Fundamenta Informaticae XXI (2001) 1001–1021 IOS Press

Theory of Additive Cellular Automata

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Abstract. This paper reports the complete characterization of additive cellular automata (ACA) that employ *xor* and *xnor* logic as the next state function. Compared to linear cellular automata (LCA) [3], which employs only *xor* logic in its next state function, an *ACA* display much wider varieties of state transition behavior and enhanced computing power. An analytical framework is developed to characterize the cyclic vector subspaces generated by an *ACA* with reference to *LCA*. It identifies the conditions on which the state transition behavior of an *ACA* differs from that of the corresponding *LCA* and also provides the theoretical analysis of the nature of difference.

I. Introduction

This work develops the theory of additive cellular automata (ACA). The theoretical framework proposed, provides the complete characterization of the cyclic state space generated by an ACA. An ACA is additive in the sense that it employs affine (also referred to as additive) transformation rather than a linear transformation implemented in a typical linear cellular automata (LCA). The theory of LCA provides the foundation of the proposed characterization of ACA.

ACA which employ xor and xnor logic to generate its next state function has been specially popular among researchers. Both ACA and its subset LCA (which employ only xor logic) have been used to develop a lot of applications in VLSI and related fields. They have been used to develop pseudo-random

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test pattern generators [18, 19], signature analyzers [6], finite state machines FSM [1], error correcting codes [5] etc. Moreover, researchers have designed CA based cipher system[15], message authenticators [9], CA based pattern classifier [11] with the help of ACA. In the process of developing the applications, there has been several works to characterize the state transition behavior of both LCA and ACA.

The analysis of linear CA, has been extensively investigated by Stone [21] as part of the exploration of linear machine. Vector space theoretic analysis of state transition behavior of this class of CA has been reported by Das [8] and subsequently by a number of researchers [3, 6, 11]. The partial characterization of ACA, that employs both the *xor* and *xnor* logic in its next state function, has been introduced in [3, 15]. However, complete characterization of the cyclic state space generated by such a CA remains untouched.

In this background, this paper reports a complete vector space theoretic analysis of ACA. We develop an elegant solution to derive its cycle structure from analysis of the given rules, defining the ACA. An ACA can have both the cyclic and non-cyclic state space. However, in the current work, we only consider the characterization of cyclic state space as the non-cyclic subspace generated by an ACA is isomorphic to that of LCA [3].

We next provide a brief introduction to additive CA along with some important results on linear CA that are relevant for the characterization of ACA in section III. The vector space theoretic analysis targeting complete characterization of ACA, is reported in section III.

II. Cellular Automata Characterization

Cellular Automata (CA) consist of a number of interconnected cells arranged spatially in a regular manner. In most general case, a CA cell can exhibit s different states and the next state of each cell depends on the present states of its k neighbors including itself. Such a CA is called an s-state k-neighborhood CA. However, Wolfram [12] worked with several features of finite CA known as 3-neighborhood (left, right and self) CA having 2 states for each cell. The state (next state) $q \in \{0,1\}$ of the i^{th} cell at time (t+1) is denoted as

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

where q_i^t denotes the state of the *i*th cell at time *t* and f_i is the next state function called the rule of the automata [22]. Since *f* is a function of 3 variables, there are 2^{2^3} or 256 possible next state functions. The structure of a 3-neighborhood *CA* cell is shown in *Fig. 1*.

Out of total 256 CA rules, 14 rules that can be realized by xor/xnor logic are called additive rules [3]. A CA designed with such rules are called additive CA (ACA). The ACA has been of special interest to researchers, as it can be characterized by matrix algebraic tools. Matrix algebraic tools are used to represent ACA that uses different rules in different cells. In the current work, we concentrate on characterizing such hybrid CA. A brief overview of this model is next outlined [3].

