

An Overview of Cryptographically Secure Pseudorandom Number generators and BBS

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ABSTRACT

In this manuscript we have presented a literature survey of cryptographically secure pseudo random number generators, their requirements regarding statistical properties and next bit test. The paper also provides a brief overview of Blum Blum Shub (BBS) Generator specifically, which is considered to be the best cryptographically secure pseudorandom number generator. We have performed the rigorous testing of BBS generator on National Institute of Science and Technology (NIST) statistical test suite 2.1.1. Scatter plot and *P-value* distribution graphs are also included in the manuscript to support the conclusion.

General Terms

Random number generation

Keywords

Blum Blum Shub generator, Cryptographically Secure Pseudo Random Bit Generator, RSA generator

1. INTRODUCTION

Random numbers are generated with the help of a computer program and are deterministic in nature, they are developed to be used in simulation to simulate natural phenomena and where a vast amount of random digits are needed. In the process of generating random digits a random seed of length l is obtained from a true random source, is then supplied to the algorithm to produce a longer bit stream of random bits of length $k \gg l$.

Many algorithms are designed to generate these long sequence of random bits but to predict the next generated bit in the sequence is difficult for an adversary but not impossible, so to protect the communication over internet, as SSL handshake process keys, Encryption keys are generated with these random number generators it is necessary that the generated random bits should be obtained from a cryptographically secure pseudo random number generator. Some Cryptographically secure pseudorandom number generators (CSPRNG) uses the source of entropy that are truly unbiased, fully random and unbreakable by any intruder, like Lavarand [1], Simon Cooper and Landon Curt Noll introduced the new version of Lavarand by replacing the lava lamps with another source of entropy a webcam with its lens cap on. The thermal "noise" produced by the webcam is captured it is digitized and then a Hash algorithm is applied over it, that do the mixing of numbers strip off unwanted sections of predictability and generates cryptographically secure random sequences. The new generator turn out to be more popular than the old lava lamp because it is license-free, patent-free, open source and easily available to user on the website. But in present

manuscript we emphasize on cryptographically secure pseudo random numbers generators based on way function problems.

One of the most important tasks of random number generators is key generation but its uses are not limited to cryptography. Depending upon nature of generators, random number generators are classified; if the random number generator is based on any one-way function, it is easy to compute $y=f(x)$ but very difficult to compute $x=f^{-1}(y)$, for example; Discrete log problem (Blum Micali Generator) [2], Quadratic residue problem (Blum Blum Shub) [3], Hash (FIPS -186) [4], Elliptic curve cryptography [5], Subset sum [6], Integer factorization (RSA generator) [7] it is said to be cryptographically secure pseudo random number generator (CSPRNG). A property of these RNGs is that there is no algorithm exist, which can find out next bit to be generated in the sequence given previous bits without knowledge of seed in polynomial time. If the RNG is to be used in simulation, it need not to be cryptographically secured but should have long cycle length and uniform distribution over the range of number domain.

A cryptographically secure pseudorandom bit is the bit that must be non-deterministic in nature and can be used for cryptographic purpose. To confirm this, next bit test is to be performed. Next-bit test is finding an polynomial time algorithm that can predict $(k+1)^{th}$ bit with probability greater than $1/2$, when given first k bits as input for bit generators and generates random bits that can be used for cryptographic application.

2. LITERATURE SURVEY

Computer's first requirement of random numbers was for simulations and numerical computations such as Monte Carlo calculations. It is significant to note that the requirement of randomness is different in cryptographic applications than that of simulation. Prediction is often a prerequisite when we talk about cryptographically secure pseudo random number generator.

First cryptographically secure pseudo random number generator [7], in terms of unpredictability, was introduced by Adi Shamir in 1981 who is one of the inventors of RSA [8]. Shamir uses the core concept of RSA i.e. the problem of Integer factorization to ensure the security of his new CSPRNG, with the assumption of the strength of this CSPRNG is equivalent to the security of the RSA cryptosystem because of intractability of the problem of Integer factorization. On the other hand, since it uses modular exponentiation of huge numbers, makes it slow as well as not suitable for practical applications.

Micali-Schnorr introduced a slight improved and modified version of RSA generator is more efficient than the RSA since $\left\lfloor N \left(1 - \frac{2}{e}\right) \right\rfloor$ bits are generated per exponential by e , where e is encryption key for RSA. However each exponentiation requires one modular squaring per bit.

Subsequently another Cryptographically secure PRBG is introduced Blum Blum Shub generator [3] also known as BBS generator which works on the concept of quadratic residue one way function. Its security also depends upon the intractability of the Integer factorization of modulus N . BBS generator requires only one modular squaring per bit, instead of one modular exponentiation. In contrast to RSA generator, BBS generator is also slow but suitable for practical applications like session key generation, Public key cryptosystem, nonce etc. Many other generators were also proposed as cryptographically secure PRBG but RSA and BBS generators are the most famous among all of them.

3. REQUIREMENTS OF CRYPTOGRAPHICALLY SECURE PRBG

A good pseudo random generator is needed to qualify on some standards to prove to be truly random. To verify this many statistical test are performed that are used to find whether there exist any correlation or not and the generated bit sequence have a good period length and uniform distribution over $U(0, 1)$. But unfortunately, there is no statistical test that can accomplish that if a PRNG passes all the tests of statistical test suite like National Institute of Science and Technology (NIST 2.1.1) [9] is flawless and cryptographically secure.

The requirement of normal PRBGs are satisfied by a cryptographically secure PRBG but contrarily is not true. They have to pass the statistical test suites as well as also have to prove the immunity over any adversary's attack, so that one should be unable to crack the PRBG.

Andrew Yao, the Knuth prize winner scientist mentioned in his paper in 1982 [10] that if a generator is passing the next bit test can pass all other polynomial time statistical test for randomness and should satisfy the property of unpredictability. It is necessary for a good PRBG to hold the property of unpredictability. There exist two types of unpredictability [11]:

- **Forward unpredictability**

Forward unpredictability is inability to predict next number in the sequence, with probability more than 1/2 i.e. 0.5 without the knowledge of seed, given that previous numbers are known.

