

1 Modeling and Simulation of Genome Evolution Using Linear
2 Boolean Functions Associated with One Dimensional Cellular
3 Automata

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6 **Abstract**

7 Structural and functional behavior of genomes could be studied using one dimensional
8 binaryvalued three neighborhood cellular automata updating rules. These updating rules are
9 linear Boolean functions, and they are applied to the adjoint sequences of adenine, (A),
10 Thymine (T), Guanine (G) and Cytosine (C) corresponding to the characteristic sequence of a
11 genome. This paper proposes the use of linear Boolean functions, and demonstrates the
12 textual or fractal behavior of genome evolution in terms of nucleotide adjoints.
13

14 *Index terms*— linear boolean functions, cellular automata, genome evolution.

15 **1 Modeling and Simulation of Genome Evolution Using Linear
16 Boolean Functions Associated with**

17 One Dimensional Cellular Automata Prashanthi Govindarajan ? , Sathya Govindarajan ? & Ethirajan
18 Govindarajan ? I. Introduction To be precise, Fig. 1 shows three levels of nucleotides. One can generate
19 64 strands of length 3. As the length increases, the number of strands he four nucleotides A, T, G, and C get
20 connected by phosphodiester bonds to form strands. Strand formation depends on innumerable factors related to
21 inter and intra cellular parameters and functions. One cannot precisely say that a particular strand gets formed
22 using such and such rules. The infinite possibilities of strand formation cannot be determined experimentally
23 or in the framework of classical genetics. One can alternatively formulate a notion of "Language of Genomes"
24 wherein one can finitely specify infinite strands, Fig. 1 shows a finitely generated quaternary tree structure of
25 strand formation of nucleic acids. T increases as per the formula $4n$, where n is the length of the strand. Strands
26 of length three are called triplet codons or 3-tuple codons. Similarly, one can think of ntuple codons where n is
27 any number.
28

29 A genome sequence is a chain of four nucleotides A, T, G and C. The numerical representation of a genome
30 sequence is a sequence of four numbers 1, 2, 3 and 4. Linear prediction of a strand could be carried out using
31 linear prediction algorithms from a sub sequence of length 8. Alternatively, one can evolve generations of genome
32 sequences from a given fulllength genome sequence using one-dimensional cellular automata rules. Section 2
33 describes the notions of adjoints of nucleotides corresponding to a genome sequence. Section 3 describes the
34 notions of cellular automata and linear Boolean functions. Section 4 provides the results of applying linear
35 Boolean functions on adjoint strings of nucleotides. Section 5 demonstrates the results of combining evolution
36 patterns of adjoint sequences dyadically. Section 6 presents various observations made from the study and
37 proposes future perspectives of cellular automatabased genome analytics.

38 Adjoint of a particular nucleotide in a genome sequence is the binary sequence obtained by substituting the
39 particular nucleotides in the genome sequence by 1's and the others by 0's. For example, let us consider a sample
40 sequence G, ??, A, T, G, A, T, T, A, C, C, A, A, G, G, C of length 16. Now the adjoint of adenine (A) is the
41 binary string $A(n) = 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0$. The adjoint of thymine (T) is the binary string $T(n)$
42 = 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0. The adjoint of guanine (G) is the binary string $G(n) = 1, 0, 0, 0, 1, 0, 0,$
43 0, 0, 0, 0, 1, 1, 0, 0. The adjoint of cytosine (C) is binary string $C(n) = 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1$.
44 The first segment of 40 nucleotides of a genome sequence of Brucella Suis 1330 is considered here for a case study.

4 VI. OBSERVATIONS AND CONCLUSIONS

45 The actual length of the genome sequence of Brucella Suis 1330 is 5806. The sample sequence is given below. $A(n) = 0110010010011000011000010000000000000000 T(n) = 000100110000000000001000001010011001100 G(n) = 100010000000011000010000010000000000000000 C(n) = 0000000001100001100010100010101100110000$

46 A cellular automaton is an idealized parallel processing system consisting of an array of numbers (1-D, 2-D
47 and more) realized using updating rules based on certain neighborhood. For example, a onedimensional cellular
48 automaton would consist of a finite length array as shown below.

51 2 III. Cellular Automata and Linear Boolean Functions

52 A cellular automaton is an idealized parallel processing system consisting of an array of numbers (1-D, 2-D and
53 more) realized using updating rules based on certain neighborhood. For example, a one dimensional cellular
54 automaton would consist of a finite length array as shown below.

