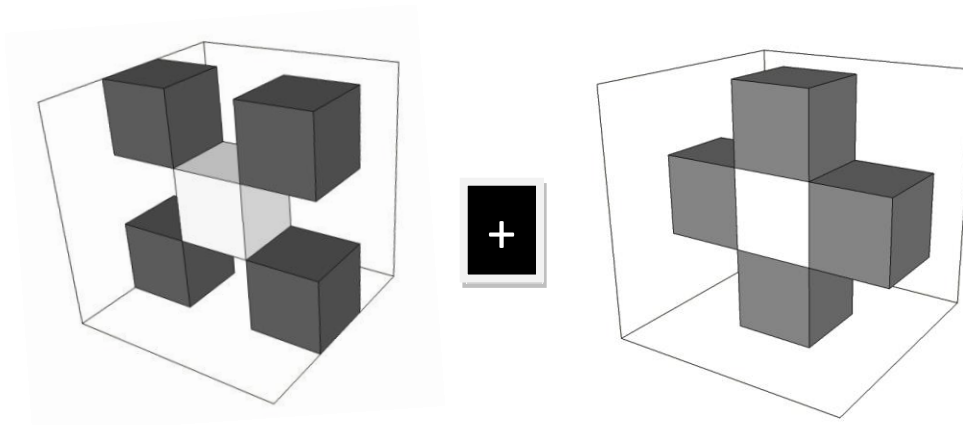


Figure 8: 2D Moore + Neumann Neighbors Illustration

In Fig. 8, We have the following participating neighbors,

- White Cell: Current cell whose value is to be up dated.
- Gray Neighbors (left): Participating Moore Neighbors (Diagonally Adjacent); Left-Top, Right-Bottom, Right-Top and Left-Bottom.
- Gray Neighbors (right): Participating Neumann Neighbors, Left, Top, Right, Bottom

The Cell structure is shown in Fig. 9

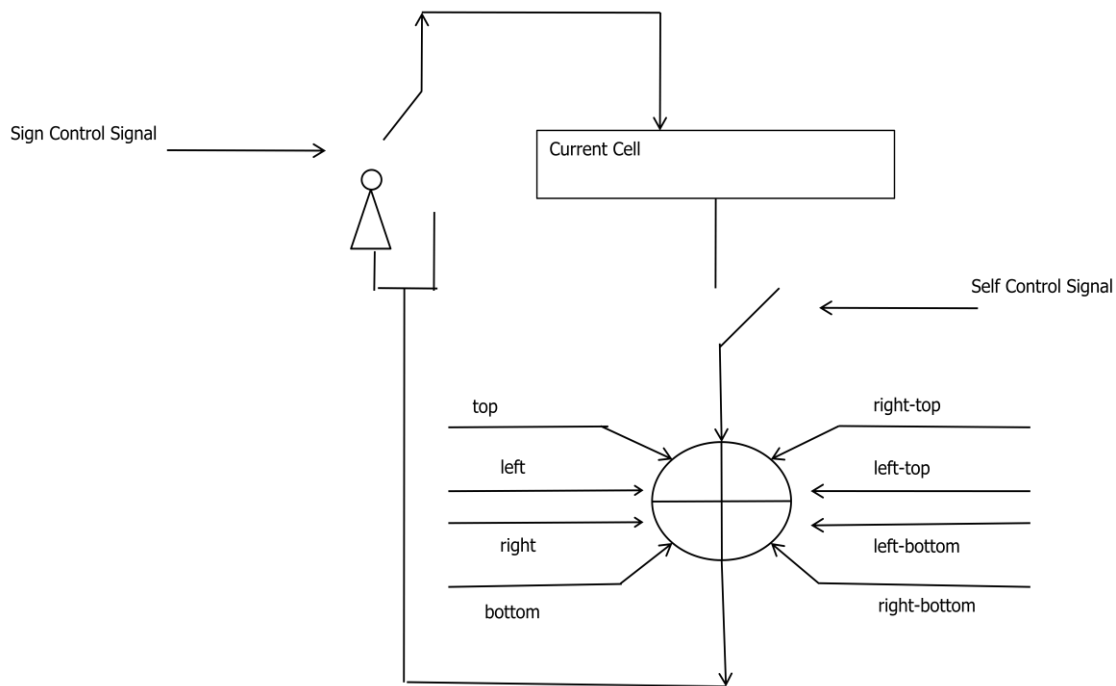


Figure 9: 2DMoore + Neumann Neighbors Cell Structure

There are again two control registers which play the same role as in 2D Neumann case

Rule	CL	SI	Left	Top	Right	Bottom	Left	Right	Left	Right
------	----	----	------	-----	-------	--------	------	-------	------	-------

							Top	Top	Bottom	Bottom
255	0	0	1	1	1	1	1	1	1	1
511	0	1	1	1	1	1	1	1	1	1
767	1	0	1	1	1	1	1	1	1	1
1023	1	1	1	1	1	1	1	1	1	1

e 2. Shows the iteration for value.

Table2: 2D Moore+ Neumann CA Rules

rules used during each calculating the cells next

5.2.3 2D Neumann Moore jumping Neighbors

Its again a two dimensional structure of 8 x 8 matrix, this time we pick between Neumann neighbors and Moore neighbors depending upon the newly introduced control register i.e. some times we pick the Moore neighbors and some times we pick the Neumann neighbors.

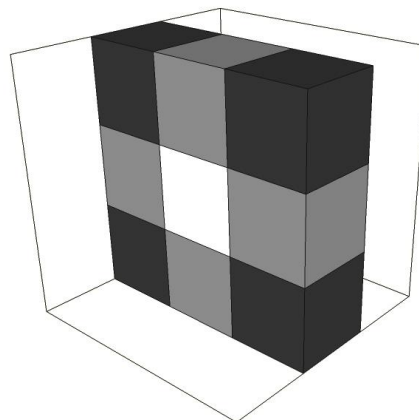


Figure 10: 2D Neumann Moore jumping Neighbors Illustration

In Fig. 10, we have the following participating neighbors,

White Cell:

Current cell whose value is to be up dated.

Light Gray Neighbors: Participating Neighbors (Non Diagonal Adjacent) Left, Top, Right and Bottom.

Black Neighbors: Participating Neighbors (Diagonally Adjacent)
Left-Top, Right-Bottom, Right-Top and Left-Bottom.

The Cell structure is shown in Fig. 11

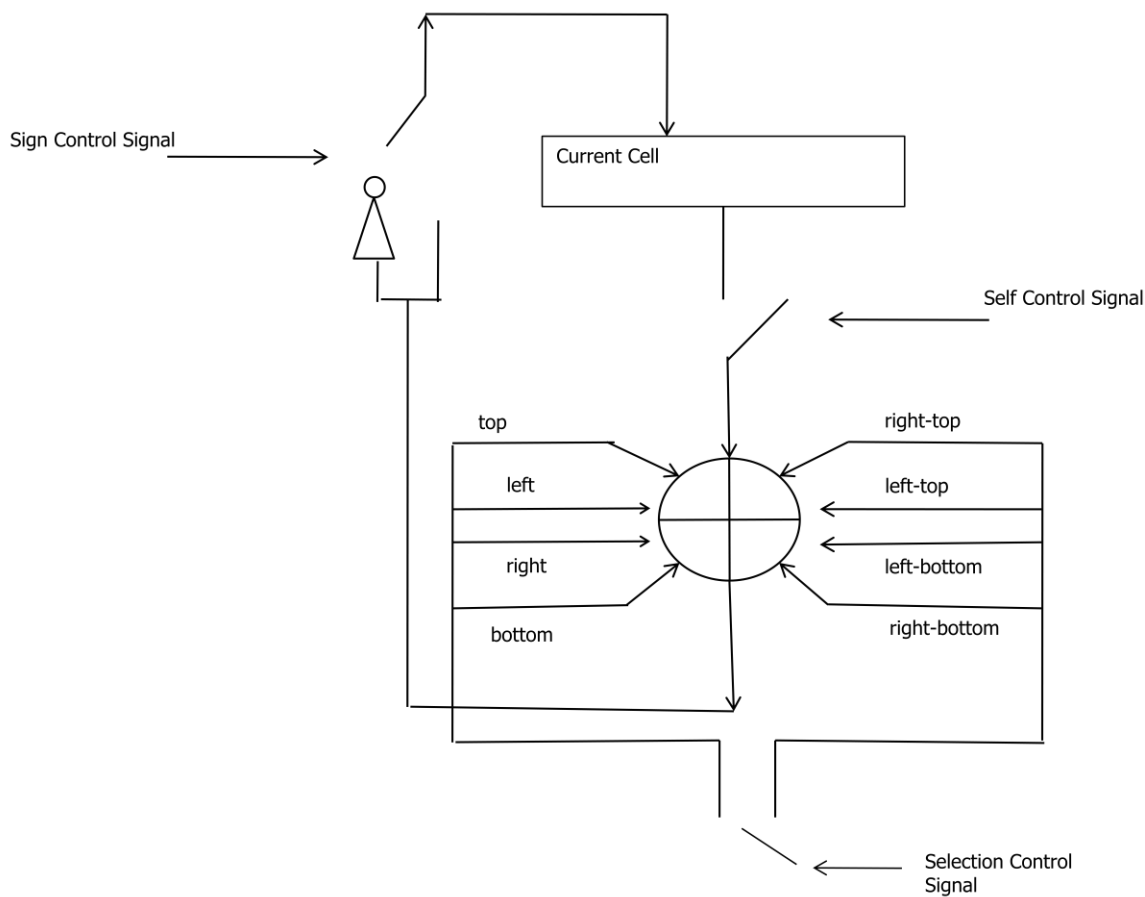


Figure 11: 2D Neumann Moore jumping Neighbors Cell Structure

This time we have three two control registers, the Sign and Self control registers play the same role as in 2D Neumann case while the third control register is used for deciding whether we opt for Diagonally Adjacent or Non Diagonally Adjacent neighbors.

Table 3. Shows the rules used during each iteration for calculating the cells next value.

Rule	CL	SI	DI	Left Top	Right Top	Left Bottom	Right Bottom	Left	Top	Right	Bottom
15	0	0	0	0	0	0	0	1	1	1	1
496	0	0	1	1	1	1	1	0	0	0	0
527	0	1	0	0	0	0	0	1	1	1	1
1008	0	1	1	1	1	1	1	0	0	0	0
1039	1	0	0	0	0	0	0	1	1	1	1
1520	1	0	1	1	1	1	1	0	0	0	0
1551	1	1	0	0	0	0	0	1	1	1	1
2032	1	1	1	1	1	1	1	0	0	0	0

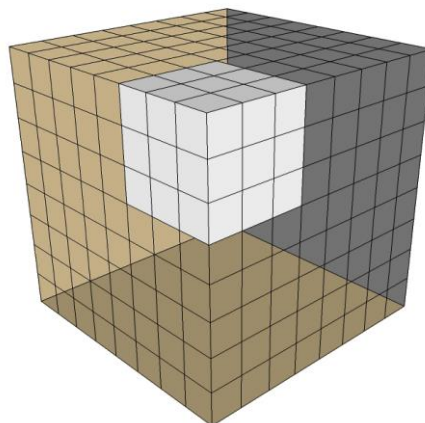
The DI is the register

newly added control

Table3: 2D Neumann Moore jumping CA Rules

5.2.4 3D

This time we
Dimensional
Random



Neumann Neighbors

are going to use a Three
Structure for generation of
numbers, it will be an array

of $8 \times 8 \times 8$ as shown in Fig. 12 consisting of 8 planes in which each plane has got 8 rows and 8 columns.

Figure 12: 3D Lattice of cells used for all the 3D PCA Algorithms

We will pick all the Neumann neighbors i.e. the six non-diagonally adjacent neighbors; four on the XY plane while two of them on Z axis. We will be using two control registers, one for Self Control and the other for Sign Control. Fig 13. Shows the diagram of participating neighbors while Fig. 14 shows the cell structure for the 3D Neumann case.

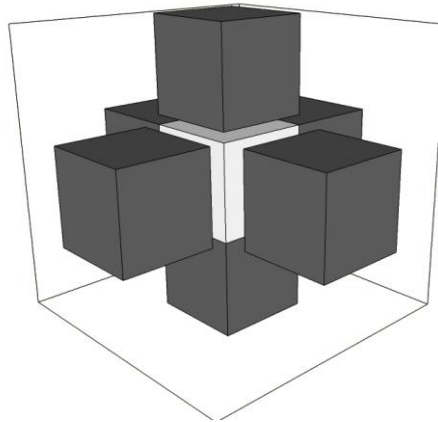


Figure 13: 3D Neumann Neighbors Illustration

In Fig. 13, we have the following participating neighbors,

White Cell: Current cell whose value is to be updated.

Gray Neighbors: Participating Neighbors (Non Diagonal Adjacent)

Left, Top, Right, Bottom, Facing Front on Z-axis, Facing Back on Z-axis.

The Cell structure is shown in Fig. 14

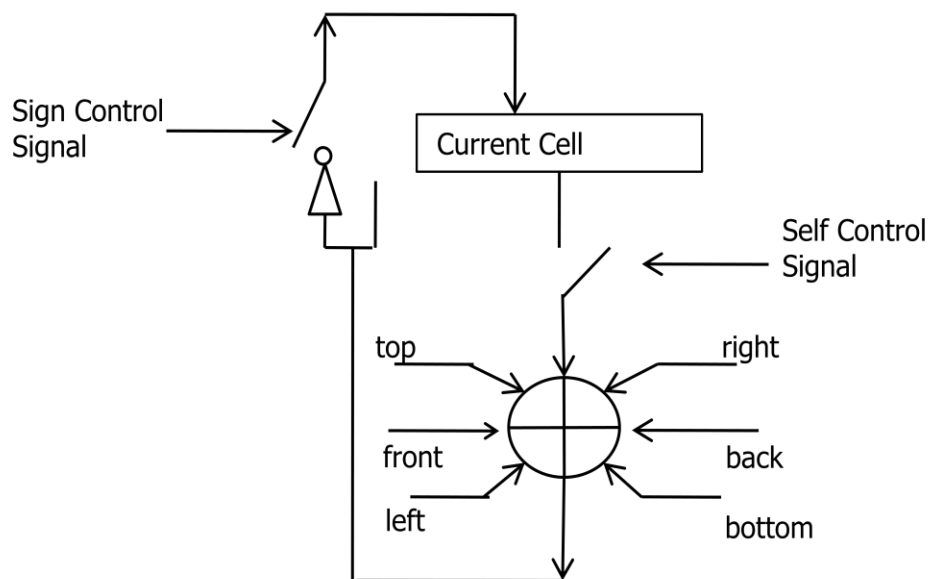


Figure 14: 3D Neumann Neighbors Cell Structure

Table 4. Shows the rules used during each iteration for calculating the cells next value.

