

# B103 Hierarchical Cellular Automata Model for VLSI Testing

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**Abstract:** In recent years, the Cellular Automata (CA) technology has been gaining wider acceptance in various fields - image processing, language recognition, pattern recognition, VLSI testing, study of fractals and chaos, etc. The user community look forward for higher versatility and robustness of CA based models employed to study hierarchical systems. In this context, this paper introduces a new concept of Hierarchical Cellular Automata (HCA). Theory of HCA is developed over the Galois extension field  $GF(2^{pqr})$ , where each HCA cell can store a symbol from the set  $\{0, 1, 2, \dots, 2^{pqr}-1\}$ . The potential application of HCA has been demonstrated for test solution of VLSI circuits specified in hierarchical structural description. The test solution is designed to meet the test requirements of the circuit by exploiting its structural dependencies.

## 1 Introduction

In the mid of twentieth century, many people in Artificial Life have been enamoured of a computing model known as the Cellular Automata (CA). The CA is sufficiently complex to develop an entire universe as sophisticated as the one in which we live. Von Neumann [1] envisioned the modeling of self-reproducing automata empowered to simulate the bacterial growth, the growth of patterns on seashells, fluid dynamics, and the voting patterns of individuals who made decisions based on their local neighbors.

One of the most important milestones in the history of development of the simple homogeneous structure of cellular automata is due to Wolfram [2]. He proposed one/two dimensional structure of simple cells, each having only two states with uniform three-neighborhood dependence. This simplified structure motivated a number of researchers [3] to explore innovative applications of the CA machine in various fields - image processing, language recognition, pattern recognition, testing of VLSI circuits, study of fractals and chaos, etc [3]. All these computing models employ  $GF(2)$  cellular automata where each cell is capable of storing either 0 or  $1 \in GF(2)$ .

While developing CA based models we observed that the power of  $GF(2)$  CA is inadequate to handle complex physical systems which by nature exhibits abstraction and hierarchy in a fundamental manner. For example, a set of proteins controls the growth and activities of every living organism. A protein is a linear sequence of amino acids in a polypeptide chain. An amino acid, in turn, is derived from a codon which is a triplet of nitrogenous bases (Adenine, Cytosine, Guanine, and Thymine). Thus, the bottom up

hierarchical structure of a protein starts from nitrogenous bases, followed by a sequence of codons/amino acids that leads to a *protein*. On the otherhand, while designing a VLSI circuit we fall back on top down hierarchy. At the highest level the VLSI chip is a collection of subsystems, each subsystem being designed with a network of modules. Each module in turn is a collection of submodules that realize the function of module. In general, a system when viewed at different levels of hierarchy & abstraction offers better insight into the system behavior. This establishes the need of a modeling tool to exploit the natural hierarchy of a system - observed in nature or artificially built.

In the above background scenario, we have introduced the concept of Hierarchical Cellular Automata (HCA) as a modeling tool. The theory of HCA machine is developed over the Galois extension field  $GF(2^{pqr})$ , where each HCA cell can store a symbol from the set  $\{0, 1, 2, \dots, 2^{pqr}-1\}$ . From the theoretical view point, the extension field structure of  $GF(2^{pqr})$  is isomorphic to that of  $GF(2^{pqr})$ . However, for engineering applications the hierarchical field structure of  $GF(2^{pqr})$  can be effectively employed to model the inherent hierarchy available in a physical system. The potential application of HCA has been demonstrated for an engineering application - VLSI testing.

The preliminaries on  $GF(2)$  cellular automata and the VLSI test problem are introduced in Section 2. Section 3 introduces the HCA and its characterization. The characterization is based on the theory of Galois extension field. For the sake of completeness, a review of the theory of extension field is also covered in this section. In Section

4, we present the application of *HCA* in designing the hierarchical Test Pattern Generator (*HCATPG*), customized for a *VLSI* circuit, as an effective solution to the *VLSI* testing. The efficiency of the proposed design is established through exhaustive experimentation.

## 2 Preliminaries

This section introduces  $GF(2)$  CA and the *VLSI* test problem that will be addressed in this paper.