- **Backward unpredictability**

In backward unpredictability seed should not be derived from the output of pseudo random number generator [12] this done because if the seed is derived then pseudo random number generator is no more secure and its randomness is compromised.

James Reeds [13] specify two notable standards of randomness in his paper "Cracking a random number generator" in 1977.

- The usual statistical standard states that a sequence of numbers, which cannot be distinguished from a sequence of true random numbers chosen from the unit interval, is considered random. Sequence should be uniformly distributed over the given space of numbers. PRNGs that

are acceptable by these standards are suitable for simulations, sampling, games, computer algorithms, medicine and similar applications.

- If a PRBG is to be used for a cryptographic application it should satisfy some key properties i.e. unpredictability is most important rather than uniform distribution and long cycle length. It is less important whether the sequence is uniformly distributed, but it is essential that the generated numbers should not contribute to adversary to find previous number or the next number going to generate.

4. BBS: THE CRYPTOGRAPHICALLY SECURE PRBG

Blum Blum Shub is a provably secure pseudo random number generator, proposed by Lenore Blum, Manuel Blum and Michael Shub in 1986 [3]. The algorithm is considered to be as secure as quadratic residue problem, an NP-complete problem. In other words, to break the BBS is equivalent to solve the quadratic residue problem, which in turn, would solve the NP-complete problem the basis of cryptography. Due to this reason the algorithm is most preferable algorithm for cryptographic purpose like key generation. This section explains the algorithm along with the quadratic residue problem.

4.1 Quadratic Residue Problem

For natural numbers a, n i.e. $a, n \in \mathbb{N}$ and $a \in \mathbb{Z}_n$, a is said to be quadratic residue n if and only if a number x with $x \in \mathbb{Z}_n$ exists and a is congruent to x squared modulo n i.e.

$$a \equiv x^2 \pmod{n} \quad (1)$$

QR_n is used to represent the set of all quadratic residues modulo n and QNR_n stands for the set of all quadratic non-residues modulo n [14]. If n is an odd prime then its QR_n would contain elements less than $n/2$.

The quadratic residue problem is to check whether a given number a is quadratic residue modulo n or not. To know this one has to find a number x for which condition given in equation (1) is satisfied, that can be done only by using brute-force technique to exhaust \mathbb{Z}_n . However, to compute its reverse i.e. finding a given numbers x, n , is quite easy to calculate, hence it is considered to be an NP-complete problem. The subsequent section illustrates how BBS uses the same problem as a one-way function.

4.2 BBS: The Algorithm

To use the concept of quadratic residue, Blum, Blum and Shub proposed to choose the two odd primes p and q , and compute $n = p \cdot q$. Then square modulo n of seed is computed and the resulting number is considered to be the first number generated. The seed is replaced with generated number in subsequent iterations and a square modulo n is computed again, generating a number per iteration. To get a bit sequence, least significant bit of generated number is extracted per iteration and is added to the generated binary sequence. Hence BBS needs only one square (or multiplication) per bit generated. Let the seed is $s, s \in \mathbb{Z}_n$, then the first number is

$$X_1 \equiv s^2 \pmod{n} \quad \text{and the bit} \quad b_1 \equiv X_1 \pmod{2}$$

For i^{th} iteration

$$X_i \equiv X_{i-1}^2 \pmod{n} \quad \text{and} \quad b_i \equiv X_i \pmod{2}$$

This way the algorithm needs to compute only one square operation to generate a bit, which is much less than any of the other cryptographically secure algorithm.

Following is the pseudo code of the algorithm:

BLUM_BLUM_SHUB (SEED):

1. $X_0 = SEED$
2. Choose two odd prime p and q
3. $n \leftarrow p \cdot q$
4. $l \leftarrow$ length of sequence
5. **for** $i \leftarrow 1$ to l
6. **do** $X_i \equiv X_{i-1}^2 \pmod n$
7. **do** $b_i \equiv X_i \pmod 2$
8. **return** $B = \langle b_1 b_2 b_3 \dots b_l \rangle$

Example

Let $p = 101, q = 97$ and seed $X_0 = 128$

$n = 101 \cdot 97 = 9797$

$X_1 \equiv 128^2 \pmod{9797}$	= 6587	$b_1 = 1$
$X_2 \equiv 6587^2 \pmod{9797}$	= 7453	$b_2 = 1$
$X_3 \equiv 7453^2 \pmod{9797}$	= 8016	$b_3 = 0$
$X_4 \equiv 8016^2 \pmod{9797}$	= 7530	$b_4 = 0$
$X_5 \equiv 7530^2 \pmod{9797}$	= 5661	$b_5 = 1$
$X_6 \equiv 5661^2 \pmod{9797}$	= 934	$b_6 = 0$
$X_7 \equiv 934^2 \pmod{9797}$	= 423	$b_7 = 1$
$X_8 \equiv 423^2 \pmod{9797}$	= 2583	$b_8 = 1$

The generate bit sequence of above example is 11001011.