An *n*-cell 1-dimensional ACA is characterized by a linear operator $[T]_{n \times n}$ matrix and an *n*-dimensional inversion vector F. T is the *characteristic* matrix of the cellular automata. The i^{th} row of T corresponds to the neighborhood relation of the i^{th} cell, where

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

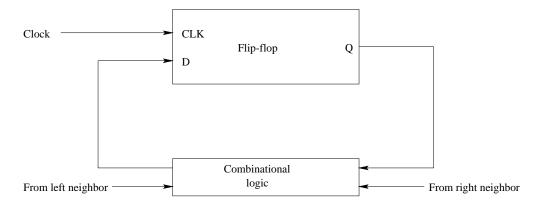


Figure 1. A 3-neighborhood CA cell

Since the CA is restricted to 3-neighborhood dependency, T[i, j] can have non-zero values for j = (i - 1), i, (i + 1). The inversion vector F of an ACA is defined as

$$F_i = \begin{cases} 1, & \text{if the next state of the } i^{th} \text{ cell results from inversion } (xnor) \\ 0, & \text{otherwise } (xor) \end{cases}$$

If s_t represents the state of the CA at the t^{th} instant of time, then the next state - that is, the state at the $(t+1)^{th}$ time instant, is given by :

$$s_{(t+1)} = T \cdot s_t + F.$$
 Therefore, $s_{(t+p)} = T^p \cdot s_t + (I + T + T^2 + \dots + T^{p-1})F,$ (1)

where $s_{(t+p)}$ is the state of CA at $(t+p)^{th}$ instant of time. For an *n*-cell CA, F is the *n* bit *inversion* vector with its i^{th} $(0 \le i \le n-1)$ bit as 1, if *xnor* rule is applied on the i^{th} cell; whereas 0 implies xor (linear) rules. The operators (., +) follow the rules defined in binary arithmetic for multiplication and addition respectively.

As the LCA is a special case of ACA, where the inversion vector F is an all 0s vector, the next state function (eq. (1)) for the LCA gets simplified to

$$s_{t+1} = T \cdot s_t \Rightarrow s_{t+p} = T^p \cdot s_t \tag{2}$$

The state transition eqs. (1) and (2) results in some global state transition behavior of the CA on the basis of which we can classify CA^1 into two categories - group and non-group CA.

II.1. Group and non-group CA

A CA that generates only cyclic states during its state transitions is known as group CA, whereas a CA generating both cyclic and non-cyclic subspaces is the non-group CA. The state transition diagram of an ACA can be characterized from its T matrix and the inversion vector F. However, the characteristic matrix (T) can directly determine whether the CA is a group or non-group CA -

$$if det(T) = 1$$
, the CA is a group CA
= 0, the CA is a non group CA

¹Henceforth, unless otherwise mentioned, the term ACA and CA are synonymously used.

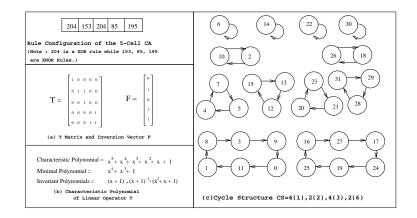


Figure 2. A 5-cell group CA and its cycle structure [4(1),2(2),4(3),2(6)] Note : The LCA formed from the T matrix also have the same cycle structure CS = [4(1),2(2),4(3),2(6)]

Group CA: For a group CA (*Fig.* 2), each CA state has a unique predecessor. That is, all the states lie on a disjoint set of cycles. The state transition behavior of a group CA is represented by the cycle structure (CS) = $[\mu_{k_1}(k_1), \mu_{k_2}(k_2), \dots, \mu_{k_m}(k_m)]$, where k_i is the cycle length of the i^{th} cyclic component of CS and μ_{k_i} is the number of such components. *Fig.* 2(c) illustrates the cycle structure of the 5-cell group CA. It has 4 cycles of length 1, 2 cycles of length 2, 3 cycles of length 3 and 2 cycles of length 6. The complete cycle structure is denoted as CS = [4(1), 2(2), 4(3), 2(6)].

Problem Definition :- tackled in this paper. In this work, we analytically compute the $CS = [\mu_{k_1}(k_1), \mu_{k_2}(k_2), \dots \dots \dots \mu_{k_m}(k_m)]$ from a given additive CA - that is from a T matrix and the F vector. Computation of CS for a linear CA has been widely studied [3, 6, 7, 20]. We consider those studies as the base while formulating our proposed scheme.

The reported analysis show that the cycle structure of LCA and ACA follows some definite relation and can be divided into two distinct categories - (1) The cycle structure repeated and not the sequence of LCA and ACA are identical (*Fig.* 2); (2) The cycle structure of LCA and ACA are different (*Fig.* 3). Details of these two aspects are reported in *section III*.

Scheme to analyze cycle structure for LCA has been developed with the underlying concept of characteristic polynomial, minimal polynomial and invariant polynomials.