Rule	CL	SI	Back	Front	Left	Top	Right	Bottom
63	0	0	1	1	1	1	1	1
127	0	1	1	1	1	1	1	1
191	1	0	1	1	1	1	1	1

255	1	1	1	1	1	1	1	1
-----	---	---	---	---	---	---	---	---

Table4: 3D Neumann CA Rules

5.2.5 3D to 2D Jumping Neumann Neighbors

3D to 2D Jumping Neumann Neighbors uses a 8 x 8 x 8 lattice of cells for Random numbers generation, this 3D cube structure picks all the 6 adjacent Neumann Neighbors just like the proposed solution number 4 does but with a subtle variation in which it jumps to the 2D Neumann structure i.e. picking only the adjacent non diagonal neighbors from XY plane only, this jumping is dependent upon the value a third control register.

Thus it could be considered as a mixture of solution two already proposed solutions, some times it behave like 2D Neumann CA and some times behave like a 3D Neumann CA, this kind of experimental jumping is done to check the quality of random numbers generated versus the efficiency of algorithm as approximately half of the times it is expected to be picking 2D Neumann neighbors (more efficient) and half of the time generating a random number while considering the z-plane neighbors (more randomness).

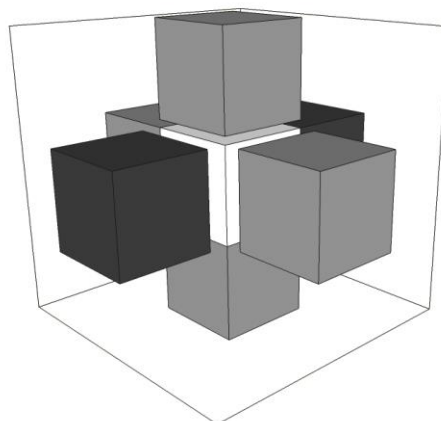


Figure 15: 3D-2D jumping Neumann Neighbors Illustration

Participating Neighbors are as follows,

- | | |
|-----------------------|--|
| White Cell: | Current cell whose value is to be updated. |
| Light Gray Neighbors: | Participating Neighbors (Non Diagonal Adjacent),
Left, Top, Right and Bottom. |
| Black Neighbors: | Participating Neighbors (Non Diagonal Adjacent),
Facing Front on Z-axis, Facing Back on Z-axis + Light
Gray Neighbors. |

Light Gray Neighbors are picked in case the current iteration is calculating for 2D Neumann while Black Neighbors along with Light Gray Neighbors are used in case 3D Neumann is selected for the current iteration; the cell structure is shown in Fig. 16.

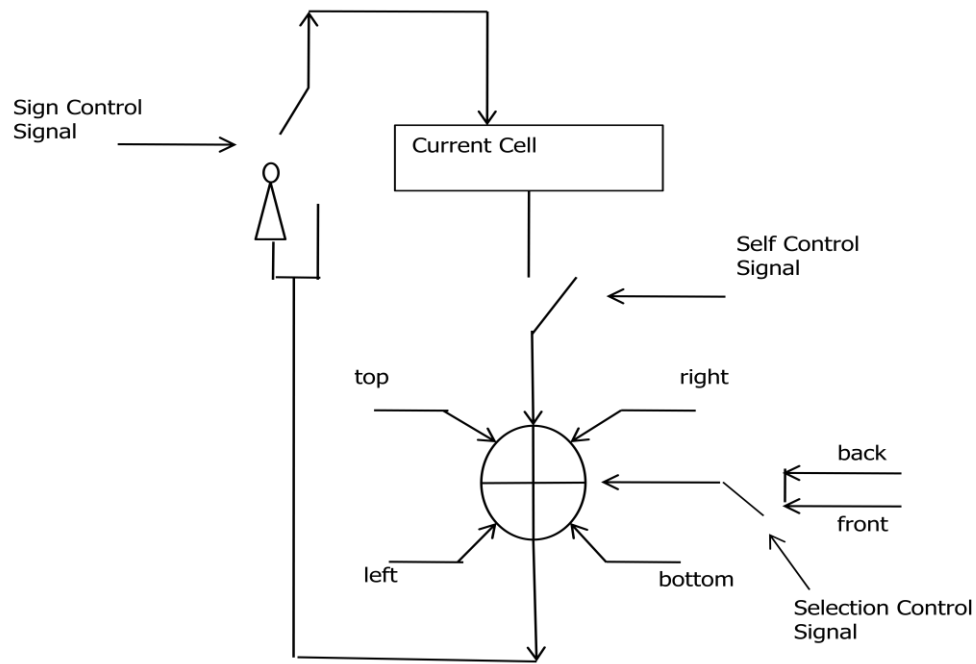


Figure 16: 3D to 2D Jumping Neumann Neighbors Cell Structure

Table 5. Shows the rules used in each iteration.

Rule	CL	SI	DI	Back	Front	Left	Top	Right	Bottom
15	0	0	0	0	0	1	1	1	1
127	0	0	1	1	1	1	1	1	1
143	0	1	0	0	0	1	1	1	1
255	0	1	1	1	1	1	1	1	1
271	1	0	0	0	0	1	1	1	1
383	1	0	1	1	1	1	1	1	1
399	1	1	0	0	0	1	1	1	1
511	1	1	1	1	1	1	1	1	1

Table5: 3D to 2D jumping Neumann CA Rules

The third control register DI is used for jumping between 3D to 2D Neumann, the SI and DI plays the same role as in simple 3D Neumann solution.

5.2.6 3D Moore + Neumann Neighbors

In this last proposed scheme we opted to select the all possible neighbors of a 3D cell, i.e. all the Diagonally Adjacent Moore neighbors along with all the Non Diagonally Adjacent Neumann neighbors, In Fig 17. We show the graphical illustration for such a configuration.

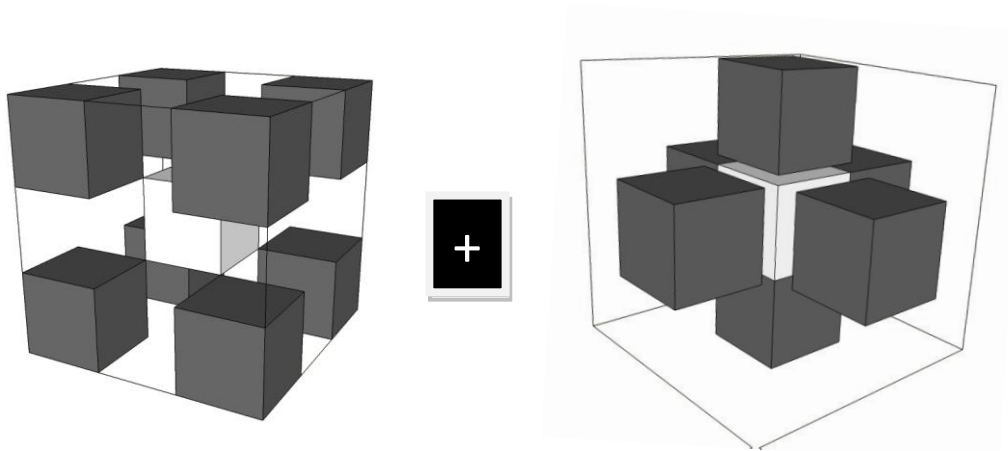


Figure 17: 3D Moore + Neumann Neighbors Illustration

Participating Neighbors are

White Cell: Current cell whose value is to be updated.

Gray Neighbors (left): Participating Neighbors (Diagonally Adjacent)

Front Left-Top and Back Right-Bottom, Front Right-Top and Back Left-Bottom, Back Left-Top and Front Right-Bottom, Back Right-Top and Front Left-Bottom.

Gray Neighbors (right): Non Diagonally adjacent Neumann

There are just two control registers used in this scheme the SI and CI i.e. self control and complement control registers with the same roles as they have played in the schemes discussed before.

The Fig. 18 shows the cell structure for 3D Moore + Neumann.

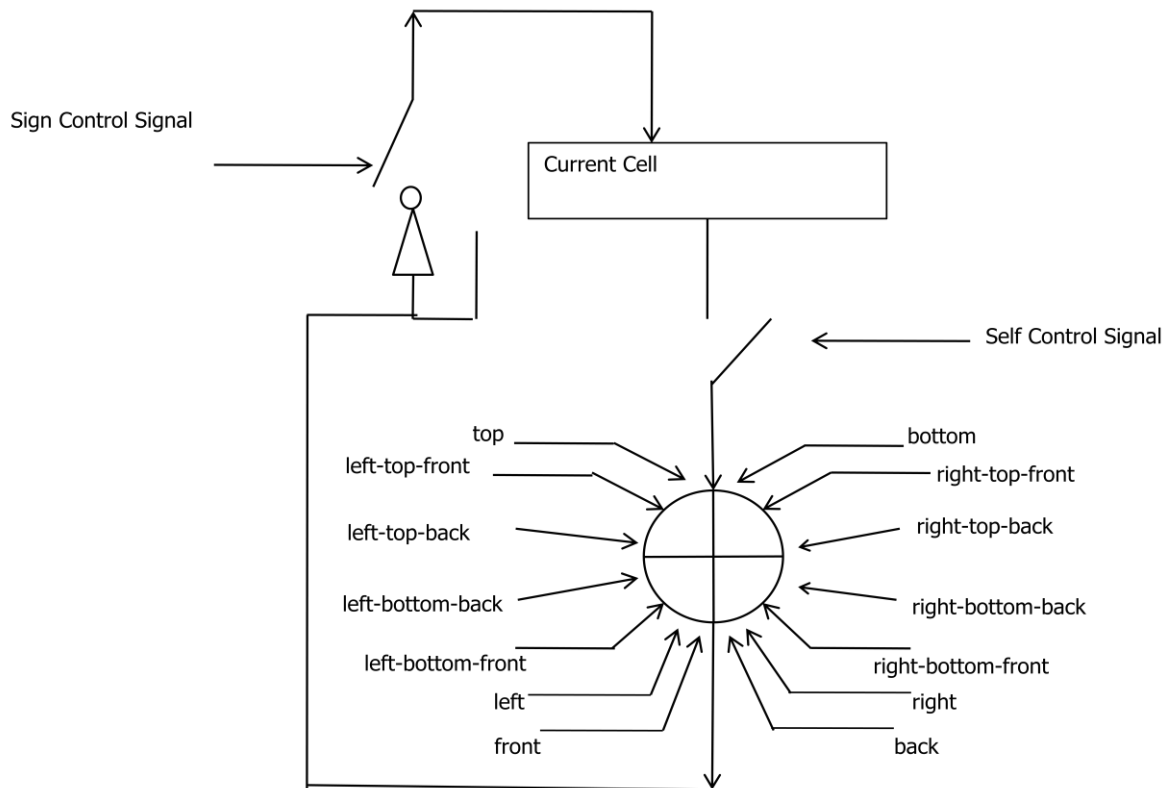


Figure 18: 3D Moore + Neumann Neighbor Cell Structure

Table 6. shows the rules used in each iteration

Rule	CL	SI	Left-Top Front	Left-Bottom Front	Right-Top Front	Right-Bottom Front	Left-Top Back	Left-Bottom Back	Right-Top Back	Right-Bottom Back
16383	0	0	1	1	1	1	1	1	1	1
32767	0	1	1	1	1	1	1	1	1	1
49151	1	0	1	1	1	1	1	1	1	1
65535	1	1	1	1	1	1	1	1	1	1
Rule	CL	SI	Back	Front	Left	Top	Right	Bottom		
16383	0	0	1	1	1	1	1	1		
32767	0	1	1	1	1	1	1	1		
49151	1	0	1	1	1	1	1	1		
65535	1	1	1	1	1	1	1	1		

Table6: 3D Moore + Neumann CA Rules

Chapter 6 Testing of Proposed Solutions

The Testing is done on DIEHARD suit of Tests; DIEHARD is Internationally Accepted for Randomness Testing and treated as de facto standard for testing RNGs.

As said DIEHARD is suit of tests, it basically comprises of 18 rigorous statistical tests, each test checks the randomness of a series of numbers on a specific parameter defined.

6.1 DIEHARD Tests Applied

We applied the following tests of DIEHARD on all the proposed solutions.

- BIRTHDAY SPACINGS TEST
- THE OVERLAPPING 5-PERMUTATION TEST
- BINARY RANK TEST
- THE BITSTREAM TEST
- OPSO(Overlapping-Pairs-Sparse-Occupancy)
- OQSO(Overlapping-Quadruples-Sparse-Occupancy)
- DNA
- COUNT-THE-1's TEST
- PARKING LOT TEST
- THE MINIMUM DISTANCE TEST
- THE 3DSPHERES TEST
- SQUEEZE test
- OVERLAPPING SUMS test
- RUNS test
- CRAPS TEST

The internal workings of these tests are available with the DIEHARD documentation and we have briefly discussed them in chapter 2 as well, here we use all the tests as Black-Box testing approach, we will be feeding the random numbers to these tests and then analyze the results obtained.

DIEHARD requires a binary file of 32 bit integers of at least 10MB size i.e. 80 million bits and gives you results in a text based output file.

Most of the tests in DIEHARD return a p-value, which is uniform on $[0,1)$. This means that to qualify for attest the p-value should fall between 0 and 1 otherwise the test is supposed to be failed.

These p-values are obtained by $p=F(X)$, where F is the assumed distribution of the sample random variable X ---often normal. But that assumed F is just an asymptotic approximation, for which the fit will be worst in the tails [8].