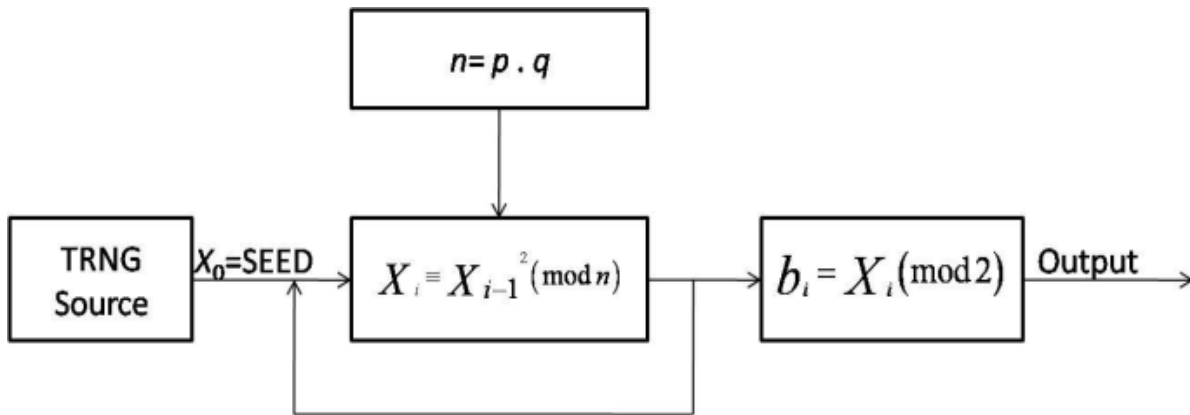


Fig 1: Schematic diagram of BBS

In this example of BBS generator, here we extract the least significant bits of the generated sequence X_i as it is mentioned in the original draft of the paper proposed by the Blum Blum and Shub in 1978. But if we extract the k least significant bits from the sequence rather than extracting one bit per squaring that is costlier and makes the functioning of generator slow, k bits per squaring increase the speed and also do not affect the security of algorithm.

5. TESTING OF BBS

The randomness of BBS is tested rigorously using statistical test suites and scatter plots. The PRBGs used for cryptographic purpose needs to be cryptographically secure and unpredictable. To be confident about randomness of number generated from a PRBG, it is important to test its output sequences. There are several of tests batteries available such as NIST statistical test suite [9], DIEHARD test [15], Donald Knuth's statistical test suite [16], and the Crypt-XS statistical test suite [17]. They all perform a number of tests to find different type of non-randomness. None of these is perfect in itself. Juan Soto [18] has shown that not all the tests are needed to be performed and the NIST statistical test suite is the best in one of these. Hence we have used NIST statistical test suite *sts-2.1.1*, regarded as most precise tests of randomness to analyze the statistical properties of BBS generator.

5.1 The NIST suite

The NIST Test Suite, consisting of 15 tests, is a statistical package that tests the randomness of (arbitrarily long) binary sequences. These sequences can be produced either by hardware or by software based random number generators.

The tests are [9]:

1. The Frequency (Monobit) Test,
2. Frequency Test within a Block,
3. The Runs Test,
4. Tests for the Longest-Run-of-Ones in a Block,
5. The Binary Matrix Rank Test,
6. The Discrete Fourier Transform (Spectral) Test,
7. The Non-overlapping Template Matching Test,
8. The Overlapping Template Matching Test,
9. Maurer's "Universal Statistical" Test,
10. The Linear Complexity Test,
11. The Serial Test,
12. The Approximate Entropy Test,

13. The Cumulative Sums (Cusums) Test,
14. The Random Excursions Test, and
15. The Random Excursions Variant Test.

Further information about these tests is not the subject of present manuscript. The test suite calculates *P-value*, the probability that a perfect random number generator would have produced a sequence less random than the sequence that was tested [9]. A significance level (α) can be chosen for the tests. If $P\text{-value} \geq \alpha$, then the null hypothesis is accepted, otherwise null hypothesis is rejected. The $\alpha = 0.01$ indicates that one would expect 1 sequence in 100 sequences to be rejected. The *P-value* of *P-values* ($P\text{-value}_T$), describes Goodness-of-fit Distribution test on the *P-values* obtained for an arbitrary statistical test (i.e., a *P-value* of the *P-values*).

5.2 Statistical Test Results of BBS

For testing of suggested algorithm, we have generated 1000 sequences, each of 10^6 bits. Each of the sequence is generated from different randomly chosen seed. The seeds are provided from a true random source - Random.org [19]. The generation of numbers has been done using C language library- GCC (GNU Compiler Collection) [20], and GMP (GNU Multiple Precision) arithmetic library [21] to handle large numbers. NIST 2.1.1 test battery is applied over each of these sequence and *P-values* for all 15 tests are computed. The significance level α is set to 0.01.

So, minimum 980 sequences must pass the test when sample size $m = 1000$ i.e. for all of the tests except random excursion and random excursion variant which have sample size of $m = 609$ and hence needs 595 sequences to pass the test for sequences to be considered random. The parameters used for testing and results of NIST suite are summarized in Table 1 and 2 respectively:

Table 1. NIST parameter List

No of sequences tested	1,000
Length of each binary sequences	1,000,000 bits
Significance level	0.01
Block size	16
Template size	9
Maximum number of templates	40

Table 2. NIST2.1.1 testing results

S. No.	Name of test	No. of sequences with P-value ≥ 0.01 (Success)	P-value of P-values	Proportion of sequences passing the test
1	Frequency test	983	0.057146	0.983
2	Block Frequency test	988	0.837781	0.988
3	Cumulative Sums test			
	1) Forward sums test	983	0.307077	0.994
	2) Reverse sums test	983	0.268917	0.983
4	Runs test	989	0.452173	0.989
5	Longest Run test	983	0.801865	0.983
6	Rank test	987	0.970302	0.987
7	FFT test	986	0.070299	0.991
8	Non-Overlapping Template matching test (Template Length = 9)			
	1. Template = 00000011	981	0.90569	0.981
	2. Template = 11000000	990	0.310049	0.99
	3. Template = 111001010	989	0.884671	0.989
	4. Template = 111001100	985	0.548314	0.985
	5. Template = 111100000	982	0.263572	0.982
	6. Template = 111101110	994	0.00087	0.994
	7. Template = 111110100	988	0.496351	0.988
	8. Template = 111011100	994	0.161703	0.994
9	Overlapping Template test	983	0.167184	0.983
10	Universal test	990	0.417219	0.99
11	Approximate Entropy test	990	0.630872	0.993
12	Random Excursions test			
	1. $x = -4$	615	0.550479	0.992