55 — — — i-1 i i+1 — — —

56 Consider an ith cell in the array. This cell has a neighbor i-1 on its left and another i+1 on its right. All three
57 put together is called a three-neighborhood. One can assign a site (cell) variable ?i-1, ?i, and ?i+1 to the three-
58 neighborhood cells. At a particular instant of time, these variables take on numerical values, say either a 0 or a 1.
59 In such a case, the variables are denoted as ?ti-1, ?ti, and ?ti+1. The value of the ith cell at the next instant of
60 time is evaluated using an updating rule that involves the present values of the ith, (i-1)th and (i+1)th cells. This
61 updating rule is essentially a linear Boolean function of three variables. One can construct 256 linear Boolean
62 functions as updating rules of one-dimensional threeneighborhood binary-valued cellular automata. Each rule
63 defines an automaton by itself. So, one-dimensional binary-valued three-neighborhood cellular automata (123CA)
64 rules could be used to model adjoints of a genome sequence. The first thirty linear Boolean functions of cellular
65 automata 123CA are listed below with their decimal equivalents.

66 3 Linear Boolean Function

67 Decimal Equivalent 0 0(?? ? ??1 ?? ? ?? ? ??+1) 1 (?? ? ??1 ?? ? ?? ? ??+1) 2 (?? ? ??1 ?? ? ??)
68 3 (?? ? ??1 ?? ? ??+1) 4 (?? ? ??1 ?? ? ??+1) 5 (?? ? ??1 ?? ? ?? ? ??+1)+(?? ? ??1 ?? ? ??)
69 ?? ??+1) 6 (?? ? ??1 ?? ? ??+1)+(?? ? ??1 ?? ? ??) 7 (?? ? ??1 ?? ? ?? ? ??+1) 8 (?? ? ??1 ?? ? ??)
70 ?? ?? ? ??+1)+(?? ? ??1 ?? ? ?? ? ??+1) 9 (?? ? ??1 ?? ? ??+1)**10**(?? ? ??1 ?? ? ??)+(?? ? ??1 ?? ? ??)
71 ??+1) (?? ? ??1 ?? ? ??) (?? ? ??1 ?? ? ??+1)+(?? ? ??1 ?? ? ??) (?? ? ??1 ?? ? ??)+(?? ? ??1 ?? ? ??)
72 ??+1) (?? ? ??1 ?? ? ?? ? ??+1) (?? ? ?? ? ??+1) (?? ? ??1 ?? ? ??) (?? ? ??1 ?? ? ?? ? ??+1)+(?? ? ??1 ?? ? ??)
73 ??+1) (?? ? ?? ? ??+1) (?? ? ?? ? ??+1)+(?? ? ??1 ?? ? ??) (?? ? ??1 ?? ? ?? ? ??+1)+(?? ? ??1 ?? ? ??)
74 ??+1) (?? ? ?? ? ??+1) (?? ? ?? ? ??+1)+(?? ? ??1 ?? ? ??+1) (?? ? ??1 ?? ? ?? ? ??+1)+(?? ? ??1 ?? ? ??)
75 ??+1) (?? ? ?? ? ??+1)+(?? ? ??1 ?? ? ?? ? ??+1) (?? ? ?? ? ??+1)+(?? ? ??1 ?? ? ??+1)+(?? ? ??1 ?? ? ??)
76 ??+1) (?? ? ??1 ?? ? ?? ? ??+1)+(?? ? ??1 ?? ? ?? ? ??+1) (?? ? ??1 ?? ? ?? ? ??+1)+(?? ? ??1 ?? ? ?? ? ??+1)
77 (?? ? ?? ? ??+1) (?? ? ??1 ?? ? ?? ? ??+1)+(?? ? ??1 ?? ? ??+1) (?? ? ?? ? ?? ? ??+1)+(?? ? ??1 ?? ? ??)
78 ??+1) (?? ? ??1 ?? ? ?? ? ??+1)+(?? ? ??1 ?? ? ??) (?? ? ?? ? ??+1)+(?? ? ??1 ?? ? ??) (?? ? ??1 ?? ? ??+1)
79 ??+1)+(?? ? ??1 ?? ? ??)+(?? ? ??1 ?? ? ??+1)

80 IV. Cellular Automata Evolutions of Genome Adjoints

81 The genome sequence of Brucella Suis 1330 is considered here for a case study. Due to space limitations, a
82 part of the genome sequence and its adjoints are shown below. As defined already, adjoint of genome sequence
83 concerning a particular nucleotide is the binary string obtained by marking a '1' in the place of that particular
84 nucleotide and by marking a '0' in the places of other nucleotides. A segment consisting of 60 nucleotides of
85 Brucella Suis 1330 is shown below.