### 2.1 $GF(2)$ Cellular Automata

The Cellular Automata (*CA*) is the simplest model of decentralized spatially extended system, made up of a number of individual components (cells). The state  $q$  of each individual unit (cell) changes over time depending on the states of its neighbors [2]. For  $q \in \{0,1\}$ , the *CA* is referred to as  $GF(2)$  *CA*. In 3-neighborhood (left, right and self) dependence, the state  $q$  of the  $i^{th}$  cell at time  $(t + 1)$  is

$$q_i^{t+1} = f(q_{i-1}^t, q_i^t, q_{i+1}^t),$$

where  $q_i^t$  denotes the state of the  $i^{th}$  cell at time  $t$  and  $f$  is the next state function called the ‘Rule’ of the automata [2].

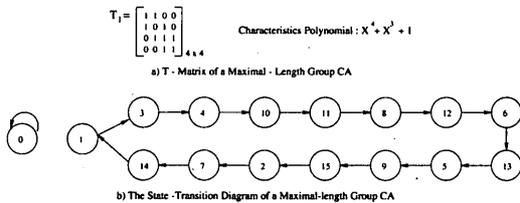


Fig. 2.1: A 4-cell maximal length group CA state transition

**Definition 1** If the next-state generating logic of CA employs only XOR then it is called a linear rule. A CA with all the cells having linear rules is called a linear CA.

An  $n$ -cell linear  $GF(2)$  CA is characterized by its characteristic matrix  $[T]_{n \times n}$  [3], where

$$T[i, j] = \begin{cases} 1, & \text{if the next state of the } i^{th} \text{ cell depends} \\ & \text{on the present state of the } j^{th} \text{ cell} \\ 0, & \text{otherwise} \end{cases}$$

The polynomial of which  $T$  is a root is called the characteristic polynomial of the CA. The characterization of  $GF(2)$  CA behavior, from its  $T$  matrix and characteristic polynomial, has been reported in [3].

**Definition 2** If all the states in the state transition graph of a CA lie in some cycles, it is called a group CA; otherwise it is a non-group CA. For a group CA,  $\det[T] \neq 0$ .

The group CA is classified as maximal and non-maximal length CA. In an  $n$ -cell maximal length CA (Fig.2.1) there is a cycle of length  $2^n - 1$  with all non-zero states.

### 2.2 The VLSI Test Problem

Rapid advances in semiconductor technology have made possible fabrication of complex *VLSI* circuits within feasible cost. However, the problem of testing *VLSI* circuits has become a major cause of concern. In recent years, the test solution incorporating *BIST* (Built-In Self Test)

methodology has been gaining increasing popularity with the test design community. Linear feedback shift registers (*LFSRs*) [4] are extensively used as the *BIST* Test Pattern Generators (*TPGs*). A wide variations of such structures have also been proposed [5, 6]. In last one decade, the simple and regular structure of Cellular Automata (*CA*) has been getting acceptance as an alternative to *LFSR* [7, 3].

*BIST* schemes are aimed to meet the basic requirements of high fault coverage with minimal test application time and low overhead. However, desired level of fault coverage for any arbitrary random logic is difficult to achieve with the conventional *BIST* structures built around *CA/LFSR*. These are typically designed without due consideration for the structure of the given *CUT* (Circuit Under Test).

The current *VLSI* chips associated with *SOC* (system on chip) application are becoming increasingly complex day by day. While designing test solution, such circuits should be viewed as a hierarchical (verilog/*VHDL*) structure. An efficient *BIST* scheme for these circuits should exploit its hierarchical structure. The Hierarchical Cellular Automata (*HCA*) can be employed as an effective tool to model the desired *BIST* structure for this class of circuits.

## 3 Hierarchical Cellular Automata

The class of CA dealt with in [3] is defined over Galois field  $GF(2)$  and can handle the elements from the set  $\{0,1\}$ . By contrast, each cell in a *HCA*, defined over Galois extension field  $GF(2^{p^q})$ , can store a symbol  $\in \{0,1,2,3, \dots, (2^{p^q} - 1)\}$ .

### 3.1 Preliminaries of Extension Field

For any positive integer  $p$ , it is possible to extend the Galois field  $GF(m)$  to a field with  $m^p$  elements, where  $m$  is a prime number. This field is known as an extension field of  $GF(m)$  and denoted as  $GF(m^p)$  [8].