	2. $x = -3$	613	0.446255	0.989
	3. $x = -2$	613	0.540661	0.989
	4. $x = -1$	617	0.258434	0.995
	5. $x = 1$	614	0.886546	0.99
	6. $x = 2$	613	0.623687	0.987
	7. $x = 3$	615	0.913523	0.992
	8. $x = 4$	614	0.808301	0.99
13	Random Excursion Variant test			
	1. $x = -9$	600	0.103035	0.985
	2. $x = -8$	602	0.3824	0.988
	3. $x = -7$	601	0.631914	0.986
	4. $x = -6$	600	0.618038	0.985
	5. $x = -5$	600	0.821041	0.985
	6. $x = -4$	600	0.414525	0.985
	7. $x = -3$	603	0.6562	0.99
	8. $x = -2$	604	0.379555	0.991
	9. $x = -1$	604	0.855534	0.991
	10. $x = 1$	605	0.652733	0.993
	11. $x = 2$	597	0.505865	0.98
	12. $x = 3$	592	0.126536	0.972
	13. $x = 4$	597	0.272297	0.98
	14. $x = 5$	598	0.312791	0.981
	15. $x = 6$	598	0.429618	0.981
	16. $x = 7$	599	0.021627	0.983
	17. $x = 8$	602	0.011722	0.988
	18. $x = 9$	604	0.855534	0.991
14	Serial test	991	0.853049	0.991
15	Linear Complexity test	992	0.298282	0.992