6.2 Testing Process

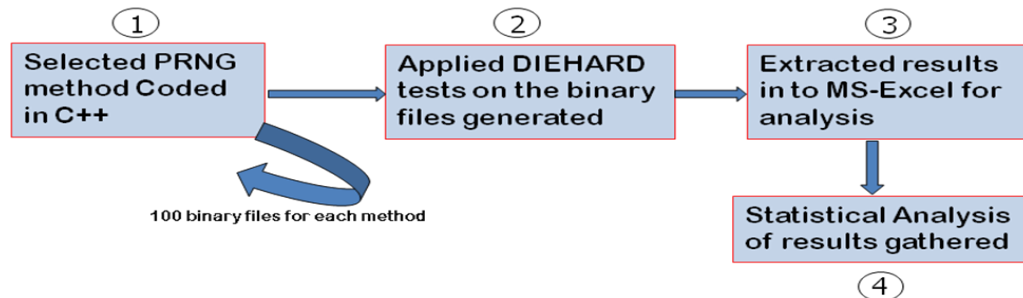


Figure 19: Testing Phases block diagram

As Fig. 19 shows we generated 100 binary files for each proposed method we discussed and then applied all the DIEHARD tests on these binary files, in the next phase we extracted results from the DIEHARD produced text files and gathered all the values for analysis.

Chapter 7 Analysis of Results

First of all we gathered 100 set of values for each method in Excel Sheet,

	A	B	C	D	E	F	G	H
1			3D CA Von Neuman Neighbours					
2		DIEHARD Test Applied						
3			Test1	Test2	Test3	Test4	Test5	Test6
4	BST	BIRTHDAY SPACINGS TEST	0.285389	0.632464	0.164649	0.154146	0.748149	0.3647
5	O5PT	THE OVERLAPPING 5-PERMUTATION TEST	0.185904	0.223082	0.231641	0.220116	0.291651	0.3130
6	BR31	BINARY RANK TEST for 31x31 matrices	0.340441	0.826918	0.723798	0.514656	0.339327	0.4148
7	BR32	BINARY RANK TEST for 32x32 matrices	0.393533	0.449978	0.758735	0.35713	0.489235	0.7621
8	BR6X8	BINARY RANK TEST for 6x8 matrices	0.627477	0.412211	0.704265	0.470615	0.695188	0.7648
9	BS	THE BITSTREAM TEST	0.4842455	0.4546095	0.453482	0.4583665	0.46576	0.4323
10	OPSO	OPSO(Overlapping-Pairs-Sparse-Occupancy)	0.47850808	0.475734783	0.470895652	0.473178261	0.459373913	0.4602
11	OQSO	OQSO(Overlapping-Quadruples-Sparse-Occupancy)	0.548003571	0.548521429	0.549582143	0.551342857	0.548053571	0.5487
12	DNA	DNA	0.485935484	0.484358065	0.485990323	0.482416129	0.479412903	0.47
13	CT1TSB-1	COUNT-THE-1's TEST on a stream of bytes (1st Set of Bytes)	0.407783	0.415675	0.428655	0.418253	0.416843	0.388
14	CT1TSB-2	COUNT-THE-1's TEST on a stream of bytes (2nd Set of Bytes)	0.076273	0.071945	0.07979	0.07825	0.095681	0.0892
15	CT1TSB	COUNT-THE-1's TEST for specific bytes	0.62028036	0.62636	0.62344152	0.62550312	0.61885336	0.6214
16	PLT	PARKING LOT TEST	0.219448	0.494477	0.447641	0.303457	0.509747	0.381
17	MDT	THE MINIMUM DISTANCE TEST	0.746275	0.636845	0.827955	0.864977	0.156177	0.1183
18	3DST	THE 3DSPHERES TEST	0.133558	0.135028	0.059775	0.354523	0.272049	0.926
19	ST	SQUEEZE test	0.542461	0.528646	0.535832	0.53534	0.532835	0.492
20	OST	OVERLAPPING SUMS test	0.05403	0.732507	0.054523	0.32787	0.127621	0.3986
21	RD	runs down	0.354139	0.275062	0.081794	0.395163	0.684918	0.7206
22	RU	runs up	0.948777	0.440351	0.778015	0.858423	0.616426	0.8378
23	CW	Craps Wins	0.087068	0.081536	0.728705	0.737526	0.744763	0.0665
24	CTG	Craps Throws/Game	0.88432	0.884638	0.580053	0.555906	0.537791	0.8317

Then we

	CU	CV	CW	CX	CY	CZ	DA	DB	DC	DD	DE	DF	DG	DH	DI
3	Test97	Test98	Test99	Test100	STDEV	AVG	AVG +/- (3*STDEV)				AVG +/- (2*STDEV)				
4	0.833222	0.266663	0.124598	0.447448	0.310241	0.494981	-0.43574	1.425705	100	Pass	-0.1255	1.115464	100	Pass	
5	0.912361	0.545414	0.203953	0.458685	0.244164	0.260486	-0.47201	0.992979	100	Pass	-0.22784	0.748815	92	Fail	
6	0.502976	0.509437	0.443056	0.632394	0.19722	0.547438	-0.04422	1.139097	100	Pass	0.152999	0.941878	92	Fail	
7	0.877634	0.320864	0.931846	0.413575	0.197617	0.546782	-0.04607	1.139633	100	Pass	0.151549	0.942016	100	Pass	
8	0.410174	0.218027	0.091691	0.779667	0.237226	0.585538	-0.12614	1.297216	100	Pass	0.111086	1.05999	96	Pass	
9	0.599397	0.498096	0.560697	0.553334	0.051308	0.507176	0.353253	0.661099	100	Pass	0.404561	0.609791	96	Pass	
10	0.370241	0.415093	0.385806	0.383932	0.071854	0.470767	0.255206	0.686328	100	Pass	0.327059	0.614474	92	Fail	
11	0.468009	0.536384	0.597663	0.560891	0.036482	0.537078	0.427633	0.646523	100	Pass	0.464114	0.610041	96	Pass	
12	0.577428	0.550089	0.488647	0.426447	0.044072	0.495401	0.363184	0.627618	100	Pass	0.407256	0.583545	100	Pass	
13	0.519537	0.18903	0.728313	0.180777	0.223879	0.359494	-0.31214	1.031113	100	Pass	-0.08826	0.807252	96	Pass	
14	0.337915	0.217746	0.057929	0.060313	0.145257	0.172064	-0.26371	0.607835	96	Fail	-0.11845	0.462578	92	Fail	
15	0.45604	0.556022	0.504329	0.619427	0.061709	0.57443	0.389303	0.759558	100	Pass	0.451012	0.697849	96	Pass	
16	0.821525	0.923384	0.537322	0.096546	0.290079	0.440724	-0.42951	1.310961	100	Pass	-0.13943	1.020882	100	Pass	
17	0.608043	0.360999	0.619713	0.3802	0.273015	0.460619	-0.35842	1.279663	100	Pass	-0.08541	1.006648	100	Pass	
18	0.877347	0.175447	0.525008	0.641199	0.286813	0.483589	-0.37685	1.344028	100	Pass	-0.09004	1.057215	100	Pass	
19	0.545387	0.513367	0.569482	0.532706	0.156713	0.590107	0.119969	1.060244	100	Pass	0.276682	0.903532	88	Fail	
20	0.015373	0.652866	0.715041	0.208704	0.283541	0.42757	-0.42305	1.278192	100	Pass	-0.13951	0.994651	100	Pass	
21	0.194317	0.893248	0.301783	0.337618	0.293729	0.432926	-0.44826	1.314112	100	Pass	-0.15453	1.020384	100	Pass	
22	0.184858	0.986531	0.578532	0.067204	0.288356	0.552818	-0.31225	1.417887	100	Pass	-0.02389	1.129531	100	Pass	
23	0.915929	0.562446	0.821733	0.262886	0.264547	0.519815	-0.27383	1.313457	100	Pass	-0.00928	1.048909	100	Pass	
24	0.621135	0.814019	0.685994	0.546909	0.24451	0.666823	-0.06671	1.400353	100	Pass	0.177803	1.155843	96	Pass	

calculated $\mu \pm 2\sigma$ and $\mu \pm 3\sigma$ and also calculated number

Figure 20: Snap shot of results

for each method, of values falling

in Safe Range for each Test (values ≥ 0.25 AND values ≤ 0.75).

7.1 Criteria 1

99.6% values should fall in $\mu \pm 3\sigma$.

95.4% values should fall in $\mu \pm 2\sigma$.

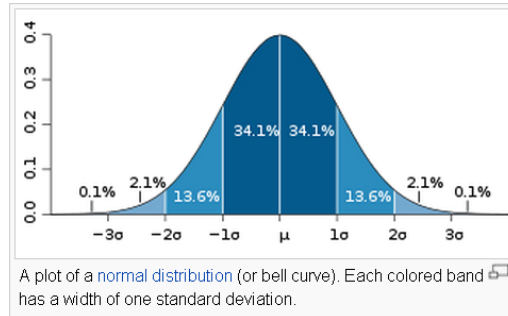
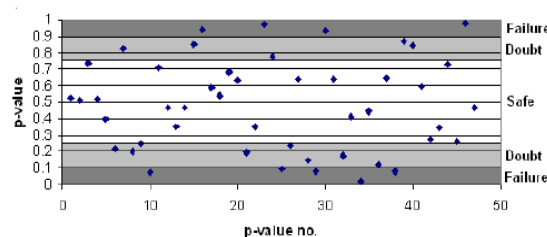


Figure 21: Courtesy: en.wikipedia.org/wiki/Standard_deviation

This criteria is applied as the p-values generated by the DIEHARD approximates to the Normal distribution[2], and it is well known fact that for a normal distribution 99.6% values should fall in $\mu \pm 3\sigma$ and 95.4% values fall in $\mu \pm 2\sigma$.

7.2 Criteria 2

Check number of values falling in Safe Range for each Test (values ≥ 0.25 AND values ≤ 0.75), i.e. out of 100 values per DIEHARD test it shows the count of how many values fall in this safe range.



Applying

criterion takes the safest values among the p-values produced by DIEHARD test.

We obtained the following out come when we applied $\mu \pm 2\sigma$ on all the tests for each method.

No of Passes Avg + - 2 Std. Dev 95% Values Fail						
Jumping 2D-3D	Rand3DVonNeuman	3D Moore + Neuman	2D Moore	2D Von Neuman Moore Jumping	2D Von Neuman	
Pass	Pass	Pass	Pass	Pass	Pass	Pass
Fail	Fail	Pass	Pass	Pass	Pass	Pass
Pass	Fail	Pass	Pass	Pass	Pass	Pass
Fail	Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass	Pass
Fail	Pass	Pass	Pass	Pass	Pass	Pass
Fail	Fail	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass	Pass
Pass	Fail	Pass	Pass	Pass	Pass	Fail
Pass	Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Fail	Pass	Pass	Pass	Pass
Pass	Fail	Pass	Pass	Pass	Pass	Fail
Pass	Pass	Pass	Fail	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Fail	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass	Pass
Failures	4	5	1	2	0	2

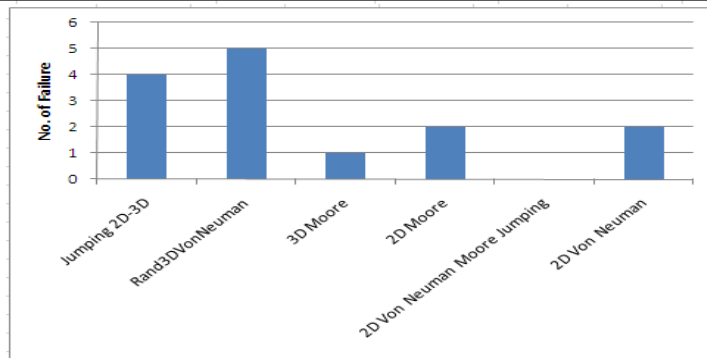


Figure 24: Results and graph showing values in the range $\mu \pm 2\sigma$

This time we get two favorable methods, 2D Neumann – Moore jumping has got zero failures while 3D Moore + Neumann has just one failure, all the other methods are more failures with 3D Neumann has the maximum failures.

Then we applied criteria 2 to check among all the 100 values per method for each test falling in the safe range i.e. (values ≥ 0.25 AND values ≤ 0.75), this type of analysis considers all the p-values instead of using average and standard deviation of all the values.

Fig 25. Discusses the results obtained using this approach.

All 100 values Comparison (values >=0.25 AND values <=0.75) Safe Range								
	1	2	3	4	5	6		
	Jumping 2D-3D	Rand 3D Von Neuman	3D Moore + Neuman	2D Moore Neighbours	2D Von Neuman Moore Jump N	2D Von Neuman Neighbors		Max Value Series
BST	68	40	56	52	44	56		1
O5PT	28	24	36	36	36	28		3, 4, 5
BR31	80	84	84	72	68	0		2, 3
BR32	84	76	72	76	80	0		1
BR6X8	60	60	44	56	52	36		1, 2
BS	100	100	100	100	100	8		1, 2, 3, 4, 5
OPSO	100	100	100	76	80	100		1, 2, 3, 6
OQSO	100	100	100	76	80	100		1, 2, 3, 6
DNA	100	100	100	76	80	100		1, 2, 3, 6
CT1TSB-1	40	52	48	40	80	28		5
CT1TSB-2	48	20	72	60	68	40		3
CT1TSB	100	100	100	100	100	100		1, 2, 3, 4, 5, 6
PLT	48	40	40	52	60	64		6
MDT	48	64	52	44	28	32		2
3DST	36	48	48	72	52	48		4
ST	44	84	48	40	44	0		2
OST	56	56	52	48	52	60		6
RD	64	48	56	56	56	60		1
RU	40	48	52	44	40	52		3, 6
CW	36	64	56	24	52	64		2, 6
CTG	60	48	44	64	52	56		4

	Jumping 2D-3D	Rand 3D Von Neuman	3D Moore + Neuman	2D Moore Neighbours	2D Von Neuman Moore Jump N	2D Von Neuman Neighbors
Max Occurance	9	10	9	5	4	8

Figure 25: Values falling in the Safe Range (values >=0.25 AND values <=0.75)

The right most columns shows which method qualifies maximum number of times in each test e.g. 2D-3D jumping Neumann qualifies with the 68 tests, so the right most column represents this information, similarly we calculated for all the methods and all the tests. At the bottom the summary is shown, 3D Von-Neumann appears 10 times in the rightmost column while 3D Moore + Neumann appears 9 times, so these two methods are the candidates from the above figure.

7.3 Comparative Analysis of Results

So we have the following results as best performing candidates.