86 The adjoints of the genome sequence segment are given below.

87 Adjoint $A(n)$ Adjoint $T(n)$ Adjoint $G(n)$ Adjoint $C(n)$

88 Cellular automata evolutions of adjoints of a genome are carried out using 256 rules of 123CA. As an example,
89 rule number 137 of 123CA, that is, $(? ? ??1 ? ? ?? ? ??+1) + (? ?? ? ??+1)$ is applied to adjoints of
90 Brucella Suis 1330 genome and results shown below in Fig. ?? .Evolution of $A(n)$ Evolution of $T(n)$ Evolution of
91 $G(n)$ Evolution of $C(n)$

92 Fig. ?? : Evolution of adjoints using rule 137 of 123CA

93 The size of the images shown in Fig. ?? is 500x500, though the actual size is 5806x500. The first 500 columns
94 of the actual images are clipped and presented here for visual clarity. From Fig. ?? , it is clear that the evolution
95 pattern of each adjoint is different. One can observe that there are certain fractal patterns in the evolutions and
96 such fractals are distributed in the images very differently. For instance, the zoomed in versions of the evolution
97 patterns of $A(n)$, $T(n)$, $G(n)$ and $C(n)$ using rule 137 are shown in Figs. ?? , 4, 5 and 6 respectively.

98 4 VI. Observations and Conclusions

99 From the above empirical study, it is observed that cellular automata modeling and simulation of evolutions
100 of adjoints of a given genome sequence and the inter-pattern operations and relations exhibit distinct patterns
101 of fractals and fractal distributions. The novel technique and results presented in this paper are outcome of

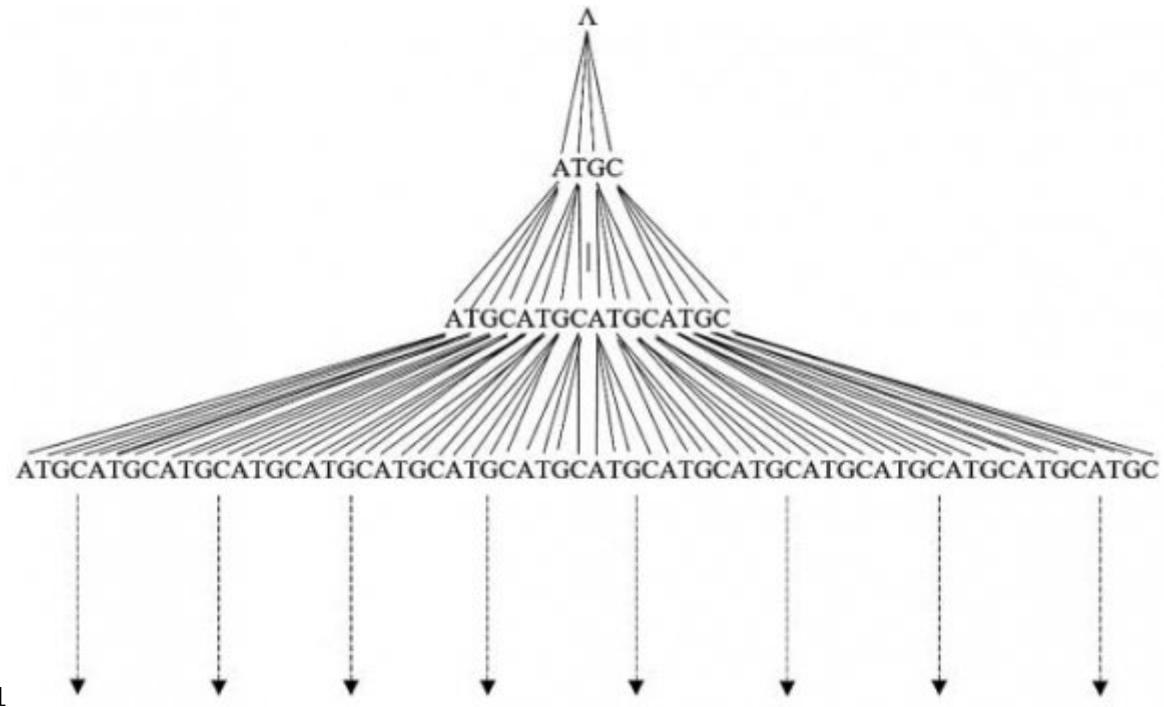


Figure 1: Fig. 1 :

