

Enhancement of Security in Cryptographic Systems by Improving Randomness

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Abstract— Randomness plays an important role in many of the applications where natural selection process needs to be simulated. The term randomness also finds an important and irreplaceable role in the area of data security and cryptography. There are various traditional random number generators. Multiplicative Congruential Generator (MCG) is one among them. The randomness of the generated sequence is enhanced by the concept of nesting introduced into the algorithm. In this work the traditional MCG is compared with Nested Multiplicative Congruential Generator (NMCG) using statistical and graphical analysis techniques. The most important advantage of NMCG is that the period of the generated sequence is infinity. In the case of MCG the period depends upon the value of multiplicand. The analysis performed also concludes that NMCG is a better random number generator when compared to MC.

Index Terms— Cryptography; Data Security; True Random Number Generator; Pseudo random number generator; Multiplicative Congruential Generator; Prime number; Kolmogorov Smirnov Test; Runs Test

I. INTRODUCTION

Random Numbers have a lot of applications where unpredictability is desired lucky draw, natural phenomena simulation, cryptography and so on. A series is said to be random if it is unpredictable, independent and uniform. Random numbers are generated by Random Number Generators. The efficiency of the generator depends on how much unpredictable, independent and uniform is the generated series.[1][2][3]

There are two types of random number generators true random number generator and pseudo random number generator. While true random number generators use real world phenomena for generation, pseudo random number generator uses computational algorithms for generation.[8] Though there are many true random number generators such as Hotbits, Laser, Random.org and so on, pseudo random number generators are used in most of the applications where randomness comes into play. [6][7][19] For better performance Pseudo Random number generators and true random number generators are used in combination also.[14][15][9] In this paper two random number generators Multiplicative Congruential Generator is compared with Nested Multiplicative Random Number Generator.[10][11] The analysis concludes that nested concept when introduced the traditional MCG has improved its performance. In nested random series the period is always infinity that is a subsequence never repeats.[12] A better random number generator promises betterment of cryptographic systems.[13]

II. MULTIPLICATIVE CONGRUENTIAL GENERATOR (MCG)

Multiplicative Congruential Generator is one of the oldest and most popular random number generator. MCG is simple and easy to implement it is a variation of traditional LCG.

Random series is generated based on a equation given below.

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

Here,

X is the sequence of random numbers

m, $0 < m$ - modulus

a, $0 < a < m$ - multiplier

$X_0, 0 \leq X_0 < m$ - seed value

MCG is fast and requires very less memory. Period of the generated series depends upon the value of 'm'. MCG is not suitable for applications like cryptography where high security is demanded because the series repeats itself after the period length.[16][17].

III. NESTED MULTIPLICATIVE CONGRUENTIAL GENERATOR (NMCG)

In Nested MCG concept of nesting is introduced into traditional MCG. The series is generated based on the equation given below. The equation is the same as that of traditional MCG. But here multiplier is not a constant value as 'a' in equation (1). 'multi' is a random number generated by another random number series.

NLCG consist of three steps:

- i. Getting the seed value
- ii. Generating the series

Getting the seed value:

Based on the current system clock value read a pixel value of the current picture captured by system camera is read. Based on the pixel value read two prime numbers $p1_0$ and $p2_0$ are generated. Here $p1_0$ is the greatest prime number less than the read pixel value and $p2_0$ is the smallest prime number greater than the read pixel value.

Seed value is given by,

$$X_0 = p1_0 * p2_0 \text{ mod } m \quad (2)$$

m is relatively prime to $p1_0 * p2_0$

Thus seed value X_0 is the product of two prime numbers. Generated seed value is cryptographically secure due to two factors, difficulty in factorizing product of two prime numbers and true randomness introduced, clock value and pixel value. The above mentioned complexity increases the efficiency of the system and makes the job of cryptanalyst difficult or rather impossible.[20]

Generating the series

$$X_i = (X_{i-1} * b_i) \text{ mod } m \quad (3)$$

$$b_i = f_i * p1_{i-1} * p2_{i-1} \quad (4)$$

f_i is the pixel value read from the image based on $p1_{i-1}$ and $p2_{i-1}$

Now based on $p1_{i-1}$ and $p2_{i-1}$ next pair of prime numbers is generated $p1_i$ and $p2_i$. Equations (3) and (4) together generate the random number series. The next element in the generated random sequence ' X_i ' not only depends on previous value X_{i-1} but also on b_i which are next element of random series generated by the equations (4). Thus another random series values contribute for final random series generation. That is one random series is nested within another series. Nesting ensures that sequence is never repeated within the final series being generated and period is infinity. NMCG algorithm is pictorially represented in the flowchart given below.