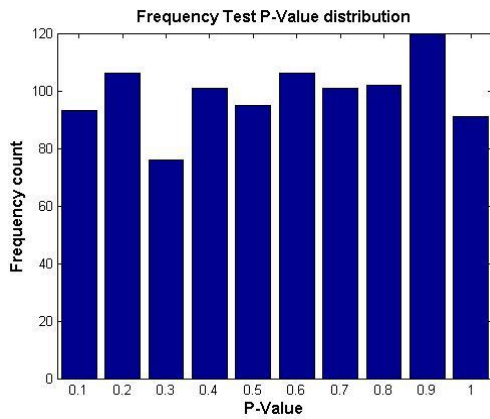


Fig 2: Frequency Test

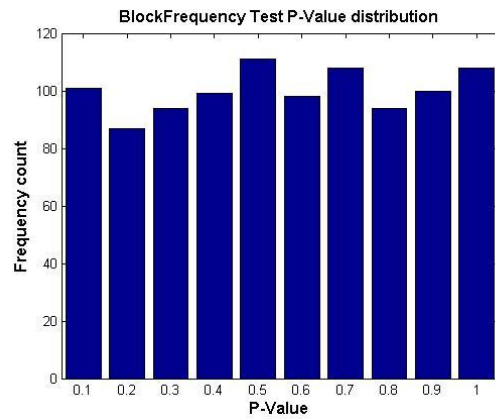


Fig 3: Block Frequency Test

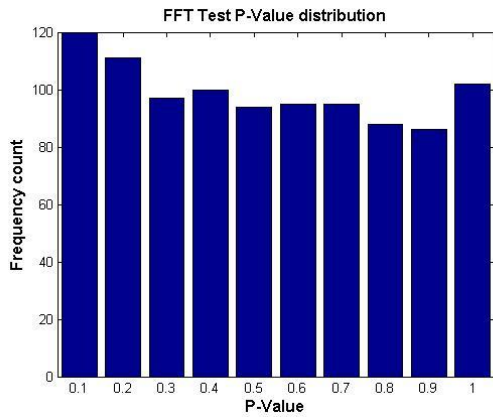


Fig 4: FFT Test

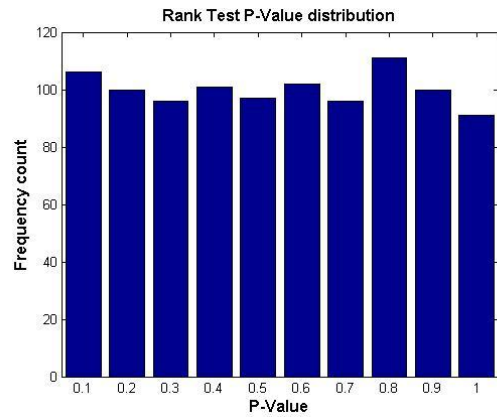


Fig 8: Rank Test

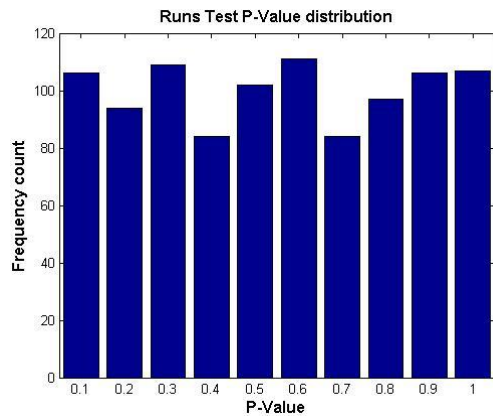


Fig 5: Runs Test

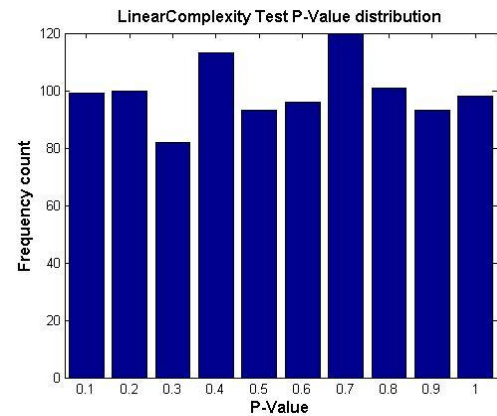


Fig 9: Linear Complexity Test

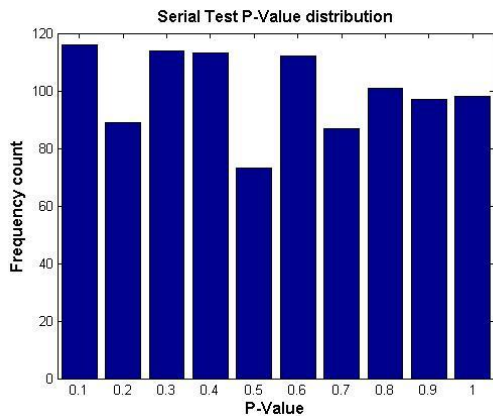


Fig 6: Serial Test

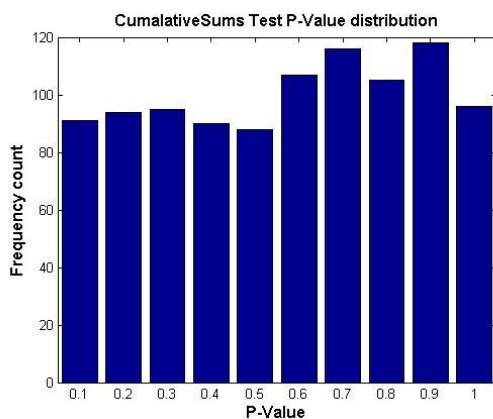


Fig 7: Cumulative Sums Test

Uniform-distribution of P -values for 1000 number of binary sequences has also been represented by histograms by dividing the complete interval of P -values $[0, 1]$ is 10 equal sub-intervals and the P -values that lie in each subinterval are plotted. For four of these tests, this distribution has been displayed in Figure 2 to 5 for four of these tests.

It is clear from the Table 2 and the Figure 2 to 5 that the P -value_T for each of these tests lies in confidence interval i.e. the tested binary sequence passes all these tests. These all figures and tables are designed using *finalAnalysisReport.txt* file generated by NIST test suite (*sts-2.1.1*), which is provided in the appendix of this manuscript. We have performed the test on some other samples also and these samples pass the tests as well.

5.3 Scatter Plot

Scatter Plots are used to show uniformity or uniform distribution of the numbers. The scatter plot of numbers generated for BBS have also been plotted in MATLAB. The motivation behind this is to show the distribution graphically rather than statistically in probabilistic terms. The figure 6 contains the scatter plot of first 1000 numbers generated by BBS with value of $p=98207$ and $q=101111$ and seed is given at random.

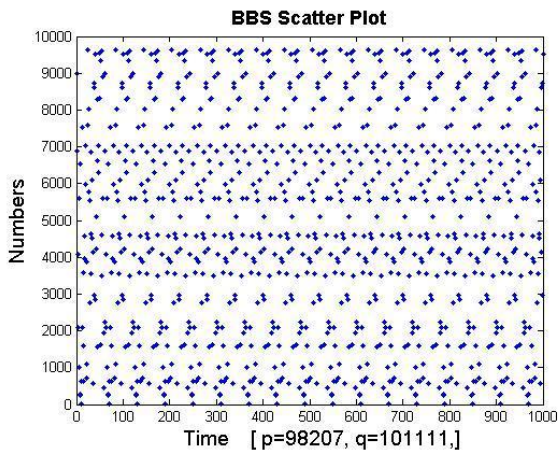


Fig 6: Scatter Plot of BBS

As the scatter plot shows, there exists a correlation between the numbers generated in subsequent iterations, but it is not undesirable as James Reed [13] proved in his paper that cryptographically secure pseudorandom number generators need not to have uniform distribution; they must be unpredictable and provably secure.

6. CONCLUSION

Since it is widely believed that no polynomial time algorithm exist that can guess the next bit, more than the probability of 0.5 generated by BBS generators and by our study we conclude that BBS, which is based on Composite quadratic residue is provably secure, is random enough to be used in cryptographic applications when computed with very large numbers. The sequence is generated in BBS by extracting least significant bits b_i from x_i , but if we extract the k least significant bits from each x_i , the security of PRBG still remain intractable as well as fast generation can also be achieved.

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APPENDIX

NIST sts-2.1.1 output file *finalAnalysisReport.txt* for Blum Blum Shub generator:

```
-----
RESULTS FOR THE UNIFORMITY OF P-VALUES AND THE PROPORTION OF PASSING SEQUENCES
-----
generator is <./data/bbs_out>
-----
```