Row Number	Column 1	Column 2	Column 3	Column 4	Column 5	Column 6
1.	GAATGATTAC	CAAGGCCAAG	CTCAAGCTCT	CCTTCCTTGG	GCTGAGCTTT	TGCCCTTGCA
2.	ATTCGTCATT	TTTTTCTGT	CACTGCTGAC	GAAAGACCCA	GCCAGCGGGG	CCCAAGTCGT
3.	CTGTCATGCC	GAGCCTTGAG	TCCAGATCGC	AGCAGCCCC	GATGCTCGGC	TCCCTTGCC
4.	CCGAGCGCT	TCGGTCCAGC	CTCTGACACT	CTGGCCCTTG	CTCAJCGCC	TCTCTTCCC
5.	GCCCCCTCCC	AGGGCACCTC	TGTGGGAGC	GGCCCCACGG	CCTGGGTGTTT	GTGTTGGACA
6.	GCTCGCGCAG	CGTGGGCGCA	GTGGAGTTTG	AGAAGGTGAA	GGTGTTCCTG	TCCCAGGTCA
7.	TCGAGTCCCT	GGAGCTGGGG	CCCAATGCCA	CCCGCGTGGG	CCTGGTCAAC	TACGCCAGCG
8.	CCGTGAAGCA	GGAGTTCCCG	CTGGGGGCC	ACGGCTCCAA	GGCCGGCGTG	CTGCAGGGCG
9.	TGCGCCGAT	CCAGGCACTG	TCCACGGGGA	CCATGACGGG	CCTGGGCAAC	CAGTTGGCA
10.	TCACCAAGGC	CTTCAGTGAA	GGCGAGGGCG	GTGCGCCAG	GTCCCCCGAC	ATCAGCAAGG
11.	TGCGTGGCCG	CCCTGCTGGG	TTCGGCTGTT	TGTCGCTCC	CACCTGTGCT	AAGAACTCTT
12.	GCCGGCACGC	TCTTGGITC	TCCCGCACAC	CCCCCGCATG	GCCGTTTTAC	TTCGGGGACC
13.	AGACCCAAGT	AAGAGAACGA	CGGGCTGACGC	TGGGATCGAA	CCCTCTTTA	CCCACCTTCC
14.	GACCCCAAGC	CTTCACAATGG	GGTGACGATG	ATTTCAGGGT	GGTGTACCTT	GGCTCCCCG
15.	CCCCGCTGAA	GTCTCTCGAA	CGCACAGCGA	GGGGTGTGAG	ATGCTTATGT	GATGTTTCAAGG
16.	GGCGTGGACA	CTGCCGCCGG	CCTGTCACCG	TTCAAAGCCC	GGCTCCACCA	CACCAACTGT
17.	GCCTCCCTGA	GCAGGGAGCT	CAGCTGTC	GCCTCAGTT	CCTCACCTAC	AAAATGGGAG
18.	CAACACAGCG	CCCTCTCAGA	GGGGCGCAGG	CAGGACTAAA	CGAGTTCATC	TGCTGAAGGC
19.	GCTCAGCACA	GGGCCCTCGGA	CCCAACAGGC	CCCATGGAGG	CGTTAGCTGA	GTTTGTATTT
20.	AGTACGCCCT	TGAGGGGGAG	GGGCTCAGAA	ACGCAAAGCA	ATGCCCAAA	GTCAACTGG

Figure 2:

Row Number	Column 1	Column 2	Column 3	Column 4	Column 5	Column 6
1.	0110010010	0110000010	0001100000	0000000000	0000000000	0000000000
2.	1000000000	0000000000	0100000000	0111010001	0001000000	0001000000
3.	0000000000	0100000000	0001010000	1001000000	0100000000	0000000000
4.	0001000000	0000000000	0000000100	0000000000	0001000000	0000000000
5.	0000000000	1000010000	0000000100	0000001000	0000000000	0000000101
6.	0000000000	0000000001	0000000000	1011000001	0000000000	0000000001
7.	0001000000	0001000000	0001000001	0000000000	00000000110	0100000100
8.	00000011001	0010000000	0000000000	1000000000	0000000000	0000100000
9.	0000000000	0010000000	0001000001	0010000000	0000000000	0100000001
10.	0010010000	0000010000	0000000000	0000000000	0000000000	1001001100
11.	0000000000	0000000000	0000000000	0000000000	0100000000	1101100000
12.	0000000000	0000000000	0000000000	0000000000	0000000000	0000000000
13.	1010001100	1101010001	0000000000	0000000000	0000000000	0001000000
14.	01000001000	0001010000	0000000000	1000000000	0000000000	0000000000
15.	000000000001	000000000001	000000000001	010000000000	100000000000	010000000000
16.	000000000001	000000000000	000000000000	0001100000	000000000001	0100100000
17.	000000000001	0010000000	0100000000	0000000000	0000000000	1111000000
18.	0110101000	000000000001	000000000000	0100100111	0010000100	00000011000
19.	000001000101	000000000001	0001101000	0001000000	000000000001	000000010000
20.	1001000000	001000000000	000000000000	100000000000	100000000000	000100000000