There exists an element  $\alpha$  in the extension field  $GF(2^p)$  that generates all the non-zero elements  $(\alpha, \alpha^2, \dots, \alpha^{2^p-1})$  of the field. The  $\alpha$  is referred to as the generator. It is the root of an irreducible polynomial  $A(x)$ , known as the generator polynomial  $A(x)$  of extension field. The coefficients of  $A(x) \in GF(2)$ . If  $A(x)$  is a primitive polynomial, all the  $2^p$  elements of the extension field are distinct.

**Definition 3 Primitive Element:** The element  $\alpha^i \in GF(2^p)$  is called the primitive element of  $GF(2^p)$  if it is a root of any primitive polynomials in degree  $p$ .

**Representations of  $GF(2^p)$  field elements:** The element  $\alpha \in GF(2^p)$  can be represented as a vector or by a  $p \times p$  matrix  $M$  having its elements  $0, 1 \in GF(2)$ . The characteristic polynomial of  $M$  is the generator polynomial  $A(x)$ . The matrix representation of an element  $\alpha^j$  ( $j = 2, 3, \dots, (2^p - 1)$ ) is given by  $M^j$ . A column vector of  $M^j$  can be used as its vector representation [9]. The decimal

counterpart of it denotes the element of  $GF(2^p)$ . Fig.3.1 illustrates the  $GF(2^2)$  elements.

**Operations in extension field:** The *star\_table* and *plus\_table* guide the multiplication and addition operations respectively on  $GF(2^p)$  elements. These tables correspond to the generator polynomial  $A(x)$ . The *star* and *plus* tables for  $x^2 + x + 1$  are shown in Table 3.1. The first row and column represent the  $GF(2^2)$  elements in decimal notation.

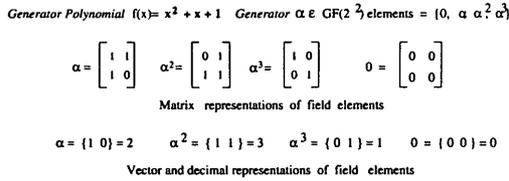


Fig. 3.1: The field elements of  $GF(2^2)$

Table 3.1: The star\_table and plus\_table for  $x^2 + x + 1$

*	0	1	2	3	+	0	1	2	3
0	0	0	0	0	0	0	1	2	3
1	0	1	2	3	1	1	0	3	2
2	0	2	3	1	2	2	3	0	1
3	0	3	1	2	3	3	2	1	0

**Extension of extension field:** The extension field  $GF(2^{pq})$  over  $GF(2)$ , where  $p$  &  $q$  positive integers, can be represented as  $GF(2^p)$  - that is, the *extension of extension field*  $GF(2^p)$  with same number of  $2^{pq}$  elements from the set  $\{0, 1, 2, \dots, (2^{pq} - 1)\}$  and isomorphic field operators *plus* & *star*. Similarly, the extension field  $GF(2^{pqr...})$  over  $GF(2)$  can be viewed as the extension field  $GF(2^{p^{qr...}})$ ;  $p, q, r, \dots$  are the positive integers. For example,  $GF(2^6)$  can be viewed as the extension of extension field  $GF(2^2)$  or  $GF(2^3)$  - that is, as  $GF(2^{2^3})$  or  $GF(2^{3^2})$ .

In  $GF(2^{pq})$  the co-efficients of the generator polynomial  $B(x) \in GF(2^p)$ . The generator  $\beta$  can be represented by a  $q \times q$  matrix with its elements  $\in GF(2^p)$  field having  $\alpha$  as its generator. The  $\alpha$  in turn, as noted earlier, can be represented by a  $p \times p$  binary matrix; in effect the elements of extension field  $GF(2^{pq})$  are hierarchically partitioned.

*Note:* From the point of extension field theory the  $GF(2^{pq})$  and  $GF(2^{p^q})$  are isomorphic. However, for the current engineering application dealing with hierarchical circuit structure, the hierarchical partitioning of field elements has distinct advantages.

### 3.2 Hierarchical CA Structure

Fig.3.2.1 shows the general structure of an  $n$ -cell hierarchical  $GF(2^{pqr...})$  CA. The interconnection among the cells are weighted in the sense that to arrive at the next state  $q_i(t+1)$  of the  $i^{th}$  cell, the present states of  $(i-1)^{th}$ ,  $i^{th}$  and  $(i+1)^{th}$  are multiplied respectively with  $w_{i-1}$ ,  $w_i$  and  $w_{i+1}$  and then added.  $w_i$ 's  $\in \{0, 1, 2, 3, \dots, (2^{pqr...}-1)\}$  are the elements of extension field  $GF(2^{pqr...})$ .