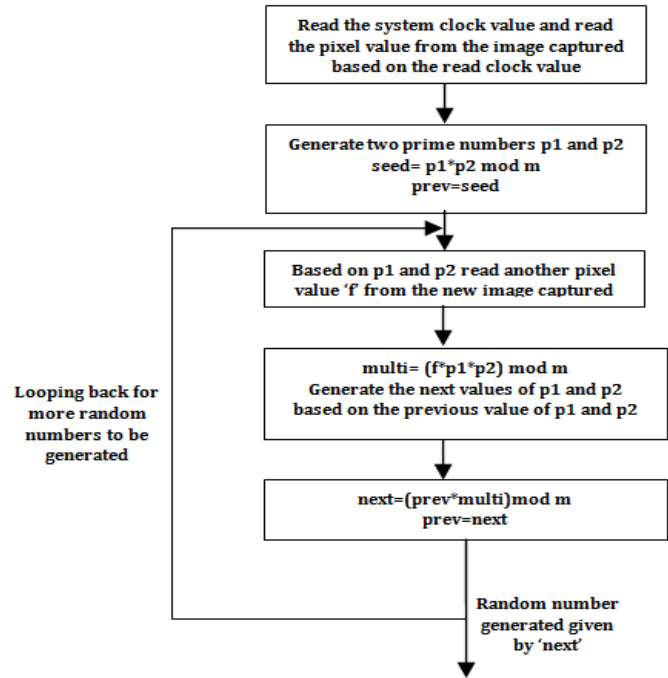


FIG. 1. Flowchart Of Nmcg

NMCG is cost effective since no expensive hardware is required. Seed value generated is sufficiently complex since true randomness and prime factorization problem is used in this step. Nested concept used during series generation makes the series unpredictable and random and subsequence never repeats in the generated sequence. [6][15] Complexities involved in this method include true entropy sources introduced in the generation, prime factorization problem, concept of nesting used in the algorithm.

IV. IMPLEMENTATION

Both mcg and nmcg were implemented in matlab and output was analyzed. System clock value was read and a pixel value was read from the current image captured based on the pixel value read. Then two prime numbers $p1_0$ and $p2_0$ was generated. Seed value is evaluated based on equation (2). The same seed value was used for mcg also.

Now the series is generated for nmcg using the equations (3) and (4) and for mcg using the equation (1). The output got for mcg and nmcg after implementing the algorithm is given below. 'm' was chosen to be 197.

Columns 1 through 15
86 107 186 82 1 84 10 96 61 93 29 114 25 108 73
Columns 16 through 30
37 183 25 129 71 172 124 153 48 89 89 86 63 169 162
Columns 31 through 45
98 126 69 6 147 96 131 114 151 131 175 101 95 9 156
Columns 46 through 60
30 173 136 101 135 95 147 68 14 45 21 146 193 61 87
Columns 61 through 75
44 80 143 171 194 87 67 111 85 114 90 37 35 176 83
Columns 76 through 90
146 40 23 81 11 119 38 1 14 54 75 155 102 146 100
Columns 91 through 105
75 175 181 157 13 190 131 45 93 11 8 123 112 2 5
Columns 106 through 120
171 14 168 61 165 142 195 148 44 101 78 19 191 97 12

FIG. 1. Output Showing The First 105 Elements In The Random Series Generated By Mcg

Columns 1 through 15
71 89 20 186 75 8 35 30 82 14 12 151 45 123 21
Columns 16 through 30
18 128 166 86 130 27 192 52 129 195 139 91 78 95 194
Columns 31 through 45
110 38 117 44 94 165 57 77 66 141 149 184 17 99 113
Columns 46 through 60
125 79 124 50 71 89 20 186 75 8 35 30 82 14 12
Columns 61 through 75
151 45 123 21 18 128 166 86 130 27 192 52 129 195 139
Columns 76 through 90
91 78 95 194 110 38 117 44 94 165 57 77 66 141 149
Columns 91 through 105
184 17 99 113 125 79 124 50 71 89 20 186 75 8 35

FIG. 2. Output showing the first 105 elements in the random series generated by nmcg

V. RESULT ANALYSIS

1) scatter diagram

Scatter diagram analysis proves that the series generated by nmcg is better when compared to mcg.