Characteristic polynomial : The characteristic polynomial f(x) of a CA is det(T + Ix), where T is the characteristic matrix.

Minimal polynomial: The minimal polynomial is the minimum degree polynomial which annihilates T. *Invariant polynomials (IP):* The characteristic polynomial f(x) comprises of polynomials $\phi_i(x)^{n_i}$ invariant to the linear operator T, where $\phi_i(x)$ is irreducible.

The example CA of Fig. 2 illustrates T, its characteristic polynomial, minimal polynomial and the invariant polynomials.

If the characteristic polynomial f(x) of a CA is expressed as a product of its invariant polynomials (IP), then

$$f(x) = x^{n_1} \cdot x^{n_2} \cdot x^{n_l} \phi_{l+1}(x)^{n_{l+1}} \cdot \phi_{\mathcal{N}}(x)^{n_{\mathcal{N}}}$$

where $\phi_i(x)$ is irreducible $(i = 1, 2, \dots, N)$. The number of *IP* comprising the characteristic polynomial

is denoted by \mathcal{N} . A factor x^{n_i} $(i = 1, 2, \dots, l)$, $(l \leq \mathcal{N})$, of f(x) represents a non-cyclic subspace whereas $\phi_i(x)^{n_i}$ corresponds to a cyclic subspace. For a group CA, there is not a single x^{n_i} components in f(x). That is, for a group CA, the characteristic polynomial is

$$f(x) = \phi_1(x)^{n_1} \phi_2(x)^{n_2} \cdots \phi_{\mathcal{N}}(x)^{n_{\mathcal{N}}}$$
(3)

The characterization of LCA state transition behavior is reported in [2, 3, 4, 6, 13, 14]. Some of the fundamental results reported in [3] and [20] are noted below. These results are important while building the analysis scheme for additive CA, described in *section III*.

II.2. Vector Space Theoretic Analysis of *LCA*

The following theorem, noted from [20], is one of the fundamental results reported for the LCA.

Theorem 1. The cycle structure of an *LCA* can be represented as

$$CS = [1(1) + \sum_{k_i} \sum_{j=0}^{m_{k_i}} \mu_{2^j \cdot k_i} (2^j \cdot k_i)]$$
(4)

where k_i is odd.

Example 1. The cycle structure of the *LCA* noted in *Fig. 2* is CS = [4(1), 2(2), 4(3), 2(6)]. We can also represent it as CS = [1(1), [3(1), 2(2)], [4(3), 2(6)]]. Here k_i s, the odd cycles, are 1 and 3. Corresponding to each odd cycle (k_i) , there is a cycle of length $2^j \cdot k_i$, in this case, j = 1 for both the odd cycles 1 and 3.

The above discussion demands definition of the following terminologies - primary cycles, secondary cycles and cycle family that are essential for the analysis of cycle structure of a CA.

Definition 1. Primary and secondary cycles : Each odd cycle (k_i) in the cycle structure is referred to as a primary cycle and the cycles which are of the form $2^j \cdot k_i$ $(j \ge 1)$ are referred to as secondary cycles.

Definition 2. *k* cycle family : All the cycles of the form $2^j \cdot k$ ($j \ge 0$), where *k* is the length of a primary cycle, are the member of a family of cycles referred to as *k*-cycle family. It is also referred to as primary cycle family.

The basic scheme for extracting the cycle structure of an LCA from its characteristic matrix T has been reported in [20]. Further, a more efficient and compact algorithm is presented in [11]. The execution steps of the algorithm are next illustrated through an example.

Example 2. Let the T matrix of a 5-cell *LCA* is

$$T = \begin{pmatrix} \begin{bmatrix} 1 \end{bmatrix} & 0 & 0 & 0 & 0 \\ 0 & \begin{bmatrix} 1 & 1 \end{bmatrix} & 0 & 0 \\ 0 & \begin{bmatrix} 0 & 1 \end{bmatrix} & 0 & 0 \\ 0 & 0 & 0 & \begin{bmatrix} 0 & 1 \end{bmatrix} \\ 0 & 0 & 0 & \begin{bmatrix} 1 & 1 \end{bmatrix} \end{pmatrix}$$

The following steps are to be executed to find the cycle structure of the LCA from its T matrix. Step 1 : Find out the characteristic polynomial the CA, illustrating each invariant polynomial separately. For the example CA, it is $(x + 1)(x + 1)^2(x^2 + x + 1)$.