In  $GF(2)$  CA,  $p=q=r...=1$ , a cell consists of one memory

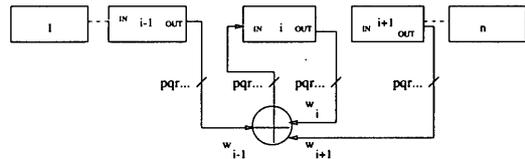


Fig. 3.2.1: General structure of an Hierarchical CA

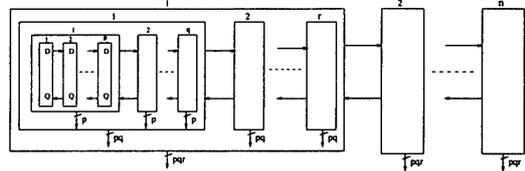


Fig. 3.2.2: Structure of a  $GF(2^{pqr...})$  Hierarchical CA cell

element ( $FF$ ) and  $w_i$ s can be either 0 (not connected) or 1 (connected). Fig.3.2.2 depicts the structure of a CA cell in three level of hierarchy -  $GF(2^{pqr...})$ . Each cell of the HCA consists of  $r$  number of subcells. A subcell in turn contains  $q$  number of next lower level subcells, each having  $p$  FFs.

As in  $GF(2)$  CA an  $n$ -cell hierarchical CA is characterized by an  $n \times n$  characteristic matrix  $T$ , where

$$T_{ij} = \begin{cases} w_{ij}, & \text{if the next state of the } i^{th} \text{ cell depends on the present state of the } j^{th} \text{ cell by a weightage } w_{ij} \in GF(2^{pqr...}) \\ 0, & \text{otherwise.} \end{cases}$$

The next state (pattern)  $X_{next}$  of a HCA can be derived as  $X_{next} = T \times X_{current}$ , where  $X_{next}$  and  $X_{current}$  are  $n$ -symbol strings. An example  $GF(2^2)$  CA with single level hierarchy follows for the sake of illustration.

generated patterns

	$GF(2^2)$	$GF(2)$
$T = \begin{bmatrix} 0 & 2 & 0 \\ 3 & 0 & 3 \\ 0 & 2 & 3 \end{bmatrix}_{3 \times 3}$	$\begin{matrix} 1 & 2 & 0 \\ 3 & 3 & 0 \\ 1 & 2 & 1 \\ 3 & 0 & 0 \\ 0 & 2 & 0 \end{matrix}$	$\begin{matrix} 01 & 10 & 00 \\ 11 & 11 & 00 \\ 01 & 10 & 01 \\ 11 & 00 & 00 \\ 00 & 10 & 00 \end{matrix}$
seed $S = \{1\ 3\ 2\}$	.....	.....

Fig. 3.2.3: The patterns generated by a 3-cell  $GF(2^2)$  CA Example 1 Let us consider the example 3-cell  $GF(2^2)$  CA of Fig.3.2.3. The  $[T]_{3 \times 3}$  matrix defines the interconnection among the CA cells. The patterns generated by the HCA with seed  $S = \{1\ 3\ 2\}$  are also shown in the figure.

### 3.3 Design of Hierarchical CA

The design of a hierarchical CA boils down to construction of  $T$  matrix of the CA. It involves two steps: (i) formation of the dependency matrix ( $D$ ) of the HCA that identifies dependencies of one CA cell on its neighbors, and (ii) specification of the weight values ( $w_i$ 's) of the dependencies. The dependency matrix of the 3-cell  $GF(2^2)$  HCA of Fig.3.2.3 is given by,  $D = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \end{pmatrix}$ , where  $D_{ij} \in GF(2) \forall i, j$ . The characteristic matrix  $T$ , as shown in Fig.3.2.3, can be generated from  $D$  by specifying the weights of the dependencies - that is, by replacing 1's of  $D$  with the non-zero weight values.

### 3.4 Characterization of HCA

Depending on the design, a HCA can show cyclic or non-cyclic behavior as in  $GF(2)$  CA.

**Theorem 1** *The hierarchical CA with characteristic matrix  $T$  is a group CA iff  $\det[T] \neq 0$ . A HCA with  $\det[T] = 0$  is a non-group CA.*

The proof is analogous to that reported in [3] for  $GF(2)$  CA.