For analyzing first 30 elements are chosen. Points on the graph are divided into four quadrants. If there are x points on the graph, count x/2 points from top to bottom and draw a horizontal line. Count x/2 points from left to right and draw a vertical line. Here 30 points are considered so lines are drawn after 15 points and graph divided into four quadrants.

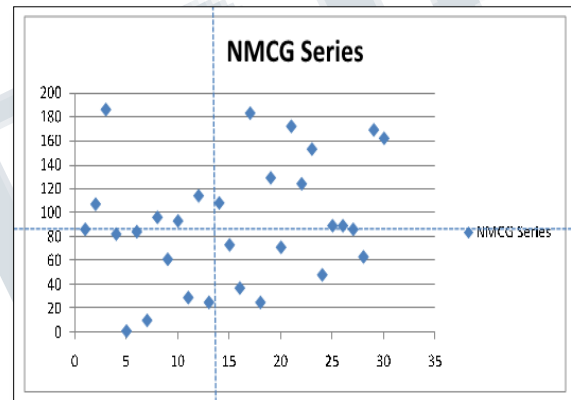


Fig. 4. Nmcg scatter diagram

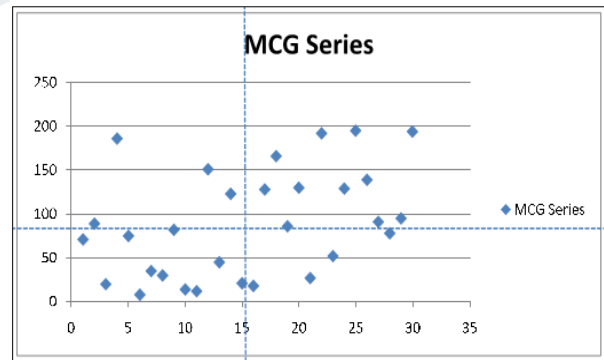


Fig. 5. Mcg scatter diagram

- Now,
- A = points in upper left + lower right (5)
- B = points in upper right + lower left (6)
- Q = smaller of a and b (7)
- N=a+b (8)

equations (5), (6), (7) and (8) is considered and the q value is compared with the limit value read from the trend test table. Limit given in the trend test table for sample of 30 is 9. Here for nmcg value of a=12 and b=18 q is 12 and is greater than the limit 9 hence the numbers in the series is drawn by random chance. But for mcg value of a=8 and b=22 q is 8 which is less than the limit value 9 hence numbers in the series are somehow related. Hence scatter diagram analysis shows that nmcg is a better random number generator than mcg.

2) bar graph analysis

the bar graph is plotted for first 200 random numbers generated both for mcg and nmcg.

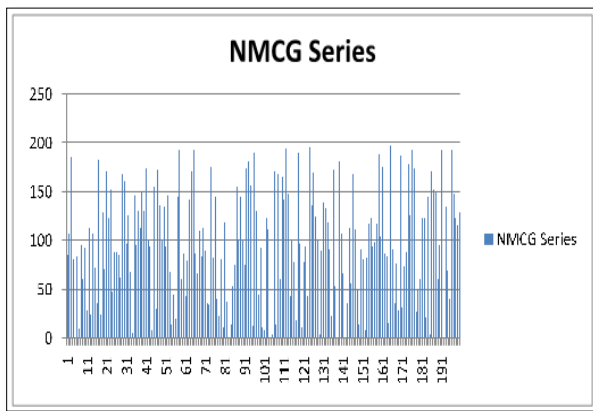


Fig. 6. Bar graph of nmcg output

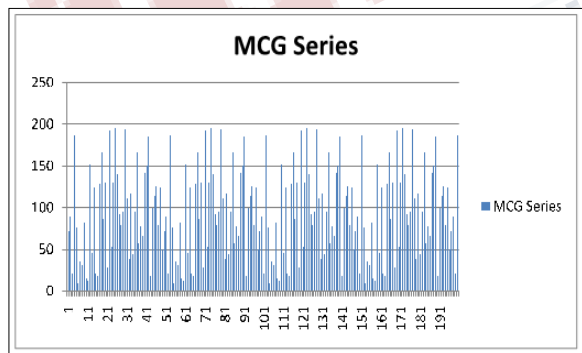


Fig. 7. Bar graph of mcg output

the plotted bar graph shows that the generated sequence is random and a subsequence never repeats for nmcg. But the case is different with mcg here the period is 49 that is subsequence repeats after every 49 numbers being. In the bar graph generated by mcg the same pattern

repeats after every 49 numbers plotted. But in the case of nmcg the subsequence never repeats. This is because the multiplicand value is never constant it forms the random number in another nested random series. Hence bar graph analysis also shows nmcg is better than mcg as a random number generator.