Criteria 1: (Tests passing in $\mu \pm 2\sigma$ range)

- 2D Neumann – Moore jumping
- 3D Moore + Neumann

Criteria 2: (Values falling in the Safe Range, values ≥ 0.25 AND values ≤ 0.75)

- 3D Neumann
- 3D Moore + Neumann

2D Moore-Neumann jumping does not perform well in Criteria 2 as it occurs only at 4 places, similarly 3D Neumann which is a candidate in Criteria 2 does not perform well in Criteria 1 by failing at 5 places, so we can safely conclude that 3D Moore + Neumann is the best performing method among all the methods we discussed.

7.4 Run Time Analysis of the Proposed Methods

The criteria 1 and criteria 2 discussed in sections 5.1 and 5.2 solely discuss the issue with respect to the quality of random numbers generated as we compared the DIEHARD produced values and applied certain statistical operations on the results, the discussion remains incomplete if we do not compare the run time efficiency of methods among each other.

The basic point is to measure the cost we have to pay in increasing the quality of randomness of a sequence i.e. how much we have to pay in terms of number of operations to get a better random numbers sequence.

To make the things simpler we can divide the methods in to three categories,

1. Working on pure 2D structures
 - 2D Neumann Neighbors
 - 2D Moore + Neumann Neighbors
 - 2D Neumann Moore jumping Neighbors

2. Working on pure 3D structures
 - 3D Neumann Neighbors
 - 3D Moore + Neumann Neighbors

3. Working on 3D to 2D transition structures
 - 3D to 2D Jumping Neumann Neighbors

7.4.1 Main Memory Requirements

Memory Requirement for pure 2D structures is common for all as we use the same data structure for all 2D cases; the difference is just in picking neighbors.

One integer takes 4 bytes, so a matrix of 8 x 8 i.e. 64 integers, would take $64 \times 4 = 256$ bytes of main memory.

For pure 3D cases and transition cases we used the grid of 8 x 8 x 8 i.e. 512 integers, would take 2048 bytes of main memory.

The memory requirements will remain constant during complete execution as after each iteration the generated numbers are flushed to the binary file.

7.4.2 Number of operations per iteration

For 2D cases we have to go through each cell thus requiring a two level nested loops, outer for rows and inner for columns, plus we need to pick 4 neighbors in case of Neumann, and 8 neighbors in case of combination of Neumann and Moore, similarly Neumann Moore jumping also requires just 4 neighbors to be picked (either Neumann's or Moore's), the cost of XOR operation and writing to binary file remains same for all the cases, thus for a 2D Neumann PCA with Self Cell participating we can calculate the running time as follows,

Statement	Cost	Time
for (int rows = 1; rows <=8; rows++)	C1	m
for (int cols = 1; cols<=8; cols++)	C2	n
Pick all the five values, the white cell + all Gray neighbors	C3	5mn
Find XOR (five values)	C4	5mn
Update the White Cell	C5	mn

The temporary variables used to store intermediate result represent the intermediate cost and that will be constant for all the cases with no significant increase in the run time of algorithm.

Exact Running Time/ Time Complexity = $c_1m + c_2n + 5c_3mn + 5c_4mn + c_5mn$

Order of Growth= $O(mn)$

As we have worked on 8 x 8 grid we have, $m = n$ therefore Order of Growth is $O(n^2)$, this will be the case for all the 2D cases as the maximum order of growth will not go beyond $O(n^2)$.

For 3D structures we have to apply 3 nested loops, first for the plane, second for the rows and third for the columns to iterate through each cell, the Neumann neighbors of each 3D cell be 6 while for Neumann + Moore the neighbors are 10 in number.

Thus for 3D Neumann we can calculate the cost as follows

Statement	Cost	Time
for (int planes = 1; planes <=8; planes++)	C0	p
for (int rows = 1; rows <=8; rows++)	C1	m
for (int cols = 1; cols<=8; cols++)	C2	n
Pick all the seven values, the white cell + all Gray neighbors	C3	7pmn
Find XOR (seven values)	C4	7pmn
Update the White Cell	C5	pmn

Exact Running Time = $c_0p + c_1m + c_2n + 7c_3pmn + 7c_4pmn + c_5pmn$

Order of Growth= $O(pmn)$

As we have worked on 8 x 8 x 8 grid we have, $p = m = n$ therefore Order of Growth is $O(n^3)$, similarly all the 3D cases will have the same order of growth as maximum order of growth will not jump beyond $O(n^3)$.

In case of 3D to 2D Jumping Neumann Neighbors, it is expected that 50% of the time it behaves like 3D and 50% of time behaves like 2D, thus the running cost will be calculated by $0.5 \times \text{Cost 2D Neumann} + 0.5 \times \text{Cost 3D Neumann}$.

Table 7 summarizes the run time calculations and order of growth for all the proposed methods.

Method	Run time = No. of Operations per iteration x Cost	Order of Growth
2D Neumann	$c_1m + c_2n + 5c_3mn + 5c_4mn + c_5mn$	$O(N^2)$
2D Neumann Moore jumping	$c_1m + c_2n + 5c_3mn + 5c_4mn + c_5mn$	$O(N^2)$
2D Moore + Neumann	$c_1m + c_2n + 9c_3mn + 9c_4mn + c_5mn$	$O(N^2)$
3D Neumann	$c_0p + c_1m + c_2n + 7c_3pmn + 7c_4pmn + c_5pmn$	$O(N^3)$
3D Moore+ Neuman	$c_0p + c_1m + c_2n + 15c_3pmn + 15c_4pmn + c_5pmn$	$O(N^3)$
3D to 2D jumping Neuman	$0.5*(c_0p + c_1m + c_2n + 7c_3pmn + 7c_4pmn + c_5pmn) + 0.5*(c_1m + c_2n + 5c_3mn + 5c_4mn + c_5mn)$	$O(N^3)$ half of iterations $O(N^2)$ half of iterations

Table7: Run time cost and order of growth

Chapter 8 Conclusion and Future Work

8.1 Conclusion

After implementing all the six methods and analysis of DIEHARD results we applied the two criteria discussed in the previous section, the first criteria is about checking number of values falling in the range of standard deviation from twice of mean, while the second criteria is to find all the values passing in the safe window i.e. in between the range of 0.25 and 0.75.

Applying the discussed criteria, the most successful candidate with maximum number of values falling in the safe range and just one less than the maximum in the $\mu \pm 2\sigma$ range is 3D Moore + Neumann, the computational cost of this method is just $O(N^3)$ with 2048MB memory requirements.

As concluding remarks we suggest 3D Moore + Neumann method for better quality random numbers generation using programmable cellular automata approach, it combines the properties of both Neumann and Moore neighbors at the same time in a three dimensional space.

This also suggests that in general the more number of non-correlated neighbors you pick the more randomness you get but at the cost of increased processing time.

Although we have chosen the best technique in terms of quality of randomness but the requirements of application at hand could be a deciding factor as some applications might require lesser randomness quality but should be efficient, if this is the case then the detail results provided in the appendix and analysis chapter can be consulted to pick the better suited PRNG.

8.2 Future Work

The work done in this research study can be further extended using the following potential research paths.

1. Implementation of other structural organizations that are not explored, various other combinations of Moore and Neumann and their joint schemes can be explored for better results.
2. Strategies proposed in this work can be tested for a better designed algorithm working with different logical operations other than XOR.
3. Implementing the same concepts for $r + 2$ adjacent neighbors, the whole literature survey suggests that all of the researchers went for $r + 1$ neighbors i.e. immediately adjacent neighbors of the current cell, no work is found on $r + 2$ or $r + 3$ neighbors, what effect would be on the results if we consider a larger set of neighbors.
4. Testing the performance of presented solutions on Hardware implementations, specifically the run time efficiencies and complexities involved if the solution is implemented in FPGA.

The points discussed above shows that there is a lot of potential in the field of Random Numbers Generation using Cellular Automata, one can easily opt any path and may find better results in terms of quality of random numbers and in terms of efficiency of a generator.

References

1. P. Hellekalek, "Good Random Number Generators are (not so) Easy to Find", *Mathematics and Computing Simulation*, 46(5-6), pp 485-505 (1998).
2. A Statistical Test Suite for Random and Pseudorandom Number Generators for Cryptographic Applications, NIST (national Institute of Standards and Technology), *Special Publication 800-22 Revision 1a*.
3. Niloy Ganguly et al. A Survey on Cellular Automata, *Center for High Performance Computing, Dresden University of Technology, Dresden, Germany*.
4. S. Nandi and P. P. Chaudhuri, "Analysis of periodic and intermediate boundary 90/150 cellular automata," *IEEE Trans. Comput.*, vol. 45, Jan. 1999.
5. Su Guan "An evolutionary approach to the design of controllable cellular automata structure for random number generation". *IEEE Trans. Evolutionary Computation*, vol. 7, no. 1, pp. 23–36, Feb. 2003
6. Tomassini, M., Sipper, M., Perrenoud, M., "On the generation of high quality random numbers by two-dimensional cellular automata." *IEEE Transactions on Computers* 49, 1146–1151 (2000)
7. Guan, S.-U., Zhang, S., Quieta, M.T, "2-D Variation With Asymmetric Neighborhood for Pseudorandom Number Generation." *IEEE Transaction on Computers* 23, 378–388 (2004).
8. The Marsaglia Random Number CDROM, with The Diehard Battery of Tests of Randomness, produced at Florida State University under a grant from The National Science Foundation (1985).
9. Dr. Mohammed M. Alani, "Testing Randomness in Ciphertext of Block-Ciphers Using DieHard Tests ", *IJCSNS International Journal of Computer Science and Network Security*, VOL.10 No.4, April 2010.
10. Sheng-Uei Guan and Syn Kiat Tan, "Pseudorandom Number Generation With Self-Programmable Cellular Automata", *IEEE TRANSACTIONS ON COMPUTER-AIDED DESIGN OF INTEGRATED CIRCUITS AND SYSTEMS*, VOL. 23, NO. 7, JULY 2004
11. Byung-Heon Kang, Dong-Ho Lee, and Chun-Pyo Hon, "High-Performance Pseudorandom Number Generator Using Two-Dimensional Cellular Automata", *4th IEEE International Symposium on Electronic Design, Test & Applications*

12. Sang-Ho Shin, Geum-Dal Park², and Kee-Young Yoo, "A Virtual Three-Dimension Cellular Automata Pseudorandom Number Generator Based on the Moore Neighborhood Method", *D.-S. Huang et al. (Eds.): ICIC 2008, LNAI 5227*, pp. 174–181, 2008. Springer-Verlag Berlin Heidelberg 2008

13. Sang-Ho Shin, Kee-Young Yoo, "Analysis of 2-State, 3-Neighborhood Cellular Automata Rules for Cryptographic Pseudorandom Number Generation", 2009 *International Conference on Computational Science and Engineering*, 978-0-7695-3823-5/09, 2009 IEEE DOI 0.1109/CSE.2009.299

Appendix A

Important piece of code explained here (for 3D Moore + Neumann Method)

Data structure to hold the random numbers and register values for the SI and DI control registers

```
int random_array [8][8][8];
```

```
int final_randNo [8][8][8];
```

```
int SI[64];
```

```
int CI[64];
```

The master loop calling doing the initialization phase and calling function which generates the random number sequences.

```
for (iteration =1; iteration<=cycles; iteration++)
```

```
{
```

```
        initialize_array();
```

```
        initialize_SI();
```

```
        initialize_CI();
```

```
        generate_random();           //This call gives 64bits of Random
```

```
}
```

The generate_random() function,

```
void generate_random(void)
```

```
{
```

```
int SI_val, CI_val;
```

```
    int control_index=0;
```

```
        int plane, row, col;
```

```
    for (plane=0; plane<=7; plane++)
```

```
    {
```

```
        for (row=0; row<=7; row++)
```

```
        {
```

```

        for (col=0; col<=7; col++)
        {
            SI_val = SI[control_index];
            CI_val = CI[control_index];
            control_index++;
            if (control_index == 64) control_index =0;
        if (SI_val == 0 && CI_val == 0)
            final_randNo[plane][row][col]= applyRule1023(plane,row, col);
        if (SI_val == 1 && CI_val == 0)
            final_randNo[plane][row][col]= applyRule2047(plane,row, col);
        if (SI_val == 0 && CI_val == 1)
            final_randNo[plane][row][col]= applyRule3071(plane,row, col);
        if (SI_val == 1 && CI_val == 1)
            final_randNo[plane][row][col]= applyRule4095(plane,row, col);
        }
    }
}

```

The Cellular Automata Rules Applied (as an example coding for rule 1023 is shown)

```

int applyRule1023(int plane, int row, int col)
{
    //Pick 10NB XOR them and update cell val, 6 Normal + 4 Diagonal
    int n1, n2, n3, n4, n5, n6, n7, n8, n9, n10, val, val1, val2, val3, val4, val5;
    int temprow, tempcol, tempplane;
    int tempval1, tempval2, tempval3, tempval4, tempval5;
    int val6, val7, val8, val9, val10;
    if (col==0)
        n1 = final_randNo[plane][row][7];
    else

```

```

        n1 = final_randNo[plane][row][col-1];
if (col==7)
        n2 = final_randNo[plane][row][0];
else
        n2 = final_randNo[plane][row][col+1];
if (row==0)
        n3 = final_randNo[plane][7][col];
else
        n3 = final_randNo[plane][row-1][col];
if (row==7)
        n4 = final_randNo[plane][0][col];
else
        n4 = final_randNo[plane][row+1][col];
if (plane==0)
        n5 = final_randNo[7][row][col];
else
        n5 = final_randNo[plane-1][row][col];
if (plane==7)
        n6 = final_randNo[0][row][col];
else
        n6 = final_randNo[plane+1][row][col];
// Calculating Indexes for Diagonal Neighbors
//Left Top Diagonal
temprow = row-1;
tempcol = col-1;
tempplane = plane-1;
if (temprow<0) temprow=7;
if (tempcol<0) tempcol=7;