Step 2 : Find out the cycle structure for each of the invariant polynomials. For the current example, these are

$$CS_{LCA(x+1)} = [1(1), 1(1)], \quad CS_{LCA(x+1)^2} = [1(1), 1(1), 1(2)], \quad CS_{LCA(x^2+x+1)} = [1(1), 1(3)]$$

Step 3 : Enumerate the complete cycle structure of the CA by successively performing cross product of cycle structures generated by each invariant polynomial [20]. Therefore, the example CA has the following cycle structure

 $CS_{LCA} = [1(1), 1(1)] \times [1(1), 1(1), 1(2)] \times [1(1), 1(3)] = [4(1), 2(2), 4(3), 2(6)].$

where \times represents the cross product operation and is defined as

Definition 3. Cross Product (\times) of two cycle structures CS_1 and CS_2 , where

$$CS_1 = [1(1) + \sum_{i_1=1}^{m_{k_{i_1}}} \mu_{k_{1i_1}}(k_{1i_1})] \text{ and } CS_2 = [1(1) + \sum_{i_2=1}^{m_{k_{i_2}}} \mu_{k_{1i_2}}(k_{1i_2})],$$

is the product of each i_1^{th} term of CS_1 and the i_2^{th} term of CS_2 . The product of $\mu_{k_{1i_1}}(k_{1i_1})$ and $\mu_{k_{2i_2}}(k_{2i_2})$ results in the cyclic component μ_k of length k following the equations [20]

$$\mu_k = \mu_{k_{1i_1}} \cdot \mu_{k_{2i_2}} \cdot gcd(k_{1i_1}, k_{2i_2}) \quad and \quad k = lcm(k_{1i_1}, k_{2i_2}) \tag{5}$$

Based on the results provided in this section, we report detail analysis of the state transition behavior of ACA that follows.

III. Vector Space Theoretic Analysis of Additive CA

Complete characterization of the state transition behavior of an ACA is reported in this section. An ACA, as noted in *section II*, is represented by the characteristic matrix T and the non-zero inversion vector F. The ACA generates more varieties of cycle structure than that of linear CA (LCA). Fig. 3 gives a typical example of cycle structure generated by ACA which is not available from an LCA. This section also highlights the variation of ACA cycle structure with that of its linear counterpart.

An ACA - C' is a group CA iff its linear counterpart C (represented by the same characteristic matrix T of C' with all 0s F Vector) is a group CA [3]. It signifies that the cycle structure of an ACA can be figured out from the analysis of state transition behavior of the corresponding LCA. Moreover, the concept of null space and its relationship with cycle length of an LCA is necessary for further analysis of cycle structure of ACA. These are reproduced from [10, 17].

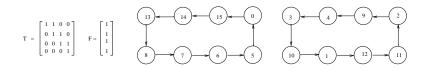


Figure 3. Additive CA and its state transition behavior. The cycle structure is CS' = 2(8). The cycle structure of the corresponding LCA is CS = [2(1), 1(2), 3(4)]

Definition 4. Null space : The null space of a matrix (T) consists of all such vectors that are transformed to the all-zero vector when premultiplied by the matrix.

Theorem III.1. [16] If an *LCA* represented by *T* has a cycle of length *k*, then the cardinality of the null space of $(T^k + I)$ denotes the number of states forming cycles of length *k* or sub-multiple of *k*.

Theorem III.2. [10] If an *LCA* is represented by *T* and for any state $\chi \neq 0$, $g(T) \cdot \chi = 0$, then g(x) and the characteristic polynomial f(x) have some common factor h(x).

The analysis of ACA state transition behavior is done with the solutions of following issues :

- A. To check whether a particular cycle length (k) is present in the cycle structure of an ACA.
- B. To find the special class of ACA(C') having the cycle structure as that of its linear counterpart C irrespective of its inversion vector F.
- C. To identify the class of C' whose cycle structure differs from that of C, and the properties of F vectors which impart this difference.
- D. To enumerate the cycle structure and depth of C'.

The following subsections report the analysis and results of above four issues.

III.1. Identification of a cycle of length (k) in ACA cycle structure

The following theorem enables us to determine whether a cycle of length k exists in an ACA or not.