**Theorem 2** *A HCA, having same non-zero weight in each row (column) of its  $T$  matrix, is a group CA iff  $\det[D] \neq 0$ , where  $D$  is the dependency matrix.*

*Proof:* Consider an  $n$ -cell  $GF(2^{p^{qr}})$  HCA with dependency matrix  $D$ . Assume, its  $T$  matrix is such that all the non-zero elements in each column are the same and in column  $i$  it is  $w_i$  ( $w_i \in GF(2^{p^{qr}})$ ). If  $v_1, v_2, \dots, v_n$  are the column vectors of  $T$ , then -

$\det[T] = \det[w_1v_1, w_2v_2, \dots, w_nv_n]$   
 $= w_1w_2 \dots w_n \det[v_1, v_2, \dots, v_n] = w_{n+1} \det[D]$ , where  $w_{n+1} = w_1w_2 \dots w_n \in GF(2^{p^{qr}})$  and  $w_{n+1} \neq 0$ . Therefore, the value of  $\det(T) \neq 0$ , iff  $\det[D] \neq 0$ . Hence, as per Theorem 1, the HCA is a group CA.

**Cycle structure:** For the HCA of Fig.3.2.3 having equal non-zero weights 3, 2, and 3 respectively in column 1, 2, and 3 of its  $T$  matrix, the  $\det[D] = 1$ . It is a group CA. In general, the cycle structure of a group CA is defined as  $[\mu_1(k_1), \mu_2(k_2), \dots]$  - that is, it has  $\mu_i$  number of cycles of length  $k_i, \forall i$ . The cycle structure of the CA of Fig.3.2.3 is  $[1(1), 1(3), 4(15)]$  - it has 1 cycle of length 1, one of length 3, and 4 cycles of length 15.

**Property 1:** A HCA designed with primitive weight while satisfying Theorem 2 generates larger length cycles.

This has been illustrated in Table 3.4. The first column shows the number of cells ( $n$ ) and the extension field parameter ( $p$ ) of different  $GF(2^p)$  CA. The HCA (with one level of hierarchy) for a particular  $n$  &  $p$  are designed for a large number of trials with primitive & non-primitive weight sets. Cycle lengths produced by the HCA for both the cases are noted in Column 2 and 3. For a given pair of  $n$  and  $p$ , an entry  $l$  in Column 2 [Column 3] reports that out of 100 randomly generated group HCA at least 25 of such HCA produce cycles of length  $\geq l$ .

Table 3.4: Cycle structure: primitive vs non-primitive weights

HCA (n, p)	cyl len ( $l$ ) with prob. of occurrence 25%	
	primitive weight	non-prim. weight
5, 8	331672625	65535
4, 8	131070	43690
6, 6	524286	58254
5, 6	17039295	298935
3, 6	262143	262143
8, 4	131070	8190
7, 4	1052415	65535
3, 4	4095	255

**Theorem 3** *A  $GF(2^{p^{qr}})$  HCA designed with same non-zero weight does not produce maximal length cycle for*

$(pqr.) > 1$ .

*Proof:* Without loss of generality let us assume that,  $T_{eq}$  be the characteristic matrix of a  $GF(2^p)$  HCA designed with same non-zero weight  $w$  and  $p > 1$ .

Case I:  $\det[T_{eq}] = 0$ : The HCA is a non group CA (Theorem 1) and thus can't generate maximal length cycle.

Case II:  $\det[T_{eq}] \neq 0$ : The HCA is a group CA (Theorem 1). If it is a maximal length CA, then

$$(T_{eq})^{2^{np}-1} = I \tag{1}$$

By elementary row and column operations it can be written as  $T_{eq} = wD$ , where  $D$  is the dependency matrix of the HCA having non-zero elements as 1.

Since,  $\det[T_{eq}] \neq 0, \det[wD] \neq 0$

that is,  $w \det[D] \neq 0 \Rightarrow \det[D] \neq 0$ .

Therefore,  $(D)^{2^n-1} = I$ , where  $D$  is a matrix in  $GF(2)$ .

From equation 1,  $[wD]^{2^{np}-1} = I$  i.e.,  $w^{2^{np}-1} \times (D^{2^{np}-1}) = I$ .