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Figure 3: Fig. 3 :Fig. 4 :Fig. 5 :Fig. 6 :Fig. 7 :

4 VI. OBSERVATIONS AND CONCLUSIONS

8910

Row Number	Column 1	Column 2	Column 3	Column 4	Column 5	Column 6
1.	0001000100	0000000000	0100000101	0011000100	0010000111	1000111000
2.	0110010011	1111110101	0001001000	0000000000	0000000000	0000001001
3.	0101010001	0000011000	1000001000	0000000001	0010010000	1001110001
4.	00000000101	1000100000	0100100001	0100000110	0100100000	1010111001
5.	00000001000	0000000010	1010000000	0000000000	0010010111	0110100000
6.	0010000000	0010000000	0100001110	0000001000	0010110010	1000000100
7.	1000010001	0000010000	0000000000	0000001000	0010010000	1000000000
8.	0001000000	00000110000	0100000000	00000010000	0000000010	0100000000
9.	1000000001	0000000010	1000000000	0001000000	0010000010	0001110000
10.	1000000000	0110001000	0000000000	0100000000	0100000000	0100000000
11.	1000100000	00001001000	1100000101	11010000100	0001101001	0000000101
12.	0000000000	1011100110	1000000000	0000000010	00001111100	1100000000
13.	0000000001	0000000000	0000100000	1000010000	0000101110	0000001100
14.	0000000000	0100000100	0010000010	0111000001	0011000011	0001000000
15.	00000010100	0010010000	0000000000	0000010000	0100110101	0010010000
16.	0000010000	0100000000	0010100000	1100000000	0001000000	0000000001
17.	0001000100	0000000011	00000000100	00010000111	0010000100	0000100000
18.	0000000000	0001010000	0000000000	0000001000	0000110010	1001000000
19.	0010000000	00000010000	0000000000	0000100000	0011000100	0111010111
20.	0010000001	1000000000	0000100000	0000000000	0100000000	0100000100

Figure 4: Fig. 8 :Fig. 9 :Fig. 10 :

Row Number	Column 1	Column 2	Column 3	Column 4	Column 5	Column 6
1.	1000100000	0001100001	0000010000	0000000011	1001010000	0100000100
2.	0000010000	0000000010	0000100100	1000100000	1000100111	00000010010
3.	0010001000	1010000101	0000100010	0100100000	1001000110	00000010000
4.	0010101000	0011000010	0000010000	0010000001	00000001000	00000000000
5.	1000000000	0111000000	0100110010	1100000110	0001010000	1001011000
6.	1000101001	0101011000	1011010001	0100110100	1101000001	00000011000
7.	0010100000	1100101111	0000001000	0001010111	0001100000	0001000101
8.	0010100100	1101000001	0010111000	0011000000	1100101001	00100011001
9.	0101001000	0001000001	0000011110	00000000111	0001100000	00100001000
10.	0000000110	00000010100	1001011101	1001000001	10000000100	0000100011
11.	0101010001	00000100111	00000010010	0010010000	00000101000	00100000000
12.	1001100010	00000011000	00000101000	00000101001	1001000000	0001111000
13.	0100000000	0010100010	0110010010	0111000100	00000000000	00000000000
14.	1000000010	00000000011	1101001001	00000001110	1100100000	11000000001
15.	00000100010	1000000100	01000001010	1011101001	0010000010	1001100011
16.	1101011000	0010010011	00010000001	00000010000	11000000000	00000010010
17.	10000000010	1001101000	00100001010	10000001000	00000000000	00000011101
18.	00000000101	00000000010	1110010011	0011000000	0101000000	0100100110
19.	1000010000	10100000110	00000000110	00000011011	0100010010	1000100000
20.	0100010000	0101111101	1110000100	0010000100	00100000000	1000000011

Figure 5:

102 prolonged research carried out in the mathematical modeling of genomes and their evolutions. It is evident that
103 one can as well look into the possibilities of genome editing using such cellular automata tools.¹

4 VI. OBSERVATIONS AND CONCLUSIONS

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