Theorem 2. [16] In an ACA with characteristic matrix T and the inversion vector F, a cycle of length k exists if

$$rank([T^{k} + I]) = rank([T^{k} + I, \mathcal{F}]), where \mathcal{F} = [I + T + T^{2} + \dots + T^{k-1}]F$$

Proof:

Let χ be a state in a cycle of length k in an ACA C'. Hence, as per eq. (1) in section II,

$$\chi = [I + T + T^2 + \dots + T^{k-1}]F + T^k \cdot \chi$$

It can be written as

$$[T^k + I] \cdot \chi = \mathcal{F}, \text{ where } \mathcal{F} = [I + T + T^2 + \dots + T^{k-1}]F$$
(6)

If a cycle of length k is to exist in C', eq. (6) should be consistent. The condition for consistency is

$$rank([T^{k} + I]) = rank([T^{k} + I, \mathcal{F}])$$
(7)

Hence the proof.

III.2. Class of ACA with the cycle structure identical to corresponding linear CA

Theorem 2 is utilized to explore a special class of C'(ACA) that has cycle structure identical to that of C(LCA) irrespective of the inversion vector F. The following theorem defines such a class of C'.

Theorem 3. The cycle structures of C'(ACA) and C(LCA) are identical if the characteristic polynomial f(x) of the T matrix does not have a factor (x + 1).

Proof:

Let k be the length of a cycle of the linear CA - C with characteristic matrix T; characteristic polynomial f(x) of which does not have a factor (x + 1). To have a cycle of length k in the corresponding ACA - C', eq. (7) has to be satisfied.

The number of vectors, forming a cycle of length k or sub-multiple of k, in the ACA/LCA are derived from the enumeration of *null space* of $\alpha_1 = (T^k + I)$ [17].

Let $(x^k + 1) = g(x) \cdot \phi_c(x)$; $\phi_c(x)$ is the largest factor of the characteristic polynomial f(x) that divides $(x^k + 1)$. Therefore g(x) and f(x) don't have any common factor. Hence for each state χ , where $(T^k + I) \cdot \chi = 0$, there is a corresponding unique state $\tilde{\chi}$, where $\phi_c(T) \cdot \tilde{\chi} = 0$ (from *theorem III.2*). Hence, the cardinality of null space of $\alpha_2 = \phi_c(T)$ is same as α_1 .

Since f(x) does not have a factor (x + 1) and $x^k + 1 = (x + 1) \cdot (1 + x + x^2 + \dots + x^{k-1})$, therefore similarly the cardinality of the null space of $\alpha_3 = [I + T + T^2 \cdots T^{k-1}] = \alpha_1 = \alpha_2$. Hence,

$$rank(T^{k}+I) = rank(T^{k}+I, I+T+T^{2}\cdots T^{k-1})$$

that is,

$$rank(T^{k}+I) = rank(T^{k}+I,\mathcal{F})$$

directly follows for any F.

Therefore, all the cycle lengths of C also exist in C'. Since the number of vectors forming each cycle length is same for the both (directly derived from the cardinality of null space), the cycle structures for both the CA are identical. Hence the proof.

The following example illustrates the result of *theorem 3*.

Example 3. Let consider a 5-cell ACA with characteristic matrix T and the inversion vector $F \neq 0$, where

The characteristic polynomial of T is $(x^3 + x + 1)(x^2 + x + 1)$. The cycle structure of the corresponding LCA is [1(1),1(3),1(7),1(21)]. The ACA, as per *theorem* 3, has the identical cycle structure as that of the LCA, irrespective of F.

Let us consider a particular cycle of length 7. As per the theorem, we enumerate $\alpha_1 = T^7 + I$, $\alpha_2 = T^3 + T + 1$ and $\alpha_3 = T^6 + T^5 + T^4 + T^3 + T^2 + T + I$, where

$T^7 + I = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0$	$ \begin{array}{cccc} 1 & 1 \\ 1 & 0 \\ 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{array} \right), \qquad \qquad T^3 $	$+T+1 = \left(\begin{array}{rrrrr} 1 & 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0$
	$T^6 + T^5 + T^4 + T^3 + T^2 +$	$T + I = \left(\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$

All the three matrices with 3^{rd} , 4^{th} & 5^{th} rows as all zeros, have cardinality of null space as 8. Therefore, rank(α_1) = rank(α_2, α_3) and rank(α_1) = rank(α_2, \mathcal{F}); $\mathcal{F}=(I + T + T^2 + T^3 + T^4 + T^5 + T^6) \cdot F$. That is, both the C' and C have cycles of length 7. The number of states having cycle length 7 or submultiple of 7 (here it is 1) is 8. Therefore, the CA cyclic components of cycle length $7 = \frac{(8-1)}{7} = 1$ (as one state forms a self loop). The complete cycle structure of C' can be shown as CS' = [1(1), 1(3), 1(7), 1(21)], which is same as that of C.