```

```
if (tempplane<0) tempplane = 7;
n7 = final_randNo[tempplane][temprow][tempcol];
//Left Bottom Diagonal
temprow = row+1;
tempcol = col-1;
tempplane = plane+1;
if (temprow>7) temprow=0;
if (tempcol<0) tempcol=7;
if (tempplane>7) tempplane = 0;
n8 = final_randNo[tempplane][temprow][tempcol];
//Right Top Diagonal
temprow = row-1;
tempcol = col+1;
tempplane= plane-1;
if (temprow<0) temprow=7;
if (tempcol>7) tempcol=0;
if (tempplane<0) tempplane = 7;
n9 = final_randNo[tempplane][temprow][tempcol];
//Right Bottom Diagonal
temprow = row+1;
tempcol = col+1;
tempplane = plane+1;
if (temprow>7) temprow=0;
if (tempcol>7) tempcol=0;
if (tempplane>7) tempplane =0;
n10 = final_randNo[tempplane][temprow][tempcol];
val1 = calc_xor(n1, n2);
val2 = calc_xor(n3, n4);
```

```
val3 = calc_xor(n5, n6);  
val4 = calc_xor(n7, n8);  
val5 = calc_xor(n9, n10);  
tempval1 = calc_xor(val1, val2);  
tempval2 = calc_xor(val3, val4);  
tempval3 = calc_xor(val5, val6);  
tempval4 = calc_xor(val7, val8);  
tempval5 = calc_xor(val9, val10);  
val = calc_xor(tempval1, tempval2);  
val = calc_xor(val, tempval3);  
val = calc_xor(val, tempval4);  
val = calc_xor(val, tempval5);  
return val;
```

```
}
```

Appendix B

2D Neumann Neighbors Results for $\mu \pm 3\sigma$

DIEHARD Test Applied	Avg - (3*Stdev)	Avg + (3*Stdev)	Test falling in the Range
BIRTHDAY SPACINGS TEST	-0.37901336	1.4573814	Pass
THE OVERLAPPING 5-PERMUTATION TEST	-0.468926744	1.577111184	Pass
BINARY RANK TEST for 31x31 matrices	1	1	Pass
BINARY RANK TEST for 32x32 matrices	1	1	Pass
BINARY RANK TEST for 6x8 matrices	-0.468123509	1.387819949	Pass
THE BITSTREAM TEST	0.64739095	0.96889153	Pass
OPSO(Overlapping-Pairs-Sparse-Occupancy)	0.314224618	0.747193078	Pass
OQSO(Overlapping-Quadruples-Sparse-Occupancy)	0.356984112	0.709225745	Pass
DNA	0.357390793	0.603823885	Pass
COUNT-THE-1's TEST on a stream of bytes (1st Set of Bytes)	-0.277511751	1.565505631	Pass
COUNT-THE-1's TEST on a stream of bytes (2nd Set of Bytes)	-0.019845513	1.435136513	Pass
COUNT-THE-1's TEST for specific bytes	0.386797919	0.716937619	Pass
PARKING LOT TEST	-0.229618712	1.165533792	Pass
THE MINIMUM DISTANCE TEST	-0.228734204	1.576438724	Pass
THE 3DSPHERES TEST	-0.495386736	1.319387096	Pass
SQUEEZE test	0.735636176	1.159628904	Pass
OVERLAPPING SUMS test	-0.361506123	1.304533283	Pass
runs down	-0.165312854	1.232009374	Pass
runs up	-0.392507335	1.248203295	Pass
Craps Wins	0.167254054	1.207466546	Pass
Craps Throws/Game	-0.280681021	1.254041701	Pass

2D Neumann Neighbors Results for $\mu \pm 2\sigma$

DIEHARD Test Applied	Avg - (2*Stdev)	Avg + (2*Stdev)	Test falling in the Range
BIRTHDAY SPACINGS TEST	-0.072947567	1.151315607	Pass
THE OVERLAPPING 5-PERMUTATION TEST	-0.127920423	1.236104863	Pass
BINARY RANK TEST for 31x31 matrices	1	1	Pass
BINARY RANK TEST for 32x32 matrices	1	1	Pass
BINARY RANK TEST for 6x8 matrices	-0.158799599	1.078496039	Pass
THE BITSTREAM TEST	0.70097438	0.9153081	Pass
OPSO(Overlapping-Pairs-Sparse-Occupancy)	0.386386028	0.675031668	Pass
OQSO(Overlapping-Quadruples-Sparse-Occupancy)	0.415691051	0.650518806	Pass
DNA	0.398462975	0.562751703	Pass
COUNT-THE-1's TEST on a stream of bytes (1st Set of Bytes)	0.029657813	1.258336067	Pass
COUNT-THE-1's TEST on a stream of bytes (2nd Set of Bytes)	0.222651491	1.192639509	Fail
COUNT-THE-1's TEST for specific bytes	0.441821202	0.661914335	Pass
PARKING LOT TEST	0.002906705	0.933008375	Pass
THE MINIMUM DISTANCE TEST	0.07212795	1.27557657	Pass
THE 3DSPHERES TEST	-0.192924431	1.016924791	Pass
SQUEEZE test	0.806301631	1.088963449	Fail
OVERLAPPING SUMS test	-0.083832889	1.026860049	Pass
runs down	0.067574184	0.999122336	Pass
runs up	-0.119055563	0.974751523	Pass
Craps Wins	0.340622803	1.034097797	Pass
Craps Throws/Game	-0.024893901	0.998254581	Pass

2D Moore Neighbors Results for $\mu \pm 3\sigma$

DIEHARD Test Applied	Avg - (3*Stdev)	Avg + (3*Stdev)	Test falling in the Range
BIRTHDAY SPACINGS TEST	-0.405930619	1.281688619	Pass
THE OVERLAPPING 5-PERMUTATION TEST	-0.450601672	1.496250472	Pass
BINARY RANK TEST for 31x31 matrices	-0.03108691	1.26347067	Pass
BINARY RANK TEST for 32x32 matrices	-0.04078692	1.24184764	Pass
BINARY RANK TEST for 6x8 matrices	-0.314643493	1.242396373	Pass
THE BITSTREAM TEST	0.308041156	0.707091764	Pass
OPSO(Overlapping-Pairs-Sparse-Occupancy)	-0.064900167	1.284560629	Pass
OQSO(Overlapping-Quadruples-Sparse-Occupancy)	-0.132204355	1.298127153	Pass
DNA	-0.104396892	1.290750123	Pass
COUNT-THE-1's TEST on a stream of bytes (1st Set of Bytes)	-0.442939472	1.516708272	Pass
COUNT-THE-1's TEST on a stream of bytes (2nd Set of Bytes)	0.135512226	1.278446094	Pass
COUNT-THE-1's TEST for specific bytes	0.372058966	0.662507546	Pass
PARKING LOT TEST	-0.347835562	1.279498442	Pass
THE MINIMUM DISTANCE TEST	-0.41950569	1.35722649	Pass
THE 3DSPHERES TEST	-0.240050889	1.099710969	Pass
SQUEEZE test	-0.369980803	0.944059443	Pass
OVERLAPPING SUMS test	-0.173670267	1.381949307	Pass
runs down	-0.321742978	1.240430418	Pass
runs up	-0.341292964	1.450865524	Pass
Craps Wins	-0.408912737	1.698362897	Pass
Craps Throws/Game	-0.231027673	1.137971113	Pass

2D Moore Neighbors Results for $\mu \pm 2\sigma$

DIEHARD Test Applied	Avg - (2*Stdev)	Avg + (2*Stdev)	Test falling in the Range
BIRTHDAY SPACINGS TEST	-0.124660746	1.000418746	Pass
THE OVERLAPPING 5-PERMUTATION TEST	-0.126126315	1.171775115	Pass
BINARY RANK TEST for 31x31 matrices	0.184672687	1.047711073	Pass
BINARY RANK TEST for 32x32 matrices	0.172985507	1.028075213	Pass
BINARY RANK TEST for 6x8 matrices	-0.055136849	0.982889729	Pass
THE BITSTREAM TEST	0.374549591	0.640583329	Pass
OPSO(Overlapping-Pairs-Sparse-Occupancy)	0.160009965	1.059650496	Pass
OQSO(Overlapping-Quadruples-Sparse-Occupancy)	0.106184229	1.059738568	Pass
DNA	0.12812761	1.05822562	Pass
COUNT-THE-1's TEST on a stream of bytes (1st Set of Bytes)	-0.116331515	1.190100315	Pass
COUNT-THE-1's TEST on a stream of bytes (2nd Set of Bytes)	0.326001204	1.087957116	Pass
COUNT-THE-1's TEST for specific bytes	0.420467063	0.614099449	Pass
PARKING LOT TEST	-0.076613228	1.008276108	Pass
THE MINIMUM DISTANCE TEST	-0.12338366	1.06110446	Pass
THE 3DSPHERES TEST	-0.016757246	0.876417326	Pass
SQUEEZE test	-0.150974095	0.725052735	Pass
OVERLAPPING SUMS test	0.085599662	1.122679378	Fail
runs down	-0.061380745	0.980068185	Pass
runs up	-0.042599883	1.152172443	Pass
Craps Wins	-0.057700131	1.347150291	Pass
Craps Throws/Game	-0.002861209	0.909804649	Fail

2D Neumann + Moore jumping Neighbors Results for $\mu \pm 3\sigma$

DIEHARD Test Applied	Avg - (3*Stdev)	Avg + (3*Stdev)	Test falling in the Range
BIRTHDAY SPACINGS TEST	-0.492415371	1.458226971	Pass
THE OVERLAPPING 5-PERMUTATION TEST	-0.477770625	1.471587345	Pass
BINARY RANK TEST for 31x31 matrices	0.002145266	1.208466334	Pass
BINARY RANK TEST for 32x32 matrices	-0.092559753	1.117303353	Pass
BINARY RANK TEST for 6x8 matrices	-0.439251463	1.294689783	Pass
THE BITSTREAM TEST	0.385449873	0.681850207	Pass
OPSO(Overlapping-Pairs-Sparse-Occupancy)	0.283655443	1.150127121	Pass
OQSO(Overlapping-Quadruples-Sparse-Occupancy)	-0.073994631	1.221349162	Pass
DNA	-0.056520442	1.22619266	Pass
COUNT-THE-1's TEST on a stream of bytes (1st Set of Bytes)	0.096587341	1.129198019	Pass
COUNT-THE-1's TEST on a stream of bytes (2nd Set of Bytes)	-0.158886658	1.336438418	Pass
COUNT-THE-1's TEST for specific bytes	0.344986305	0.687405765	Pass
PARKING LOT TEST	-0.292434096	1.363166336	Pass
THE MINIMUM DISTANCE TEST	-0.253056069	1.553231349	Pass
THE 3DSPHERES TEST	-0.312896071	1.367575591	Pass
SQUEEZE test	-0.007536914	1.381652994	Pass
OVERLAPPING SUMS test	-0.205477795	1.374255955	Pass
runs down	-0.332370843	1.201641243	Pass
runs up	-0.381712114	1.309248594	Pass
Craps Wins	-0.356770545	1.352534625	Pass
Craps Throws/Game	-0.125391412	1.445069812	Pass

2D Neumann + Moore jumping Neighbors Results for $\mu \pm 2\sigma$

DIEHARD Test Applied	Avg - (2*Stdev)	Avg + (2*Stdev)	Test falling in the Range
BIRTHDAY SPACINGS TEST	-0.167308314	1.133119914	Pass
THE OVERLAPPING 5-PERMUTATION TEST	-0.15287763	1.14669435	Pass
BINARY RANK TEST for 31x31 matrices	0.203198777	1.007412823	Pass
BINARY RANK TEST for 32x32 matrices	0.109084098	0.915659502	Pass
BINARY RANK TEST for 6x8 matrices	-0.150261255	1.005699575	Pass
THE BITSTREAM TEST	0.434849929	0.632450151	Pass
OPSO(Overlapping-Pairs-Sparse-Occupancy)	0.428067389	1.005715175	Pass
OQSO(Overlapping-Quadruples-Sparse-Occupancy)	0.141896001	1.00545853	Pass
DNA	0.157265075	1.012407143	Pass
COUNT-THE-1's TEST on a stream of bytes (1st Set of Bytes)	0.26868912	0.95709624	Pass
COUNT-THE-1's TEST on a stream of bytes (2nd Set of Bytes)	0.090334188	1.087217572	Pass
COUNT-THE-1's TEST for specific bytes	0.402056215	0.630335855	Pass
PARKING LOT TEST	-0.016500691	1.087232931	Pass
THE MINIMUM DISTANCE TEST	0.047991834	1.252183446	Pass
THE 3DSPHERES TEST	-0.032817461	1.087496981	Pass
SQUEEZE test	0.223994737	1.150121343	Pass
OVERLAPPING SUMS test	0.057811163	1.110966997	Pass
runs down	-0.076702162	0.945972562	Pass
runs up	-0.099885329	1.027421809	Pass
Craps Wins	-0.07188635	1.06765043	Pass
Craps Throws/Game	0.136352125	1.183326275	Pass

Jumping 2D-3D Neumann Neighbors Results for $\mu \pm 3\sigma$

DIEHARD Test Applied	Avg - (3*Stdev)	Avg + (3*Stdev)	Test falling in the Range
BIRTHDAY SPACINGS TEST	-0.284166685	1.269533085	Pass
THE OVERLAPPING 5-PERMUTATION TEST	-0.236278358	1.635656958	Pass
BINARY RANK TEST for 31x31 matrices	-0.13134028	1.1971842	Pass
BINARY RANK TEST for 32x32 matrices	-0.028839558	1.124761558	Pass
BINARY RANK TEST for 6x8 matrices	-0.185201521	1.279951361	Pass
THE BITSTREAM TEST	0.310501995	0.612340325	Pass
OPSO(Overlapping-Pairs-Sparse-Occupancy)	0.353045146	0.746977115	Pass
OQSO(Overlapping-Quadruples-Sparse-Occupancy)	0.398038539	0.645760318	Pass
DNA	0.372102329	0.68452451	Pass
COUNT-THE-1's TEST on a stream of bytes (1st Set of Bytes)	-0.552849191	1.344949431	Pass
COUNT-THE-1's TEST on a stream of bytes (2nd Set of Bytes)	-0.445457402	1.250162282	Pass
COUNT-THE-1's TEST for specific bytes	0.365247026	0.726112433	Pass
PARKING LOT TEST	-0.327767566	1.405126046	Pass
THE MINIMUM DISTANCE TEST	-0.376181538	1.404031858	Pass
THE 3DSPHERES TEST	-0.452852651	1.460075291	Pass
SQUEEZE test	-0.162542352	1.340220752	Pass
OVERLAPPING SUMS test	-0.300312996	1.192523236	Pass
runs down	-0.335455436	1.388049276	Pass
runs up	-0.437017668	1.414065588	Pass
Craps Wins	-0.105606359	1.498981639	Pass
Craps Throws/Game	0.020667064	1.304900456	Pass