As  $w^{(2^p-1)} = 1$  ( $w \in GF(2^p)$  & '1' is the identity element in  $GF(2^p)$ ) and  $(D)^{2^n-1} = I$ , then

$[wD]^k = I$ , where  $k = \text{lcm}(2^p - 1, 2^n - 1)$ . It implies,  $[T_{eq}]^k = I$ . But, as  $k < (2^{np} - 1)$ ,  $T_{eq}$  can't generate maximal length cycle of length  $(2^{np} - 1)$ .

**Theorem 4** *If an  $n$ -cell HCA with characteristic matrix  $T$  has cycle structure  $[1(1), \mu_1(m)]$ , where  $\mu_1 m = (2^{npqr} - 1)$ , and  $p, q, r, \dots$  are the extension field parameters, then the HCA designed with  $[T]^\mu$  has cycles  $[1(1), \mu_1 \mu(k)]$ , where  $\mu k = m$ .*

*Proof:* From the cycle structure of  $T$  it is obvious that for any non-zero state  $x$ ,  $[T]^m x = x$ , that is  $[T]^m = I$ . Now, if  $\mu k = m$ , then  $[T]^{\mu k} = [T]^\mu k = I$ , which means  $[T]^\mu$  may have cycle of length  $k$  or factors of  $k$ . Let us assume that  $[T]^\mu$  has cycle of length  $k_1$ , where  $k_1$  is a factor of  $k$  and  $k_1 k_2 = k$ .

Then for some state  $x_1$ ,  $[T]^\mu k_1 x_1 = x_1$ , that is

$$[T]^\mu k/k_2 x_1 = [T]^\mu k_1/k_2 x_1 = [T]^\mu 1/k_2 x_1 = x_1$$

But,  $[T]^m x = x, \forall x$ . Therefore,  $k_2 = 1$

Hence,  $[T]^\mu$  has the cycle structure as  $[1(1), \mu_1 \mu(k)]$  since  $T$  has the cycle structure  $[1(1), \mu_1(m)]$ .

**Lemma 1:** If  $\beta$  is a primitive element in the extension field then  $\beta^2, \beta^4, \beta^8, \dots$  are also primitive

*Proof:* Without loss of generality let us assume that  $\beta$  is a primitive element in  $GF(2^{p^q})$ ,  $p$  and  $q$  being the extension field parameters.  $\beta$  is primitive - that is, the  $\beta, \beta^2, \beta^3, \dots, \beta^{(2^{pq}-1)}$  all are unique and the cycle structure of  $\beta$  is  $[1(1), 1(m)]$ , where  $m = (2^{pq} - 1)$ . From Theorem 4 it follows that  $\beta^\mu$  has the cycle structure  $[1(1), \mu(k)]$ , where  $\mu k = m$ , as  $\mu_1 = 1$ . Since,  $m$  is odd, then  $2^i$  will be prime to  $m$  for  $i = 1, 2, 3, \dots, (pq-1)$ . Hence,  $\beta^{2^i}$  has the cycle structure as  $[1(1), 1(m)]$  - that is,  $\beta^{2^i}$  is primitive.

### 3.5 Randomness Property of HCA

The randomness property of the HCA based pattern generator, formed for different values of  $p, q$ , and  $n$ , are studied based on the metric proposed in DiehardC [10].

A comparative study on randomness quality of maximal length  $GF(2)$  CA,  $GF(2^p)$  CA and  $GF(2^{p^q})$  CA is presented in Table 3.5.

Table 3.5: Randomness Test

Randomness Test	Pattern size $N = 32$		
	$GF(2)$	$GF(2^4)$	$GF(2^{2^4})$
Overlap Sum	pass	pass	pass
Run	pass	pass	pass
3Dsphere	fail	pass	pass
Parking lot	fail	fail	pass
B'day spacing	fail	pass	pass
Count 1's	fail	pass	pass
B rank(6x8)	fail	pass	pass
B rank(31x31)	fail	fail	pass
B rank(32x32)	fail	fail	pass
Count 1's(byte)	fail	fail	pass
Bit stream	fail	pass	pass
Craps	pass	pass	pass
Minimum Dist	fail	pass	pass
Overlap 5-permut	pass	fail	pass
OPSO	fail	fail	pass
OQSO	fail	fail	pass
DNA	pass	pass	pass
Squeeze	fail	pass	pass

The 18 different tests are shown in Column 1 of the table. Each test produces a set of 'p' values. For a pattern set with good randomness quality, the values of p's will be uniformly distributed between 0.001 and 0.999. The entry *pass*, in a column, means that the 'p' value is evenly distributed on [0,1] for at least 75% cases of total number of seeds tried for a pattern generator. The results of Table 3.5 report that the randomness quality of the patterns generated by the HCA is better than that of  $GF(2)$  CA.