III.3. Class of ACA with cycle structure different from that of LCA

From *theorem 3*, it is obvious that the cycle structure of the linear CA - C and the ACA - C' can differ only if the characteristic polynomial of the CA has a factor of (x + 1). The study of the role of (x + 1)factor helps identification of the class of ACA for which the cycle structure differs from that of the corresponding LCA. Let us concentrate on the LCA/ACA having a single invariant polynomial as $(x + 1)^n$. The generalization follows subsequently. The following terminologies are defined relating the cycle structure of LCA/ACA.

• $C(x + 1)^n$: The *LCA* with characteristic matrix *T*, the characteristic & minimal polynomial as $(x + 1)^n$ has the cycle structure [3]

$$CS = [1(1) + \sum_{j=0}^{m} \mu_{2^{j}}(2^{j})], \quad where \ m = \lceil \log_{2}(n) \rceil,$$
(8)

rank of
$$(T + I) = n - 1$$
, and rank of $(T + I)^{i} = n - i$. (9)

The $C\phi(x)$ and $C'\phi(x)$ represent the *LCA* and *ACA* respectively with characteristic polynomial $\phi(x)$.

• $[\mathbf{F}^{\mathbf{k}}]$: The set of inversion vectors which annihilates only $(x+1)^k$ - that is,

$$(T+I)^k \cdot \chi = 0, \text{ where } (T+I)^{k'} \cdot \chi \neq 0 \text{ and } k' < k \& \chi \in [F^k].$$
 (10)

• $\mathbf{Car}[\mathbf{F}^{\mathbf{k}}]$: Cardinality of the set F^k .

Based on the results of the *theorem* 2, we next identify the condition responsible for the difference in cycle structure of $C(x+1)^n$ and $C'(x+1)^n$.

Theorem 4. The ACA $C'(x+1)^n$, characterized by T and the inversion vector F, and $LCA C(x+1)^n$ have the identical cycle structure if

$$rank(T+I) = rank((T+I), F)$$
(11)

Proof:

In order to test whether the cycle structure of $LCA C(x + 1)^n$ and $ACA C'(x + 1)^n$ are identical, the consistency of eq. (6) is checked for the existence of a cycle of length k in C'. The LCA C has cycle(s) of length k, where from eq. (8),

$$k = 2^{j}, \quad j = (0, 1, 2, \cdots, m), \quad m = \lceil log_{2}(n) \rceil.$$

Since the cycle length (k) of $C(x + 1)^n$ is of the form 2^j , the relation for consistency (eq. (6)) can be rewritten as

$$(T^{2^{j}} + I) \cdot \chi = (I + T + T^{2} + \dots + T^{2^{j}-1})F$$
(12)

where $j = (0, 1, 2, \dots, m)$ and $m = \lceil log_2(n) \rceil$. The relation has to be consistent $\forall j$ to establish the equivalence of cycle structures of C & C'. Since

$$(T^{2^{j}} + I) = (T + I)^{2^{j}}$$
 and $(I + T + T^{2} + \dots + T^{2^{j}-1}) = (T + I)^{2^{j}-1}$,

eq. (12) can be rewritten as

$$(T+I)^{2^{j}} \cdot \chi = (T+I)^{2^{j}-1}F \Rightarrow (T+I)^{2^{j}-1}[(T+I) \cdot \chi = F]$$
(13)

It implies

$$(T+I)\cdot\chi = F\tag{14}$$

is to be consistent. That is, it should satisfy rank(T + I) = rank((T + I), F). Hence the proof.

Eq. (11) characterizes the F vector responsible for imparting the difference in cycle structure between LCA and ACA. The following theorem notes the characterization of F vectors.