Jumping 2D-3D Neumann Neighbors Results for $\mu \pm 2\sigma$

DIEHARD Test Applied	Avg - (2*Stdev)	Avg + (2*Stdev)	Test falling in the Range
BIRTHDAY SPACINGS TEST	-0.025216724	1.010583124	Pass
THE OVERLAPPING 5-PERMUTATION TEST	0.075710861	1.323667739	Fail
BINARY RANK TEST for 31x31 matrices	0.090080467	0.975763453	Pass
BINARY RANK TEST for 32x32 matrices	0.163427295	0.932494705	Fail
BINARY RANK TEST for 6x8 matrices	0.058990626	1.035759214	Pass
THE BITSTREAM TEST	0.360808383	0.562033937	Fail
OPSO(Overlapping-Pairs-Sparse-Occupancy)	0.418700474	0.681321787	Fail
OQSO(Overlapping-Quadruples-Sparse-Occupancy)	0.439325502	0.604473355	Pass
DNA	0.424172692	0.632454146	Pass
COUNT-THE-1's TEST on a stream of bytes (1st Set of Bytes)	-0.236549421	1.028649661	Pass
COUNT-THE-1's TEST on a stream of bytes (2nd Set of Bytes)	-0.162854121	0.967559001	Pass
COUNT-THE-1's TEST for specific bytes	0.425391261	0.665968199	Pass
PARKING LOT TEST	-0.038951964	1.116310444	Pass
THE MINIMUM DISTANCE TEST	-0.079479305	1.107329625	Pass
THE 3DSPHERES TEST	-0.134031327	1.141253967	Pass
SQUEEZE test	0.087918165	1.089760235	Pass
OVERLAPPING SUMS test	-0.051506958	0.943717198	Pass
runs down	-0.04820465	1.10079849	Pass
runs up	-0.128503792	1.105551712	Pass
Craps Wins	0.161824974	1.231550306	Pass
Craps Throws/Game	0.234705963	1.090861557	Pass

3D Neumann Neighbors Results for $\mu \pm 3\sigma$

DIEHARD Test Applied	Avg - (3*Stdev)	Avg + (3*Stdev)	Test falling in the Range
BIRTHDAY SPACINGS TEST	-0.435741614	1.425704574	Pass
THE OVERLAPPING 5-PERMUTATION TEST	-0.472007462	0.992979062	Pass
BINARY RANK TEST for 31x31 matrices	-0.044220687	1.139097487	Pass
BINARY RANK TEST for 32x32 matrices	-0.046068254	1.139632974	Pass
BINARY RANK TEST for 6x8 matrices	-0.126140145	1.297216305	Pass
THE BITSTREAM TEST	0.353253297	0.661098503	Pass
OPSO(Overlapping-Pairs-Sparse-Occupancy)	0.255205702	0.686328167	Pass
OQSO(Overlapping-Quadruples-Sparse-Occupancy)	0.427632583	0.64652256	Pass
DNA	0.363183714	0.627617834	Pass
COUNT-THE-1's TEST on a stream of bytes (1st Set of Bytes)	-0.312142744	1.031130424	Pass
COUNT-THE-1's TEST on a stream of bytes (2nd Set of Bytes)	-0.26370657	0.60783529	Fail
COUNT-THE-1's TEST for specific bytes	0.389302573	0.759557811	Pass
PARKING LOT TEST	-0.429513575	1.310961175	Pass
THE MINIMUM DISTANCE TEST	-0.35842469	1.27966301	Pass
THE 3DSPHERES TEST	-0.376849976	1.344027976	Pass
SQUEEZE test	0.119969075	1.060244365	Pass
OVERLAPPING SUMS test	-0.42305232	1.27819208	Pass
runs down	-0.448259736	1.314112296	Pass
runs up	-0.312251235	1.417887235	Pass
Craps Wins	-0.273826748	1.313456508	Pass
Craps Throws/Game	-0.066707395	1.400352675	Pass

3D Neumann Neighbors Results for $\mu \pm 2\sigma$

DIEHARD Test Applied	Avg - (2*Stdev)	Avg + (2*Stdev)	Test falling in the Range
BIRTHDAY SPACINGS TEST	-0.125500582	1.115463542	Pass
THE OVERLAPPING 5-PERMUTATION TEST	-0.227843042	0.748814642	Fail
BINARY RANK TEST for 31x31 matrices	0.152999009	0.941877791	Fail
BINARY RANK TEST for 32x32 matrices	0.151548618	0.942016102	Pass
BINARY RANK TEST for 6x8 matrices	0.11108593	1.05999023	Pass
THE BITSTREAM TEST	0.404560831	0.609790969	Pass
OPSO(Overlapping-Pairs-Sparse-Occupancy)	0.327059447	0.614474423	Fail
OQSO(Overlapping-Quadruples-Sparse-Occupancy)	0.464114246	0.610040897	Pass
DNA	0.407256067	0.583545481	Pass
COUNT-THE-1's TEST on a stream of bytes (1st Set of Bytes)	-0.088263882	0.807251562	Pass
COUNT-THE-1's TEST on a stream of bytes (2nd Set of Bytes)	-0.118449593	0.462578313	Fail
COUNT-THE-1's TEST for specific bytes	0.45101178	0.697848604	Pass
PARKING LOT TEST	-0.13943445	1.02088205	Pass
THE MINIMUM DISTANCE TEST	-0.085410074	1.006648394	Pass
THE 3DSPHERES TEST	-0.090036984	1.057214984	Pass
SQUEEZE test	0.276681623	0.903531817	Fail
OVERLAPPING SUMS test	-0.139511587	0.994651347	Pass
runs down	-0.154531064	1.020383624	Pass
runs up	-0.023894824	1.129530824	Pass
Craps Wins	-0.009279539	1.048909299	Pass
Craps Throws/Game	0.177802617	1.155842663	Pass

3D Neumann + Moore Neighbors Results for $\mu \pm 3\sigma$

DIEHARD Test Applied	Avg - (3*Stdev)	Avg + (3*Stdev)	Test falling in the Range
BIRTHDAY SPACINGS TEST	-0.362109399	1.266893879	Pass
THE OVERLAPPING 5-PERMUTATION TEST	-0.56245372	1.41907132	Pass
BINARY RANK TEST for 31x31 matrices	0.134424213	1.109243787	Pass
BINARY RANK TEST for 32x32 matrices	-0.080318823	1.221617143	Pass
BINARY RANK TEST for 6x8 matrices	-0.429557295	1.502997695	Pass
THE BITSTREAM TEST	0.255794304	0.625934056	Pass
OPSO(Overlapping-Pairs-Sparse-Occupancy)	0.266061272	0.694059771	Pass
OQSO(Overlapping-Quadruples-Sparse-Occupancy)	0.373966009	0.58667942	Pass
DNA	0.377457885	0.614176179	Pass
COUNT-THE-1's TEST on a stream of bytes (1st Set of Bytes)	-0.239838667	1.356826667	Pass
COUNT-THE-1's TEST on a stream of bytes (2nd Set of Bytes)	-0.207017927	1.156700007	Pass
COUNT-THE-1's TEST for specific bytes	0.3960481	0.69391325	Pass
PARKING LOT TEST	-0.403298639	1.430999359	Pass
THE MINIMUM DISTANCE TEST	-0.275662406	1.421238886	Pass
THE 3DSPHERES TEST	-0.399816278	1.040410598	Pass
SQUEEZE test	-0.007541349	1.324701509	Pass
OVERLAPPING SUMS test	-0.45990008	1.39001168	Pass
runs down	-0.291639118	1.367565118	Pass
runs up	-0.423079908	1.312661828	Pass
Craps Wins	-0.342775048	0.946304648	Pass
Craps Throws/Game	-0.310330192	1.212486672	Pass

3D Neumann + Moore Neighbors Results for $\mu \pm 2\sigma$

DIEHARD Test Applied	Avg - (2*Stdev)	Avg + (2*Stdev)	Test falling in the Range
BIRTHDAY SPACINGS TEST	-0.090608852	0.995393332	Pass
THE OVERLAPPING 5-PERMUTATION TEST	-0.232199547	1.088817147	Pass
BINARY RANK TEST for 31x31 matrices	0.296894142	0.946773858	Pass
BINARY RANK TEST for 32x32 matrices	0.136670504	1.004627816	Pass
BINARY RANK TEST for 6x8 matrices	-0.107464797	1.180905197	Pass
THE BITSTREAM TEST	0.317484263	0.564244097	Pass
OPSO(Overlapping-Pairs-Sparse-Occupancy)	0.337394356	0.622726688	Pass
OQSO(Overlapping-Quadruples-Sparse-Occupancy)	0.409418244	0.551227185	Pass
DNA	0.416910934	0.57472313	Pass
COUNT-THE-1's TEST on a stream of bytes (1st Set of Bytes)	0.026272222	1.090715778	Pass
COUNT-THE-1's TEST on a stream of bytes (2nd Set of Bytes)	0.020268395	0.929413685	Pass
COUNT-THE-1's TEST for specific bytes	0.445692292	0.644269058	Pass
PARKING LOT TEST	-0.097582306	1.125283026	Pass
THE MINIMUM DISTANCE TEST	0.007154476	1.138422004	Pass
THE 3DSPHERES TEST	-0.159778465	0.800372785	Fail
SQUEEZE test	0.214499128	1.102661032	Pass
OVERLAPPING SUMS test	-0.151581453	1.081693053	Pass
runs down	-0.015105078	1.091031078	Pass
runs up	-0.133789618	1.023371538	Pass
Craps Wins	-0.127928432	0.731458032	Pass
Craps Throws/Game	-0.056527381	0.958683861	Pass

Summary of results for $\mu \pm 3\sigma$

No of Passes Avg + - 3 Std. Dev 99% Values Fall

Jumping 2D-3D	Rand3DVonNeuman	3D Moore + Neumann	2D Moore	2D Von Neumann Moore Jumping	2D Von Neumann
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Fail	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Failures	0	1	0	0	0

Summary of results for $\mu \pm 2\sigma$

No of Passes Avg + - 2 Std. Dev 95% Values Fail					
Jumping 2D-3D	Rand3DVonNeuman	3D Moore + Neumann	2D Moore	2D Von Neumann Moore Jumping	2D Von Neumann
Pass	Pass	Pass	Pass	Pass	Pass
Fail	Fail	Pass	Pass	Pass	Pass
Pass	Fail	Pass	Pass	Pass	Pass
Fail	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Fail	Pass	Pass	Pass	Pass	Pass
Fail	Fail	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Fail	Pass	Pass	Pass	Fail
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Fail	Pass	Pass	Pass
Pass	Fail	Pass	Pass	Pass	Fail
Pass	Pass	Pass	Fail	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Pass	Pass	Pass
Pass	Pass	Pass	Fail	Pass	Pass
Failures	4	5	1	0	2

Appendix C

Results for Safe range value (values ≥ 0.25 AND values ≤ 0.75)
DIEHARD Tests Name used in abbreviated form.

All 100 values Comparison (values ≥ 0.25 AND values ≤ 0.75) Safe Range

	1	2	3	4	5	6
	Jumping 2D-3D	Rand 3D Von Neumann	3D Moore + Neumann	2D Moore Neighbours	2D Von Neumann Moore Jump N	2D Von Neumann Neighbors
BST	68	40	56	52	44	56
O5PT	28	24	36	36	36	28
BR31	80	84	84	72	68	0
BR32	84	76	72	76	80	0
BR6X8	60	60	44	56	52	36
BS	100	100	100	100	100	8
OPSO	100	100	100	76	80	100
OQSO	100	100	100	76	80	100
DNA	100	100	100	76	80	100
CT1TSB-1	40	52	48	40	80	28
CT1TSB-2	48	20	72	60	68	40
CT1TSB	100	100	100	100	100	100
PLT	48	40	40	52	60	64
MDT	48	64	52	44	28	32
3DST	36	48	48	72	52	48
ST	44	84	48	40	44	0
OST	56	56	52	48	52	60
RD	64	48	56	56	56	60
RU	40	48	52	44	40	52
CW	36	64	56	24	52	64
CTG	60	48	44	64	52	56

	Jumping 2D-3D	Rand 3D Von Neumann	3D Moore + Neumann	2D Moore Neighbours	2D Von Neumann Moore Jump N	2D Von Neumann Neighbors
Max Occurance	9	10	9	5	4	8

Appendix D

Sample of Binary file (Snap shot taken for 3D Moore + Neumann Method)