3) kolmogorov smirnov test

Ks test can be used to check the randomness of the numbers generated by a rng that is allowed to take on any value within a certain interval, leading to a continuous cdf [16]. Ks test is conducted on both nmcg and mcg results and the analysis table is given below.[18]

H0 = sequence being tested is random

Ha = sequence being tested is not random

Table i. Ks test analysis of nmcg

j	random	normalised	(j/n-xj)	xj-(j-1)/n
1	1	0.005076142	0.02825719	0.00507614
2	10	0.050761421	0.01590525	0.01742809
3	25	0.126903553	-0.0269036	0.06023689
4	25	0.126903553	0.00642978	0.02690355
5	29	0.147208122	0.01945854	0.01387479
6	37	0.187817259	0.01218274	0.02115059
7	48	0.243654822	-0.0103215	0.04365482
8	61	0.30964467	-0.042978	0.07631134
9	63	0.319796954	-0.019797	0.05313029
10	71	0.360406091	-0.0270728	0.06040609
11	73	0.370558376	-0.0038917	0.03722504
12	82	0.416243655	-0.0162437	0.04957699
13	84	0.426395939	0.00693739	0.02639594
14	86	0.436548223	0.03011844	0.00321489
15	86	0.436548223	0.06345178	-0.0301184
16	89	0.45177665	0.08155668	-0.0482234
17	89	0.45177665	0.11489002	-0.0815567
18	93	0.472081218	0.12791878	-0.0945854
19	96	0.487309645	0.14602369	-0.1126904
20	107	0.543147208	0.12351946	-0.0901861
21	108	0.54822335	0.15177665	-0.1184433
22	114	0.578680203	0.15465313	-0.1213198
23	124	0.629441624	0.13722504	-0.1038917
24	129	0.654822335	0.14517766	-0.1118443
25	153	0.776649746	0.05668359	-0.0233503
26	162	0.822335025	0.04433164	-0.0109983
27	169	0.85786802	0.04213198	-0.0087986
28	172	0.873096447	0.06023689	-0.0269036
29	183	0.92893401	0.03773266	-0.0043993
30	186	0.944162437	0.05583756	-0.0225042

From the table generated for nmcg output

$k+ = 0.1546531$

$k- = 0.0763113$

From ks test table at n=30 and 1-α=0.9

$k=0.21756$

$K+ < k$ and $k- < k$ hence sequence generated by nmcg is random and pass ks test

Table ii. K s test analysis of mcg

From the table generated for mcg output

$K+ = 0.1510998$

$K- = 0.0774958$

From ks test table at n=30 and 1-α=0.9

$k=0.21756$

$K+ < k$ and $k- < k$ hence sequence generated by mcg is random and pass ks test

The output analysis shows that the sequence generated by

both mcg and nmcg are random.

4) Runs Test

Run can be defined as a series of increasing values or a series of decreasing values. Length of the run is the number of increasing, or decreasing, values. For starting the runs test median of first thirty elements are found out. If a value in the series is less than median then it is denoted by -1 otherwise +1. Now runs are counted for this series of +1 and -1 and hypothesis testing is done.

Ho : sequence is random

Ha : sequence is not random

From the sequence generated by mcg 30 samples are taken and median is calculated. Median is got as 84. Now all the values greater than 84 is denoted as +1 and values less than 84 as -1. Number of runs is got as 18. Now, n1, number of -1, is 14 and n2, number of +1 is 16. From runs table the test is passed if the number of runs is between 10 and 22. Here number of runs is 18 and hence the generated sequence is not random. For nmcg median is 87.5, n1 is 15 and n2 is 15. Number of runs is 20 which is between 10 and 22, hence the generated sequence is random. Thus the runs test also proves that both nmcg and mcg are good random number generators.

VI. CONCLUSION

In this paper Nested Multiplicative Congruential Generator (NMCG) is compared with Multiplicative Congruential Generator (MCG). True random source, prime factorization problem and nesting all together contributes to the enhanced behavior of NMCG. For NMCG the period of the generated sequence is always infinity but for MCG the period depends upon the value of multiplicand chosen. Period is infinity for NMCG since the value of multiplicand is never constant. For applications that require high degree of randomness the period needs to be infinity that is a subsequence should never repeat. The presence of period makes the work of cryptanalyst easier. Statistical and Graphical analysis conducted on the generated sequence also proves NMCG to be good random number generator.

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