C1	C2	C3	C4	C5	C6	C7	C8	C9	C10	P-VALUE	PROPORTION	STATISTICAL TEST
93	106	76	101	95	106	101	102	129	91	0.057146	983/1000	Frequency
101	87	94	99	111	98	108	94	100	108	0.837781	988/1000	BlockFrequency
91	94	95	90	88	107	116	105	118	96	0.307077	983/1000	CumulativeSums
95	92	95	89	84	119	105	113	108	100	0.268917	983/1000	CumulativeSums
106	94	109	84	102	111	84	97	106	107	0.452173	989/1000	Runs
108	101	105	98	110	102	105	88	88	95	0.801865	983/1000	LongestRun
106	100	96	101	97	102	96	111	100	91	0.970302	987/1000	Rank
132	111	97	100	94	95	95	88	86	102	0.070299	986/1000	FFT
98	107	95	106	106	100	86	105	103	94	0.900569	989/1000	NonOverlappingTemplate
102	83	101	88	99	117	111	101	109	89	0.310049	990/1000	NonOverlappingTemplate
109	88	99	97	106	97	111	100	94	99	0.884671	989/1000	NonOverlappingTemplate
115	100	93	101	99	100	92	94	117	89	0.548314	985/1000	NonOverlappingTemplate
97	113	120	83	110	95	92	95	94	101	0.263572	982/1000	NonOverlappingTemplate
107	111	92	98	106	94	90	91	99	112	0.703417	991/1000	NonOverlappingTemplate
71	98	89	97	86	129	89	121	115	105	0.000870	994/1000	NonOverlappingTemplate
117	116	91	100	94	91	88	106	100	97	0.408275	983/1000	NonOverlappingTemplate
100	86	94	117	91	111	101	93	99	108	0.496351	988/1000	NonOverlappingTemplate
104	98	96	87	104	106	102	92	96	115	0.773405	986/1000	NonOverlappingTemplate
107	106	97	92	98	105	101	86	112	96	0.794391	992/1000	NonOverlappingTemplate
91	100	123	93	100	95	105	94	88	111	0.342451	993/1000	NonOverlappingTemplate
98	99	98	96	111	85	110	87	98	118	0.377007	991/1000	NonOverlappingTemplate
80	123	97	96	105	99	114	98	99	89	0.161703	994/1000	NonOverlappingTemplate
103	105	95	86	93	102	113	87	93	123	0.200115	991/1000	NonOverlappingTemplate
98	106	100	98	104	107	92	89	94	112	0.856359	985/1000	NonOverlappingTemplate
116	108	104	102	84	86	85	102	118	95	0.133404	989/1000	NonOverlappingTemplate
110	98	105	100	98	90	80	94	116	109	0.345650	988/1000	NonOverlappingTemplate
105	111	105	88	94	103	91	101	84	118	0.332970	990/1000	NonOverlappingTemplate
106	101	97	110	93	90	95	102	104	102	0.944274	992/1000	NonOverlappingTemplate
104	94	98	95	112	87	98	112	100	100	0.796268	993/1000	NonOverlappingTemplate
109	106	112	101	94	85	85	101	105	102	0.556460	988/1000	NonOverlappingTemplate
98	100	113	100	106	92	95	92	111	93	0.805569	990/1000	NonOverlappingTemplate
104	90	82	105	109	117	100	109	87	97	0.279844	991/1000	NonOverlappingTemplate
88	123	103	98	91	109	98	91	92	107	0.314544	994/1000	NonOverlappingTemplate
103	99	90	111	106	101	91	94	101	104	0.910091	989/1000	NonOverlappingTemplate
97	99	99	84	106	104	94	104	104	109	0.861264	990/1000	NonOverlappingTemplate
121	107	91	98	95	98	114	95	84	97	0.282626	991/1000	NonOverlappingTemplate
115	101	96	93	94	106	105	97	82	111	0.492436	987/1000	NonOverlappingTemplate
97	101	103	96	88	94	96	89	109	127	0.235589	988/1000	NonOverlappingTemplate
93	101	88	94	99	103	106	90	101	125	0.348869	992/1000	NonOverlappingTemplate
103	122	102	105	101	104	88	100	90	85	0.344048	986/1000	NonOverlappingTemplate
89	103	89	98	102	96	100	123	100	100	0.530120	988/1000	NonOverlappingTemplate
102	91	95	103	99	115	100	100	102	93	0.912724	995/1000	NonOverlappingTemplate
92	117	112	108	98	95	102	85	94	97	0.471146	991/1000	NonOverlappingTemplate
84	99	113	99	94	111	103	99	95	103	0.711601	991/1000	NonOverlappingTemplate
97	75	105	97	97	115	105	107	99	103	0.361938	989/1000	NonOverlappingTemplate
94	111	118	97	93	85	95	101	94	112	0.375313	989/1000	NonOverlappingTemplate
98	98	92	118	109	105	86	98	104	92	0.552383	990/1000	NonOverlappingTemplate
101	104	99	99	109	86	94	96	117	95	0.676615	987/1000	NonOverlappingTemplate
87	110	79	93	84	110	114	107	100	116	0.063615	991/1000	NonOverlappingTemplate
92	91	107	96	105	94	108	103	109	95	0.875539	994/1000	NonOverlappingTemplate
91	102	104	97	95	96	105	107	97	106	0.975012	996/1000	NonOverlappingTemplate
111	90	105	89	105	98	102	94	115	91	0.593478	989/1000	NonOverlappingTemplate
92	120	107	88	107	94	90	93	99	110	0.357000	991/1000	NonOverlappingTemplate
101	101	97	81	110	102	94	107	96	111	0.639202	990/1000	NonOverlappingTemplate
129	106	100	83	92	102	104	90	100	94	0.127393	991/1000	NonOverlappingTemplate
119	85	90	97	104	86	101	109	109	100	0.296834	987/1000	NonOverlappingTemplate
101	111	113	99	100	104	94	100	101	77	0.461612	991/1000	NonOverlappingTemplate
108	97	100	99	107	82	111	93	118	85	0.233162	990/1000	NonOverlappingTemplate
106	107	93	96	116	109	82	99	90	102	0.422638	988/1000	NonOverlappingTemplate
87	88	111	105	107	94	94	99	123	92	0.228367	990/1000	NonOverlappingTemplate
84	123	107	94	90	99	122	94	97	90	0.066882	988/1000	NonOverlappingTemplate
110	114	90	86	95	106	106	87	89	117	0.177628	989/1000	NonOverlappingTemplate
103	112	117	101	86	92	101	110	88	90	0.313041	991/1000	NonOverlappingTemplate
106	94	101	90	115	95	100	104	96	99	0.870856	991/1000	NonOverlappingTemplate
77	115	117	89	105	91	101	103	119	83	0.022760	993/1000	NonOverlappingTemplate