	00	01	02	03	04	05	06	07	08	09	0a	0b	0c	0d	0e	0f	
00000000	52	41	80	42	45	ad	f0	24	60	4a	0b	a9	fc	b0	34	3b	RAÆBE-8ç`J.ëù°4;
00000010	59	f6	ce	6f	6d	68	ba	2a	21	be	23	d9	c1	58	40	2d	Yôïomh°*!%#ÙÁX@-
00000020	17	78	11	b3	a8	1a	22	91	b3	79	20	af	0c	16	2d	3c	.x.'.'."''y'`..-<
00000030	78	fa	e1	49	a9	92	d4	7d	4a	a5	d9	5e	27	67	41	25	xúáIe'ô)JÏÛ^'gA%
00000040	5e	0c	52	6d	cb	f9	81	56	64	5c	ae	8a	90	1a	43	62	^.RmËù□Vd\øŠ□.Cb
00000050	f8	cc	8d	4a	9f	3b	88	d6	06	58	bd	e2	1a	1b	03	75	øï□Jÿ;^ö.X%â...u
00000060	6a	8b	52	69	e4	20	6f	99	59	e3	3c	4a	1b	b5	c3	f8	j<Riâ o**Yâ<J.µÃø
00000070	f0	6f	e9	1f	eb	4c	48	b4	48	e3	e0	c4	87	fc	27	7a	ðoé.èLH'HããÀ+ù'z
00000080	e4	0c	58	79	a4	87	06	04	19	77	c2	84	fb	01	c3	0e	ä.Xy#t...wÃ.ù.Ã.
00000090	5f	12	06	ce	ad	12	fb	49	d2	2a	67	27	8a	e5	1c	8d	...î-.ûIô*g'šã.□
000000a0	c6	7a	f8	ac	13	16	96	fd	5e	d5	5a	9b	a7	17	41	07	Æzø-...-ý^ôZ>S.A.
000000b0	e9	f6	e2	bb	6f	2b	78	05	e1	7f	3f	84	ac	3c	6e	63	éôã>o+x.á□?¿-<nc
000000c0	dd	c8	0b	a2	9e	d2	32	c7	7e	85	46	fb	cf	8d	9c	47	ÝÈ.<žô2ç~...FûI□œG
000000d0	c9	0d	17	79	b6	c3	21	de	eb	e9	2d	5b	4e	79	22	25	É...yŕÃ!Pëé-[Ny""%
000000e0	cb	90	d5	74	18	d7	fb	d8	0a	48	4e	a9	26	36	12	15	Ë□ô't.×ûø.HNø<6..
000000f0	03	6a	6e	33	fa	2b	b7	36	cc	c4	a0	77	4a	e1	29	cd	.jn3ú+·6îÃ wJá)Í
00000100	14	3e	23	d2	e4	38	54	45	ad	34	46	7c	a0	f0	3c	9b	.>#ôã8TE-4F 8<>
00000110	5d	25	8b	f0	e4	a3	cb	55	77	65	cb	0b	31	af	ac	ad]*<ðã&ËUweË.l'--
00000120	22	b3	30	d4	47	3e	7e	bf	81	05	1f	6e	05	b3	21	c0	""ôôG>~ç□.n.'!À
00000130	57	80	1c	ca	47	3a	44	2f	66	ef	a7	b8	18	02	0d	29	W€.ËG:D/fiS,...)
00000140	b9	fb	42	c9	55	2f	3e	bd	07	4c	7f	75	94	65	eb	01	'ûBËU/>%.L□u"eë.
00000150	70	04	91	e4	d1	ef	2b	44	73	78	34	13	73	d3	95	d8	p.'ãÑi+Dsx4.só·ø
00000160	01	3d	06	f0	b7	d3	ca	b3	3c	41	2f	df	c9	13	20	2f	.=.ð·ôË°<A/BË. /
00000170	de	9d	77	8f	59	5f	e0	c8	2f	b6	5f	b3	8a	49	d9	cc	P□w□Y_àÈ/ŕ_'šIÙÏ
00000180	e2	af	7e	d9	17	c2	e1	7d	54	c9	ba	50	27	f5	bd	e1	ã^-Û.Ãá)TË°P'ð%á
00000190	2b	10	da	b4	5d	51	3d	20	16	b6	1d	72	73	b8	19	cc	+.'Û'JQ=.ŕ.r.s_.î

Appendix E

Sample of DIEHARD Output file (Snap shot taken for 3D Moore + Neumann Method)

BIRTHDAY SPACINGS TEST, M= 512 N=2**24 LAMBDA= 2.0000

Results for file1

For a sample of size 500: mean
file1 using bits 1 to 24 2.068
duplicate number number
spacings observed expected
0 67. 67.668
1 118. 135.335
2 143. 135.335
3 103. 90.224
4 40. 45.112
5 17. 18.045
6 to INF 12. 8.282
Chisquare with 6 d.o.f. = 6.78 p-value= .658276
.....

For a sample of size 500: mean
file1 using bits 2 to 25 1.938
duplicate number number
spacings observed expected
0 60. 67.668
1 148. 135.335
2 145. 135.335
3 92. 90.224
4 27. 45.112
5 22. 18.045
6 to INF 6. 8.282
Chisquare with 6 d.o.f. = 11.55 p-value= .927114
.....

For a sample of size 500: mean
file1 using bits 3 to 26 1.956
duplicate number number
spacings observed expected
0 71. 67.668
1 142. 135.335
2 124. 135.335
3 99. 90.224
4 38. 45.112
5 21. 18.045
6 to INF 5. 8.282
Chisquare with 6 d.o.f. = 5.20 p-value= .481707
.....

For a sample of size 500: mean
file1 using bits 4 to 27 1.960
duplicate number number
spacings observed expected

0	58.	67.668
1	137.	135.335
2	158.	135.335
3	90.	90.224
4	41.	45.112
5	10.	18.045
6 to INF	6.	8.282

Chisquare with 6 d.o.f. = 9.79 p-value= .866126

.....

For a sample of size 500: mean

file1	using bits 5 to 28	2.116
duplicate	number	number
spacings	observed	expected
0	68.	67.668
1	133.	135.335
2	123.	135.335
3	83.	90.224
4	53.	45.112
5	27.	18.045
6 to INF	13.	8.282

Chisquare with 6 d.o.f. = 10.26 p-value= .885733

.....

For a sample of size 500: mean

file1	using bits 6 to 29	1.976
duplicate	number	number
spacings	observed	expected
0	72.	67.668
1	133.	135.335
2	130.	135.335
3	98.	90.224
4	42.	45.112
5	19.	18.045
6 to INF	6.	8.282

Chisquare with 6 d.o.f. = 2.09 p-value= .088971

.....

For a sample of size 500: mean

file1	using bits 7 to 30	1.990
duplicate	number	number
spacings	observed	expected
0	59.	67.668
1	148.	135.335
2	141.	135.335
3	77.	90.224
4	51.	45.112
5	15.	18.045
6 to INF	9.	8.282

Chisquare with 6 d.o.f. = 5.82 p-value= .555793

.....

For a sample of size 500: mean

file1	using bits 8 to 31	2.032
duplicate	number	number

spacings	observed	expected
0	62.	67.668
1	137.	135.335
2	140.	135.335
3	88.	90.224
4	41.	45.112
5	22.	18.045
6 to INF	10.	8.282

Chisquare with 6 d.o.f. = 2.31 p-value= .110796

.....

For a sample of size 500: mean

file1	using bits	9 to 32	2.090
duplicate	number	number	

spacings	observed	expected
0	47.	67.668
1	148.	135.335
2	139.	135.335
3	85.	90.224
4	52.	45.112
5	18.	18.045
6 to INF	11.	8.282

Chisquare with 6 d.o.f. = 9.84 p-value= .868594

.....

The 9 p-values were

.658276	.927114	.481707	.866126	.885733
.088971	.555793	.110796	.868594	

A KSTEST for the 9 p-values yields .691006

OPERM5 test for file file1

For a sample of 1,000,000 consecutive 5-tuples,
 chisquare for 99 degrees of freedom= 81.134; p-value= .095538

OPERM5 test for file file1

For a sample of 1,000,000 consecutive 5-tuples,
 chisquare for 99 degrees of freedom= 94.586; p-value= .393169

Binary rank test for file1

Rank test for 31x31 binary matrices:
 rows from leftmost 31 bits of each 32-bit integer

rank	observed	expected	(o-e)^2/e	sum
28	195	211.4	1.274968	1.275
29	5161	5134.0	.141886	1.417
30	23196	23103.0	.373989	1.791
31	11448	11551.5	.927783	2.719

chisquare= 2.719 for 3 d. of f.; p-value= .612521

Binary rank test for file1

Rank test for 32x32 binary matrices:
 rows from leftmost 32 bits of each 32-bit integer

rank	observed	expected	$(o-e)^2/e$	sum
29	195	211.4	1.274968	1.275
30	5107	5134.0	.142102	1.417
31	23086	23103.0	.012578	1.430
32	11612	11551.5	.316607	1.746

chisquare= 1.746 for 3 d. of f.; p-value= .466475

Binary Rank Test for file1

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 1 to 8

	OBSERVED	EXPECTED	$(O-E)^2/E$	SUM
$r \leq 4$	932	944.3	.160	.160
$r = 5$	22075	21743.9	5.042	5.202
$r = 6$	76993	77311.8	1.315	6.517

$$p=1-\exp(-SUM/2)= .96155$$

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 2 to 9

	OBSERVED	EXPECTED	$(O-E)^2/E$	SUM
$r \leq 4$	922	944.3	.527	.527
$r = 5$	21848	21743.9	.498	1.025
$r = 6$	77230	77311.8	.087	1.112

$$p=1-\exp(-SUM/2)= .42639$$

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 3 to 10

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
r<=4	952	944.3	.063	.063
r =5	21952	21743.9	1.992	2.054
r =6	77096	77311.8	.602	2.657
p=1-exp(-SUM/2)= .73509				

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 4 to 11

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
r<=4	965	944.3	.454	.454
r =5	21562	21743.9	1.522	1.975
r =6	77473	77311.8	.336	2.312
p=1-exp(-SUM/2)= .68518				

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 5 to 12

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
r<=4	969	944.3	.646	.646
r =5	21465	21743.9	3.577	4.223
r =6	77566	77311.8	.836	5.059
p=1-exp(-SUM/2)= .92031				

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 6 to 13

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
--	----------	----------	-----------	-----

r<=4 945 944.3 .001 .001

r =5 21674 21743.9 .225 .225

r =6 77381 77311.8 .062 .287

$$p=1-\exp(-\text{SUM}/2)= .13375$$

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 7 to 14

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
--	----------	----------	-----------	-----

r<=4	976	944.3	1.064	1.064
------	-----	-------	-------	-------

r =5	21475	21743.9	3.325	4.389
------	-------	---------	-------	-------

r =6	77549	77311.8	.728	5.117
------	-------	---------	------	-------

$$p=1-\exp(-\text{SUM}/2)= .92259$$

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 8 to 15

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
--	----------	----------	-----------	-----

r<=4	924	944.3	.436	.436
------	-----	-------	------	------

r =5	21936	21743.9	1.697	2.134
------	-------	---------	-------	-------

r =6	77140	77311.8	.382	2.515
------	-------	---------	------	-------

$$p=1-\exp(-\text{SUM}/2)= .71569$$

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 9 to 16

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
--	----------	----------	-----------	-----

r<=4	961	944.3	.295	.295
------	-----	-------	------	------

r =5	22042	21743.9	4.087	4.382
------	-------	---------	-------	-------

r =6 76997 77311.8 1.282 5.664

$$p=1-\exp(-\text{SUM}/2)= .94110$$

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 10 to 17

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
r<=4	954	944.3	.100	.100
r =5	21496	21743.9	2.826	2.926
r =6	77550	77311.8	.734	3.660

$$p=1-\exp(-\text{SUM}/2)= .83957$$

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 11 to 18

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
r<=4	905	944.3	1.636	1.636
r =5	21575	21743.9	1.312	2.948
r =6	77520	77311.8	.561	3.508

$$p=1-\exp(-\text{SUM}/2)= .82695$$

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 12 to 19

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
r<=4	954	944.3	.100	.100
r =5	21935	21743.9	1.680	1.779
r =6	77111	77311.8	.522	2.301

$$p=1-\exp(-\text{SUM}/2)= .68347$$

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 13 to 20

	OBSERVED	EXPECTED	(O-E) ² /E	SUM
r<=4	934	944.3	.112	.112
r =5	21884	21743.9	.903	1.015
r =6	77182	77311.8	.218	1.233
p=1-exp(-SUM/2)= .46017				

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 14 to 21

	OBSERVED	EXPECTED	(O-E) ² /E	SUM
r<=4	962	944.3	.332	.332
r =5	21859	21743.9	.609	.941
r =6	77179	77311.8	.228	1.169
p=1-exp(-SUM/2)= .44265				

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 15 to 22

	OBSERVED	EXPECTED	(O-E) ² /E	SUM
r<=4	948	944.3	.014	.014
r =5	21705	21743.9	.070	.084
r =6	77347	77311.8	.016	.100
p=1-exp(-SUM/2)= .04882				

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 16 to 23

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
r<=4	901	944.3	1.986	1.986
r =5	21797	21743.9	.130	2.115
r =6	77302	77311.8	.001	2.117
p=1-exp(-SUM/2)= .65294				

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 17 to 24

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
r<=4	876	944.3	4.940	4.940
r =5	21771	21743.9	.034	4.974
r =6	77353	77311.8	.022	4.996
p=1-exp(-SUM/2)= .91775				

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 18 to 25

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
r<=4	919	944.3	.678	.678
r =5	21678	21743.9	.200	.878
r =6	77403	77311.8	.108	.985
p=1-exp(-SUM/2)= .38897				

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 19 to 26

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
--	----------	----------	-----------	-----

r<=4	959	944.3	.229	.229
------	-----	-------	------	------

r =5	21604	21743.9	.900	1.129
------	-------	---------	------	-------

r =6	77437	77311.8	.203	1.332
------	-------	---------	------	-------

$$p=1-\exp(-\text{SUM}/2)= .48615$$

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 20 to 27

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
--	----------	----------	-----------	-----

r<=4	904	944.3	1.720	1.720
------	-----	-------	-------	-------

r =5	21675	21743.9	.218	1.938
------	-------	---------	------	-------

r =6	77421	77311.8	.154	2.093
------	-------	---------	------	-------

$$p=1-\exp(-\text{SUM}/2)= .64876$$

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 21 to 28

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
--	----------	----------	-----------	-----

r<=4	895	944.3	2.574	2.574
------	-----	-------	-------	-------

r =5	21801	21743.9	.150	2.724
------	-------	---------	------	-------

r =6	77304	77311.8	.001	2.725
------	-------	---------	------	-------

$$p=1-\exp(-\text{SUM}/2)= .74394$$

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 22 to 29

	OBSERVED	EXPECTED	(O-E)^2/E	SUM
--	----------	----------	-----------	-----

r<=4	932	944.3	.160	.160
------	-----	-------	------	------

r =5	21729	21743.9	.010	.170
------	-------	---------	------	------

r =6 77339 77311.8 .010 .180

$$p=1-\exp(-\text{SUM}/2)= .08608$$

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 23 to 30

	OBSERVED	EXPECTED	(O-E) ² /E	SUM
r<=4	967	944.3	.546	.546
r =5	21710	21743.9	.053	.598
r =6	77323	77311.8	.002	.600

$$p=1-\exp(-\text{SUM}/2)= .25922$$

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 24 to 31

	OBSERVED	EXPECTED	(O-E) ² /E	SUM
r<=4	920	944.3	.625	.625
r =5	21709	21743.9	.056	.681
r =6	77371	77311.8	.045	.727

$$p=1-\exp(-\text{SUM}/2)= .30467$$

Rank of a 6x8 binary matrix,

rows formed from eight bits of the RNG file1

b-rank test for bits 25 to 32

	OBSERVED	EXPECTED	(O-E) ² /E	SUM
r<=4	946	944.3	.003	.003
r =5	21697	21743.9	.101	.104
r =6	77357	77311.8	.026	.131

$$p=1-\exp(-\text{SUM}/2)= .06323$$

TEST SUMMARY, 25 tests on 100,000 random 6x8 matrices

These should be 25 uniform [0,1] random variables:

.961546	.426393	.735095	.685179	.920306
.133747	.922587	.715689	.941104	.839569
.826948	.683471	.460169	.442649	.048820
.652939	.917749	.388968	.486151	.648756
.743944	.086080	.259218	.304666	.063231

brank test summary for file1

The KS test for those 25 supposed UNI's yields

KS p-value= .670466

THE OVERLAPPING 20-tuples BITSTREAM TEST, 20 BITS PER WORD, N words

This test uses $N=2^{21}$ and samples the bitstream 20 times.