Theorem 5. The cycle structure of $ACA C'(x+1)^n$ differs from that of its linear counterpart $C(x+1)^n$ if and only if the inversion vector F of $ACA \in [F^n]$.

Proof:

In order to prove the theorem, the consistency of eq. (14) is checked for inversion vector (1) $F \in [F^{n'}]$, where n' < n, and (2) $F \in [F^n]$.

Comment: The proof establishes the fact that in the first case both C and C' have the same cycle structure while in the second case the cycle structure of C' differs from that of C. The proof is established by checking consistency for every cycle length $(k = 2^j, j = \{0, 1, 2, \dots, m\})$

Case 1: Inversion vector $F \in [F^{n'}]$, where n' < n. We show that eq. 14 $((T + I) \cdot \chi = F)$ is consistent iff $\chi \in [F^{n'+1}]$. If $\chi \in [F^{n'+1}]$, then

$$(T+I)^{n'+1} \cdot \chi = 0 \Rightarrow (T+I)^{n'} \cdot [(T+I) \cdot \chi] = 0$$
(15)

As per the definition of $[F^{n'+1}]$ (noted in page 1010), the constituent vectors become zero only when premultiplied by $(T + I)^{n'+1}$ and higher factors. Therefore, $(T + I) \cdot \chi \neq 0$ and the vector y is formed by enumerating the equation

$$(T+I) \cdot \chi = y \tag{16}$$

Moreover, eq. (15) can be rewritten as

$$(T+I)^{n'} \cdot y = 0 \quad \Rightarrow \quad y \in [F^{n'}] \quad i.e. \quad (T+I)^{n''} \cdot y \neq 0 \quad n'' < n \tag{17}$$

Since rank of (T + I) is (n - 1), χ_i (i = 1, 2 and $\chi_1, \chi_2 \in [F^{n'+1}]$), while premultiplied with (T + I) generates the same y in eq. (16). That is,

$$(T+I)\cdot\chi_1 = (T+I)\cdot\chi_2 = y$$

Hence, exploiting the all possible pairs of χ , we obtain the set of y having cardinality $\frac{Car(F^{n'+1})}{2}$, where the cardinality of $[F^{n'+1}]$ is denoted as $Car(F^{n'+1})$.

Since rank of $(T + I)^{n'+1}$ is one less than that of $(T + I)^{n'}$ (eq. (9)), therefore

$$Car(F^{n'+1}) = 2 \times Car(F^{n'})$$

Therefore, $Car(y) = Car(F^{n'})$ - that is, the full set of $(F^{n'})$ is represented by y. It implies, the relation

$$(T+I) \cdot \chi = F$$

is consistent for all values of $F \in [F^{n'}]$ and consequently, eq. (13) is consistent for all $F \in [F^{n'}]$, n' < n.

Case 2 : Checking consistency of cycle length for ACA $C'(x+1)^n$ with inversion vector $F \in [F^n]$: In this case, it is shown that eq. (14) is inconsistent for all cycle length $2^j \leq n$. Multiplying either side of eq. (14) with $(T+I)^{n-1}$ we obtain

$$(T+I)^n \chi = (T+I)^{n-1} F.$$
(18)

It is inconsistent since the left hand side (LHS) = 0 while right hand side $(RHS) \neq 0$. \Box

Theorem 5 leads to the following corollary that specifies the cycle structure of $C'(x+1)^n$.

Corollary III.1. The cycle structure of $C'(x+1)^n$ with $F \in F^n$ is

$$CS' = \mu'(2^{\mathcal{M}}); \quad \mu' = 2^{n-\mathcal{M}} \quad and \quad \mathcal{M} = \lfloor \log_2(n) \rfloor + 1$$
(19)

where the cycle structure of C(x + 1) (eq. (8), page 1009) is

$$CS = [1(1) + \sum_{j=0}^{m} \mu_{2^{j}}(2^{j})] \text{ where } m = \lceil log_{2}(n) \rceil.$$

Proof:

Since the characteristic and minimal polynomial of C is $(x + 1)^n$ - that is, $(T + I)^n$ is always = 0. Therefore, eq. (14) becomes consistent (LHS = RHS = 0), when we premultiply both sides by $(T+I)^n$. Consequently, eq. (13) is consistent if $2^j - 1 \ge n$ - that is, cycles of length 2^j exist in C' if $2^j - 1 \ge n$. The minimum value of j for which the equation becomes consistent is represented by \mathcal{M} , where $2^{\mathcal{M}-1} \le n < 2^{\mathcal{M}}$.