116	99	74	99	128	96	90	101	88	109	0.014550	988/1000	NonOverlappingTemplate
110	103	117	96	103	98	86	107	90	90	0.463512	995/1000	NonOverlappingTemplate
86	117	98	113	87	104	96	113	94	92	0.257004	994/1000	NonOverlappingTemplate
104	96	92	78	108	96	92	106	118	110	0.222480	994/1000	NonOverlappingTemplate
111	98	87	106	110	93	95	95	109	96	0.713641	995/1000	NonOverlappingTemplate
100	108	110	115	96	108	95	90	95	83	0.429923	986/1000	NonOverlappingTemplate
88	109	91	98	102	107	92	102	106	105	0.841226	990/1000	NonOverlappingTemplate
99	115	90	93	117	104	100	102	94	86	0.422638	989/1000	NonOverlappingTemplate
82	101	96	99	111	96	105	98	115	97	0.593478	990/1000	NonOverlappingTemplate
89	86	97	94	93	120	89	101	108	123	0.084037	989/1000	NonOverlappingTemplate
103	103	94	122	102	100	90	93	101	92	0.579021	989/1000	NonOverlappingTemplate
119	86	84	103	91	101	97	109	106	104	0.314544	983/1000	NonOverlappingTemplate
108	106	108	89	91	89	105	103	105	96	0.777265	987/1000	NonOverlappingTemplate
102	115	74	106	102	104	91	92	109	105	0.206629	987/1000	NonOverlappingTemplate
106	91	110	77	123	93	103	103	103	91	0.111389	994/1000	NonOverlappingTemplate
90	90	93	100	108	106	110	90	104	109	0.693142	984/1000	NonOverlappingTemplate
93	104	107	91	86	104	96	105	99	115	0.664168	992/1000	NonOverlappingTemplate
98	107	95	106	107	98	87	105	103	94	0.907419	989/1000	NonOverlappingTemplate
114	97	84	90	110	105	83	101	120	96	0.125200	988/1000	NonOverlappingTemplate
101	92	85	94	110	111	109	95	118	85	0.212184	993/1000	NonOverlappingTemplate
107	83	94	107	102	108	104	110	88	97	0.574903	988/1000	NonOverlappingTemplate
107	118	102	99	88	101	88	81	117	99	0.154629	984/1000	NonOverlappingTemplate
113	94	113	104	103	95	105	97	88	88	0.589341	987/1000	NonOverlappingTemplate
103	87	91	96	92	109	98	116	99	109	0.572847	993/1000	NonOverlappingTemplate
117	101	109	77	109	98	115	78	97	99	0.048093	982/1000	NonOverlappingTemplate
108	126	86	94	91	90	98	102	98	107	0.205531	993/1000	NonOverlappingTemplate
92	102	104	98	92	107	103	96	102	104	0.981940	991/1000	NonOverlappingTemplate
91	112	108	97	111	104	106	83	100	88	0.433590	985/1000	NonOverlappingTemplate
85	102	98	104	104	112	83	111	104	97	0.490483	995/1000	NonOverlappingTemplate
95	102	99	96	103	92	103	112	100	98	0.973055	992/1000	NonOverlappingTemplate
123	92	107	120	91	86	91	97	93	100	0.103138	984/1000	NonOverlappingTemplate
92	95	98	110	96	103	106	108	102	90	0.896345	986/1000	NonOverlappingTemplate
105	115	104	105	80	120	88	107	94	82	0.058243	992/1000	NonOverlappingTemplate
90	98	98	97	109	87	96	100	113	112	0.641284	992/1000	NonOverlappingTemplate
110	109	91	92	104	88	105	120	92	89	0.278461	985/1000	NonOverlappingTemplate
105	108	94	97	114	100	87	86	103	106	0.595549	989/1000	NonOverlappingTemplate
98	105	95	92	106	99	104	106	87	108	0.883171	983/1000	NonOverlappingTemplate
100	114	97	105	97	97	112	100	88	90	0.703417	989/1000	NonOverlappingTemplate
93	97	101	119	98	103	98	79	109	103	0.377007	987/1000	NonOverlappingTemplate
100	103	92	109	98	114	98	94	91	101	0.854708	993/1000	NonOverlappingTemplate
111	103	108	98	107	103	93	98	94	85	0.769527	982/1000	NonOverlappingTemplate
109	105	104	105	98	94	94	110	93	88	0.820143	995/1000	NonOverlappingTemplate
92	104	103	95	96	105	97	99	112	97	0.956729	990/1000	NonOverlappingTemplate
96	98	94	90	89	127	114	89	104	99	0.145326	989/1000	NonOverlappingTemplate
97	101	109	107	94	110	101	73	103	105	0.319084	985/1000	NonOverlappingTemplate
101	86	97	116	106	105	97	97	98	97	0.784927	987/1000	NonOverlappingTemplate
93	82	94	114	91	101	94	95	114	122	0.112708	993/1000	NonOverlappingTemplate
101	115	106	80	93	111	104	97	107	86	0.274341	990/1000	NonOverlappingTemplate
108	101	96	115	89	90	99	104	92	106	0.695200	990/1000	NonOverlappingTemplate
104	93	94	93	100	90	106	112	101	107	0.851383	985/1000	NonOverlappingTemplate
133	108	105	102	97	100	91	98	85	81	0.028625	988/1000	NonOverlappingTemplate
94	103	101	108	90	100	91	104	115	94	0.771469	990/1000	NonOverlappingTemplate
103	112	101	109	112	103	100	80	82	98	0.264901	994/1000	NonOverlappingTemplate
100	90	101	119	113	110	96	90	82	99	0.229559	989/1000	NonOverlappingTemplate
107	112	106	92	87	109	88	85	111	103	0.317565	987/1000	NonOverlappingTemplate
111	95	102	101	98	87	111	88	108	99	0.684890	987/1000	NonOverlappingTemplate
110	103	89	95	94	115	98	97	102	97	0.796268	991/1000	NonOverlappingTemplate
102	122	98	92	100	99	87	100	89	111	0.377007	990/1000	NonOverlappingTemplate
111	87	100	90	107	111	107	91	93	103	0.587274	997/1000	NonOverlappingTemplate
105	103	114	103	94	102	100	95	98	86	0.830808	988/1000	NonOverlappingTemplate
105	104	100	108	82	92	89	101	110	109	0.538182	988/1000	NonOverlappingTemplate
107	97	87	97	115	98	94	85	104	116	0.352107	991/1000	NonOverlappingTemplate
124	119	89	95	88	92	100	100	107	86	0.081510	991/1000	NonOverlappingTemplate
104	108	116	106	90	95	113	90	84	94	0.305599	990/1000	NonOverlappingTemplate
102	91	83	98	105	99	107	113	101	101	0.715679	986/1000	NonOverlappingTemplate
111	103	99	114	100	81	119	96	76	101	0.058612	992/1000	NonOverlappingTemplate
90	111	94	91	102	102	101	100	108	101	0.903338	990/1000	NonOverlappingTemplate
97	123	106	88	102	93	96	81	111	103	0.172816	987/1000	NonOverlappingTemplate
105	106	105	93	107	79	103	96	104	102	0.668321	992/1000	NonOverlappingTemplate
122	106	80	94	86	91	98	100	118	105	0.069863	986/1000	NonOverlappingTemplate
109	107	110	98	91	106	86	99	94	100	0.755819	991/1000	NonOverlappingTemplate
111	105	115	83	109	104	99	113	76	85	0.044508	992/1000	NonOverlappingTemplate
98	93	91	103	102	89	115	98	105	106	0.781106	993/1000	NonOverlappingTemplate
86	110	109	106	108	91	86	90	91	123	0.101311	993/1000	NonOverlappingTemplate
115	103	82	84	104	102	104	105	95	106	0.404728	982/1000	NonOverlappingTemplate
106	115	81	116	97	100	87	90	105	103	0.219006	992/1000	NonOverlappingTemplate
123	108	96	89	107	109	88	95	85	100	0.184549	979/1000	* NonOverlappingTemplate