#### 4 Design of Customized HCATPG

The proposed scheme extracts the clustering of primary inputs (PIs) and the structural dependencies among the different hierarchical modules of the CUT. Consider the example circuit of Fig.4 with 36 primary inputs (PIs). To design a conventional pseudo-random TPG for this circuit we can use a 36-bit length LFSR/ $GF(2)$  CA. In practice, for all types of CUT with 36 PIs, the same pseudo-random pattern generator or its variations are used as the TPG. It can be seen that the PIs of Fig.4 get grouped into four 9-bit buses (A, B, C, and E). It is logical to assume that all the 36 PIs of the CUT are not independent so far as their functionality is concerned. The 9 PIs of type A input to module  $M_1$  are functionally similar and can be considered to form a cluster of PIs rather than 9 independent PI lines - each carrying a single bit. The similar consideration is valid for B, C and E. The guiding motivation is - instead of feeding the PIs of a cluster of 9-bit bus from 9 cells of a 36-cell  $GF(2)$  CA, we propose to feed the cluster from a cell of the 4-cell  $GF(2^9)$  CA.

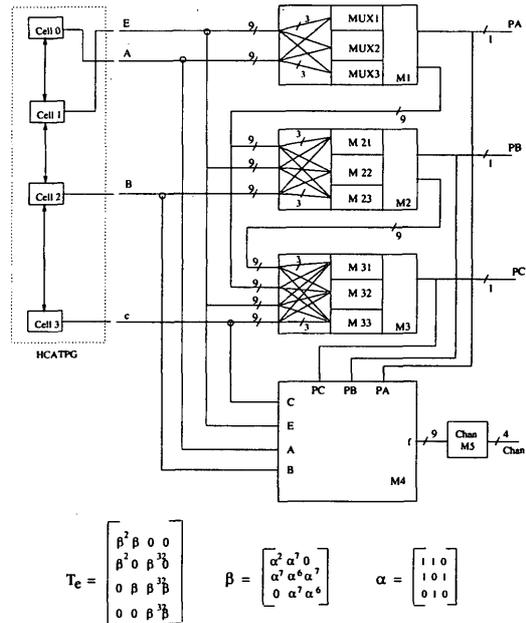


Fig. 4: High-level model for a CUT & the HCATPG

Next, on further analysis of the module  $M_1$ , it can be observed that the 9-members of cluster A [E] are divided in three groups (sub-clusters), each group containing 3 PIs and fed into three different sub-modules ( $MUX_1$ ,  $MUX_2$ , and  $MUX_3$ ). Similar cases can be observed for other blocks ( $M_2$ ,  $M_3$ ,...). It is very much logical to assume that within a cluster, the intra sub-cluster PIs are functionally more closer in comparison to the inter sub-cluster PIs. The functional dependencies of the PIs in a sub-cluster are to be reflected in designing the TPG cells. This problem can be solved if the  $GF(2^9)$  CA cell is viewed hierarchically as a  $GF(2^{3^3})$  CA cell. Without any loss of generality, we shall restrict to two level of hierarchy for designing HCATPG.

The major design steps to arrive at the desired customized HCATPG are as follows:

- (1) Selection of  $p$ ,  $q$ , and  $n$  for  $GF(2^{p^q})$  HCATPG.
- (2) Identifying the dependency matrix  $D$ .
- (3) Designing the characteristic matrix  $T$ .

##### 4.1 Selection of $p$ , $q$ and $n$

This step executes partitioning of primary inputs to form input clusters & sub-clusters based on their functional information and then identifies the cardinalities of the clusters & sub-clusters. The most frequent cardinality of the clusters ( $c_1, c_2, \dots, c_k$ ) is chosen as the value of  $(p \times q)$ , whereas the most frequent cardinality of sub-clusters within a cluster is taken as the value of the parameter  $p$ . The value of  $n$  (number of TPG cells) is fixed as per the expression

$$n = \lceil (|c_1|/pq) \rceil + \lceil (|c_2|/pq) \rceil + \dots + \lceil (|c_k|/pq) \rceil + \lceil (|restPIs|/pq) \rceil$$

For testing the circuit of Fig.4, the HCATPG is a 4-cell  $GF(2^{3^3})$  CA. The cells are marked as 0, 1, 2, and 3.