No. missing words should average 141909. with $\sigma=428$.

tst no 1:	141373 missing words,	-1.25 sigmas from mean,	p-value= .10508
tst no 2:	141963 missing words,	.13 sigmas from mean,	p-value= .54990
tst no 3:	142320 missing words,	.96 sigmas from mean,	p-value= .83135
tst no 4:	142154 missing words,	.57 sigmas from mean,	p-value= .71622
tst no 5:	142078 missing words,	.39 sigmas from mean,	p-value= .65324
tst no 6:	141919 missing words,	.02 sigmas from mean,	p-value= .50901
tst no 7:	142436 missing words,	1.23 sigmas from mean,	p-value= .89075
tst no 8:	141492 missing words,	-.98 sigmas from mean,	p-value= .16476
tst no 9:	141610 missing words,	-.70 sigmas from mean,	p-value= .24216
tst no 10:	142457 missing words,	1.28 sigmas from mean,	p-value= .89966
tst no 11:	141985 missing words,	.18 sigmas from mean,	p-value= .57017
tst no 12:	141711 missing words,	-.46 sigmas from mean,	p-value= .32154

tst no 13: 141792 missing words, -.27 sigmas from mean, p-value= .39199
 tst no 14: 141981 missing words, .17 sigmas from mean, p-value= .56650
 tst no 15: 141545 missing words, -.85 sigmas from mean, p-value= .19732
 tst no 16: 142313 missing words, .94 sigmas from mean, p-value= .82720
 tst no 17: 142317 missing words, .95 sigmas from mean, p-value= .82958
 tst no 18: 142655 missing words, 1.74 sigmas from mean, p-value= .95927
 tst no 19: 141925 missing words, .04 sigmas from mean, p-value= .51460
 tst no 20: 141489 missing words, -.98 sigmas from mean, p-value= .16303

OPSO test for generator file1

Output: No. missing words (mw), equiv normal variate (z), p-value (p)

OPSO for file1	using bits 23 to 32	142263	1.220	.8887
OPSO for file1	using bits 22 to 31	141778	-.453	.3253
OPSO for file1	using bits 21 to 30	141893	-.056	.4775
OPSO for file1	using bits 20 to 29	141881	-.098	.4611
OPSO for file1	using bits 19 to 28	141745	-.567	.2855
OPSO for file1	using bits 18 to 27	141796	-.391	.3480
OPSO for file1	using bits 17 to 26	142021	.385	.6499
OPSO for file1	using bits 16 to 25	141711	-.684	.2470
OPSO for file1	using bits 15 to 24	141651	-.891	.1865
OPSO for file1	using bits 14 to 23	141793	-.401	.3442
OPSO for file1	using bits 13 to 22	141866	-.149	.4406
OPSO for file1	using bits 12 to 21	141986	.264	.6043
OPSO for file1	using bits 11 to 20	141616	-1.011	.1559
OPSO for file1	using bits 10 to 19	141921	.040	.5161

OPSO for file1	using bits 9 to 18	141857	-.180	.4284
OPSO for file1	using bits 8 to 17	142073	.564	.7138
OPSO for file1	using bits 7 to 16	141483	-1.470	.0708
OPSO for file1	using bits 6 to 15	141471	-1.511	.0653
OPSO for file1	using bits 5 to 14	141986	.264	.6043
OPSO for file1	using bits 4 to 13	141602	-1.060	.1446
OPSO for file1	using bits 3 to 12	141685	-.774	.2196
OPSO for file1	using bits 2 to 11	141473	-1.505	.0662
OPSO for file1	using bits 1 to 10	141567	-1.180	.1189

OQSO test for generator file1

Output: No. missing words (mw), equiv normal variate (z), p-value (p)

OQSO for file1	using bits 28 to 32	141863	-.157	.4376
OQSO for file1	using bits 27 to 31	142315	1.375	.9155
OQSO for file1	using bits 26 to 30	141982	.246	.5973
OQSO for file1	using bits 25 to 29	142036	.429	.6662
OQSO for file1	using bits 24 to 28	142131	.751	.7738
OQSO for file1	using bits 23 to 27	142027	.399	.6550
OQSO for file1	using bits 22 to 26	142093	.623	.7332
OQSO for file1	using bits 21 to 25	142086	.599	.7254
OQSO for file1	using bits 20 to 24	141587	-1.093	.1373
OQSO for file1	using bits 19 to 23	142192	.958	.8310
OQSO for file1	using bits 18 to 22	141632	-.940	.1736
OQSO for file1	using bits 17 to 21	141389	-1.764	.0389
OQSO for file1	using bits 16 to 20	142029	.406	.6575
OQSO for file1	using bits 15 to 19	141930	.070	.5279
OQSO for file1	using bits 14 to 18	141778	-.445	.3281

OQSO for file1	using bits 13 to 17	142085	.595	.7242
OQSO for file1	using bits 12 to 16	141945	.121	.5481
OQSO for file1	using bits 11 to 15	142048	.470	.6808
OQSO for file1	using bits 10 to 14	141764	-.493	.3111
OQSO for file1	using bits 9 to 13	142145	.799	.7878
OQSO for file1	using bits 8 to 12	141706	-.689	.2453
OQSO for file1	using bits 7 to 11	142063	.521	.6988
OQSO for file1	using bits 6 to 10	141460	-1.523	.0639
OQSO for file1	using bits 5 to 9	142149	.812	.7917
OQSO for file1	using bits 4 to 8	141911	.006	.5023
OQSO for file1	using bits 3 to 7	142017	.365	.6424
OQSO for file1	using bits 2 to 6	141732	-.601	.2739
OQSO for file1	using bits 1 to 5	141645	-.896	.1851

DNA test for generator file1

Output: No. missing words (mw), equiv normal variate (z), p-value (p)

DNA for file1	using bits 20 to 21	141539	-1.092	.1373
DNA for file1	using bits 19 to 20	142153	.719	.7639
DNA for file1	using bits 18 to 19	142427	1.527	.9366
DNA for file1	using bits 17 to 18	141747	-.479	.3160
DNA for file1	using bits 16 to 17	142454	1.607	.9459
DNA for file1	using bits 15 to 16	142369	1.356	.9124
DNA for file1	using bits 14 to 15	142260	1.034	.8495
DNA for file1	using bits 13 to 14	141607	-.892	.1862
DNA for file1	using bits 12 to 13	141859	-.148	.4410
DNA for file1	using bits 11 to 12	141649	-.768	.2213
DNA for file1	using bits 10 to 11	141656	-.747	.2274

DNA for file1	using bits 9 to 10	141961	.152	.5606
DNA for file1	using bits 8 to 9	141525	-1.134	.1285
DNA for file1	using bits 7 to 8	142007	.288	.6134
DNA for file1	using bits 6 to 7	142050	.415	.6609
DNA for file1	using bits 5 to 6	142105	.577	.7181
DNA for file1	using bits 4 to 5	141334	-1.697	.0448
DNA for file1	using bits 3 to 4	141754	-.458	.3234
DNA for file1	using bits 2 to 3	141784	-.370	.3558
DNA for file1	using bits 1 to 2	141747	-.479	.3160

Test results for file1

Chi-square with $5^5-5^4=2500$ d.of f. for sample size:2560000

chisquare equiv normal p-value

Results fo COUNT-THE-1's in successive bytes:

byte stream for file1	2405.69	-1.334	.091132
byte stream for file1	2502.18	.031	.512312

Results for COUNT-THE-1's in specified bytes:

bits 1 to 8	2485.26	-.208	.417458
bits 2 to 9	2533.00	.467	.679619
bits 3 to 10	2459.37	-.575	.282770
bits 4 to 11	2523.00	.325	.627492
bits 5 to 12	2423.42	-1.083	.139397
bits 6 to 13	2490.86	-.129	.448556
bits 7 to 14	2518.10	.256	.600999
bits 8 to 15	2656.09	2.207	.986360

bits 9 to 16	2614.31	1.617	.947014
bits 10 to 17	2429.86	-.992	.160633

CDPARK: result of ten tests on file file1

Of 12,000 tries, the average no. of successes
should be 3523 with sigma=21.9

Successes: 3522 z-score: -.046 p-value: .481790

Successes: 3537 z-score: .639 p-value: .738676

Successes: 3526 z-score: .137 p-value: .554479

Successes: 3534 z-score: .502 p-value: .692266

Successes: 3505 z-score: -.822 p-value: .205562

Successes: 3546 z-score: 1.050 p-value: .853193

Successes: 3536 z-score: .594 p-value: .723613

Successes: 3513 z-score: -.457 p-value: .323972

Successes: 3562 z-score: 1.781 p-value: .962529

Successes: 3526 z-score: .137 p-value: .554479

square size avg. no. parked sample sigma

100. 3530.700 15.411

KSTEST for the above 10: p= .638628

This is the MINIMUM DISTANCE test

for random integers in the file file1

Sample no. d^2 avg equiv uni

5 1.3638 1.0519 .746056

10 .6086 .7828 .457574

15	.3159	.6406	.272010
20	.1821	.9221	.167223
25	.0348	.8639	.034414
30	.0262	.8330	.026021
35	.8224	1.0543	.562435
40	.9775	1.1022	.625586
45	.0502	1.0318	.049235
50	.9500	.9946	.615120
55	.2766	.9520	.242659
60	.0701	.8943	.068018
65	.2568	.8797	.227451
70	3.7634	.8857	.977230
75	7.1344	.9775	.999231
80	1.3870	.9895	.751923
85	.5250	1.0089	.409974
90	2.8711	1.0000	.944175
95	.0495	1.0115	.048536
100	2.0191	1.0343	.868568

MINIMUM DISTANCE TEST for file1

Result of KS test on 20 transformed mindist²'s:

p-value= .533812

The 3DSPHERES test for file file1

sample no: 1	$r^3=$ 38.193	p-value= .72004
sample no: 2	$r^3=$ 65.735	p-value= .88821
sample no: 3	$r^3=$ 24.364	p-value= .55609
sample no: 4	$r^3=$ 12.159	p-value= .33323
sample no: 5	$r^3=$ 10.890	p-value= .30441
sample no: 6	$r^3=$ 23.250	p-value= .53929
sample no: 7	$r^3=$ 77.702	p-value= .92499
sample no: 8	$r^3=$ 2.731	p-value= .08700
sample no: 9	$r^3=$ 25.410	p-value= .57131
sample no: 10	$r^3=$ 20.434	p-value= .49396
sample no: 11	$r^3=$ 9.450	p-value= .27021
sample no: 12	$r^3=$ 6.442	p-value= .19323
sample no: 13	$r^3=$ 11.870	p-value= .32676
sample no: 14	$r^3=$ 1.689	p-value= .05474
sample no: 15	$r^3=$ 92.819	p-value= .95468
sample no: 16	$r^3=$ 6.956	p-value= .20694
sample no: 17	$r^3=$ 11.314	p-value= .31418
sample no: 18	$r^3=$ 31.576	p-value= .65095
sample no: 19	$r^3=$ 2.126	p-value= .06841
sample no: 20	$r^3=$ 30.239	p-value= .63504

A KS test is applied to those 20 p-values.

3DSPHERES test for file file1

p-value= .241794

RESULTS OF SQUEEZE TEST FOR file1

Table of standardized frequency counts

$(\text{obs-exp})/\sqrt{\text{exp}})^2$

for j taking values $\leq 6, 7, 8, \dots, 47, \geq 48$:

2.7	.5	.8	-2.4	-.5	-.1
.7	-1.1	.1	-.4	-1.9	-1.6
1.2	.6	-.8	-.7	1.0	-.2
.8	-1.6	-.8	.2	1.5	1.5
-.2	.9	.0	-1.7	2.9	1.3
.7	.4	-1.1	.9	-.5	1.3
.3	.5	.1	-.1	.9	.0

2.7

Chi-square with 42 degrees of freedom: 62.622

z-score= 2.250 p-value= .978858

Test no. 1 p-value .430451

Test no. 2 p-value .217270

Test no. 3 p-value .622948

Test no. 4 p-value .057847

Test no. 5 p-value .793093

Test no. 6 p-value .142356

Test no. 7 p-value .004947

Test no. 8 p-value .792843

Test no. 9 p-value .983215

Test no. 10 p-value .974643

Results of the OSUM test for file1

KSTEST on the above 10 p-values: .638171

Run test for file1 :

runs up; ks test for 10 p's: .949285

runs down; ks test for 10 p's: .135459

Run test for file1 :

runs up; ks test for 10 p's: .109780

runs down; ks test for 10 p's: .911528

Results of craps test for file1

No. of wins: Observed Expected

98561 98585.86

98561= No. of wins, z-score= -.111 pvalue= .45573

Analysis of Throws-per-Game:

Chisq= 15.24 for 20 degrees of freedom, p= .23739

Throws	Observed	Expected	Chisq	Sum
1	66795	66666.7	.247	.247
2	37579	37654.3	.151	.398
3	27003	26954.7	.086	.484
4	19205	19313.5	.609	1.093
5	13753	13851.4	.699	1.793
6	10135	9943.5	3.686	5.479
7	7128	7145.0	.041	5.520
8	5035	5139.1	2.108	7.627
9	3695	3699.9	.006	7.633
10	2733	2666.3	1.669	9.302
11	1867	1923.3	1.650	10.952

12	1370	1388.7	.253	11.205
13	1025	1003.7	.451	11.656
14	726	726.1	.000	11.656
15	516	525.8	.184	11.840
16	404	381.2	1.370	13.210
17	297	276.5	1.514	14.724
18	206	200.8	.133	14.857
19	140	146.0	.245	15.102
20	107	106.2	.006	15.108
21	281	287.1	.130	15.238

SUMMARY FOR file1

p-value for no. of wins: .455735

p-value for throws/game: .237386