93	122	92	113	87	95	109	90	97	102	0.240501	990/1000	NonOverlappingTemplate
97	101	105	103	100	106	105	88	96	99	0.976266	990/1000	NonOverlappingTemplate
112	87	72	121	115	107	103	84	102	97	0.013102	991/1000	NonOverlappingTemplate
108	123	99	96	95	76	106	90	103	104	0.132640	990/1000	NonOverlappingTemplate
100	120	87	116	90	106	103	87	95	96	0.224821	981/1000	NonOverlappingTemplate
86	107	93	99	106	97	91	95	109	117	0.518106	994/1000	NonOverlappingTemplate
100	95	96	106	105	97	92	94	116	99	0.861264	992/1000	NonOverlappingTemplate
90	95	94	106	108	95	94	107	101	110	0.858002	993/1000	NonOverlappingTemplate
95	86	98	109	79	98	104	100	124	107	0.125200	988/1000	NonOverlappingTemplate
102	99	95	112	111	101	99	99	86	96	0.825505	986/1000	NonOverlappingTemplate
89	86	108	118	106	92	94	128	84	95	0.024688	994/1000	NonOverlappingTemplate
97	103	90	91	106	88	110	103	99	113	0.680755	996/1000	NonOverlappingTemplate
99	108	106	89	86	105	118	93	96	100	0.502247	992/1000	NonOverlappingTemplate
92	103	110	90	86	104	96	105	99	115	0.583145	992/1000	NonOverlappingTemplate
122	83	102	114	98	84	102	100	102	93	0.167184	983/1000	OverlappingTemplate
124	101	93	91	106	97	91	106	93	98	0.417219	990/1000	Universal
99	95	103	104	103	93	93	102	88	120	0.630872	990/1000	ApproximateEntropy
66	69	57	55	69	68	60	57	70	49	0.550479	615/620	RandomExcursions
65	60	57	76	63	73	59	59	59	49	0.446255	613/620	RandomExcursions
59	71	56	55	69	69	71	58	51	61	0.540661	613/620	RandomExcursions
66	57	61	75	71	52	74	55	55	54	0.258434	617/620	RandomExcursions
65	65	67	67	58	66	65	51	56	60	0.886546	614/620	RandomExcursions
67	50	69	57	63	54	59	67	72	62	0.623687	613/620	RandomExcursions
63	67	52	55	65	64	66	67	63	58	0.913523	615/620	RandomExcursions
58	67	53	63	55	61	59	65	73	66	0.808301	614/620	RandomExcursions
56	70	62	64	63	54	63	47	73	68	0.446255	612/620	RandomExcursionsVariant
70	50	58	68	56	60	66	57	69	66	0.684024	613/620	RandomExcursionsVariant
58	65	53	70	56	62	72	55	61	68	0.707249	615/620	RandomExcursionsVariant
56	69	58	69	50	63	67	55	63	70	0.637119	611/620	RandomExcursionsVariant
63	67	67	64	54	57	56	61	72	59	0.861473	610/620	RandomExcursionsVariant
62	67	68	50	61	64	61	60	64	63	0.938551	610/620	RandomExcursionsVariant
66	67	49	55	66	74	57	69	71	46	0.145858	613/620	RandomExcursionsVariant
64	60	63	65	53	62	76	60	62	55	0.777948	614/620	RandomExcursionsVariant
49	65	66	58	78	66	56	62	54	66	0.379967	612/620	RandomExcursionsVariant
57	63	52	68	55	64	66	75	66	54	0.560348	615/620	RandomExcursionsVariant
55	56	61	57	77	63	59	76	54	62	0.369073	612/620	RandomExcursionsVariant
48	71	68	58	65	67	44	59	68	72	0.144531	614/620	RandomExcursionsVariant
71	53	70	60	65	59	57	59	61	65	0.858847	617/620	RandomExcursionsVariant
64	62	62	52	67	67	60	62	59	65	0.970348	617/620	RandomExcursionsVariant
61	61	64	58	66	52	72	59	68	59	0.858847	616/620	RandomExcursionsVariant
52	71	52	61	67	78	55	61	68	55	0.258434	617/620	RandomExcursionsVariant
53	64	56	64	64	59	63	66	61	70	0.938551	618/620	RandomExcursionsVariant
55	60	64	58	60	63	61	61	72	66	0.957555	617/620	RandomExcursionsVariant
116	89	114	113	73	112	87	101	97	98	0.035406	986/1000	Serial
107	108	108	97	78	89	123	86	107	97	0.072514	988/1000	Serial
99	100	82	113	93	96	125	101	93	98	0.192724	991/1000	LinearComplexity

 The minimum pass rate for each statistical test with the exception of the random excursion (variant) test is approximately = 980 for a sample size = 1000 binary sequences.

The minimum pass rate for the random excursion (variant) test is approximately = 606 for a sample size = 620 binary sequences.

For further guidelines construct a probability table using the MAPLE program provided in the addendum section of the documentation.