## 4.2 Identification of Dependency matrix

If two *PI* clusters enter into the same circuit module, the structural dependencies is said to exist between them and these are referred to as *dependent clusters*. It is observed that the dependent clusters closely interact among themselves to detect the faults of the circuit module. Therefore, the *CA* cells feeding the *dependent clusters* must have dependencies among themselves. For Fig.4, the resulting *D* matrix is

$$D_e = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 \end{bmatrix}$$

where *PI* clusters *A*, *E*, *B*, and *C* are respectively fed from the Cells 0, 1, 2, and 3.

## 4.3 Design of Characteristic Matrix *T*

The non-zero value '1' of the *D* matrix is next replaced with a weight value ( $\beta^j$ ) from the set  $W = \{\beta, \beta^2, \beta^4, \dots\}$ , to design the *T* matrix of the *HCA*. The basic algorithmic steps to arrive at the characteristic matrix *T* of the *HCATPG* are noted in [9]. In the design, to customize the *HCATPG*, the weights of *T* are tuned depending on the test requirement of the *CUT*. The tuning is performed in two steps: the *major* and *minor* tuning.

**Major tuning:** This step involves the fixing of the weight value ( $\beta^j$ ) for a *HCATPG* cell. For example, to design the *T* matrix of the *HCATPG* for the *CUT* of Fig.4, the 1's of the *D<sub>e</sub>* are replaced by the weights from the set  $\{\beta, \beta^2, \beta^4, \dots\}$ . The *T<sub>e</sub>* noted in Fig.4 results in maximum fault coverage for the *CUT* and the weight values  $\beta^2$ ,  $\beta$ ,  $\beta^{32}$ , and  $\beta$  are fixed for the *HCATPG* cell 0, 1, 2, and 3 respectively.

**Minor tuning:** The  $\beta_{matrix}^j$  is a  $q \times q$  matrix having its elements from the set  $\{0, \alpha, \alpha^2, \dots, \alpha^{(2^q-1)}\} \in GF(2^q)$ . The minor tuning of *TPG* is done by structural modifications of the weights fixed through major tuning - that is,  $\beta^j$  is fine tuned by changing the relative positions of  $\alpha^i$ 's within the  $\beta^j$  to improve the fault efficiency of the design. The final structure of the weight  $\beta$  for the example design is shown in Fig.4.

## 4.4 Experimental Results

The basic requirement for the proposed customized on-chip test pattern generator is the availability of the hierarchical structural description of the *CUT*. The circuits specified in Column 1 of Table 4.4 are designed from *ISCAS & ITC* benchmark circuits to get the hierarchical structural net lists. Table 4.4 depicts the summary of the fault coverage shown by *GF(2)* *CA* and *GF(2<sup>q</sup>)* *HCA* based designs in columns 4 and 5 respectively. A *CUT* is tested for three seeds (taken randomly) for both the designs. The fault coverage figures for a *CUT* are achieved on applying a fixed number of test vectors mentioned under the column heading *Test Vec.* Detail experimental setup is noted in [9]. It is observed that

the *HCATPG* resulted in higher fault coverage than that could be achieved with *GF(2)* *CA* based *TPG*.

Table 4.4: Test results of customized *HCATPG*

Circuit name	# PI/PO	Test Vec.	Fault coverage (%)	
			<i>HCA</i>	<i>GF(2)</i>
c1	36/7	400	99.18	98.57
c2	32/32	80	99.56	99.48
c3	41/32	450	98.95	98.29
c4	33/25	3000	99.12	98.76
c5	67/48	600	99.34	99.34
c6	82/64	700	98.95	98.81
c7	72/14	400	99.08	98.47
s1	24/4	2500	99.28	98.75
s2	23/24	1400	97.76	97.20

c represents combinational, s represents sequential

## 5 Conclusion

The paper presents an innovative concept of hierarchical cellular automata. The theory of *extension field* is utilized in designing the hierarchical *CA* (*HCA*). The *HCA* can be employed to model inherent hierarchy within a physical system. The efficiency of the *HCA* machine is tested in an engineering application - *VLSI* testing.

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