- 1. If $n = 2^{\lceil \log_2 n \rceil}$, then n is of the form 2^j ; where j is an integer. Therefore, $\lceil \log_2 n \rceil = \lfloor \log_2 n \rfloor = \log_2 n$. In that case, $\mathcal{M} = \lfloor \log_2 n \rfloor + 1$.
- 2. If $n < 2^{\lceil \log_2 n \rceil}$, then $\mathcal{M} = \lceil \log_2 n \rceil = \lfloor \log_2 n \rfloor + 1$.

Considering (1) and (2), it can be found that $\mathcal{M} = |\log_2 n| + 1$.

Since all the states fall in the null space of $(T+I)^{2^{\mathcal{M}}}$, as per *theorem III.1*, the ACA can have cycles only of length $2^{\mathcal{M}}$ and the number of cyclic components (μ') is $2^n/2^{\mathcal{M}} = 2^{n-\mathcal{M}}$.

The \mathcal{M} is denoted as the *minimum additive factor*. The following example illustrates the results of *theorem 5* and *corollary III.1*.

Example 4. Let us consider the following T matrix and two inversion vectors F_1 and F_2 of the 4-cell ACA -

$$T = \begin{pmatrix} 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{pmatrix} \quad F_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} \quad F_2 = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 0 \end{pmatrix}.$$

The characteristic polynomial and minimal polynomial of the T is $(x + 1)^4$.

The cycle structure of the LCA counterpart with characteristic matrix T is CS = [2(1), 1(2), 3(4)].

The inversion vector F_1 annihilates $(T + I)^4$. The ACA with characteristic matrix T and the inversion vector F_1 has the cycle structure [2(8)] (Fig. 3). However, F_2 cannot annihilate $(T+I)^4$. The ACA with characteristic matrix T and the inversion vector F_2 has its cycle structure CS' identical to CS generated by the LCA.

Theorem 5 & corollary III.1 can be extended to a more generalized class of LCA/ACA. As per theorem 3, it is noted that the cycle structure of ACA - C' can differ from C only if the characteristic polynomial of C/C' contains a factor of (x+1). That is, the characteristic polynomial can be represented as

$$f(x) = (x+1)^{n_1} \cdot (x+1)^{n_l} \phi_{l+1}(x)^{n_{l+1}} \cdot \phi_{\mathcal{N}}(x)^{n_{\mathcal{N}}}$$

where (i) each $\phi_i(x)^{n_i}$ is an invariant polynomial and (ii) the irreducible factor of the first l invariant polynomials is (x + 1). The following theorem states the nature of cycle structure of such *LCA* and *ACA*.

Theorem III.3. The cycle structure of *LCA*, with characteristic matrix T and a factor (x + 1) in the characteristic polynomial, is

$$CS = [1(1) + \sum_{k_i} \sum_{j=0}^{m_{k_i}} \mu_{2^j \cdot k_i} (2^j \cdot k_i)]; \quad k_1 = 1;$$
(20)

whereas the cycle structure of ACA, with the characteristic matrix T & inversion vector F and having the largest (x + 1)-invariant polynomial annihilated by F as $(x + 1)^{n_i}$, is

$$CS' = \sum_{k_i} \sum_{j=\mathcal{M}}^{m_{k_i}} \mu'_{2^j \cdot k_i} (2^j \cdot k_i), \quad k_1 = 1$$
(21)

where

$$\mu'_{2^{\mathcal{M}}\cdot k_{i}} = \begin{cases} \sum_{j=0}^{\mathcal{M}} (\mu_{2^{j}\cdot k_{i}} \times 2^{j} \cdot k_{i}) + 1 \\ \frac{j=0}{2^{\mathcal{M}\cdot k_{i}}} & \text{for } \mathbf{k} = 1 \\ \sum_{j=0}^{\mathcal{M}} \mu_{2^{j}\cdot k_{i}} \times 2^{j} \cdot k_{i} \\ \frac{j=0}{2^{\mathcal{M}\cdot k_{i}}} & \text{otherwise} \end{cases}$$

and $\mu'_{2^{j}\cdot k_{i}} = \mu_{2^{j}\cdot k_{i}} \quad j > \mathcal{M}, \quad \mathcal{M} = \lfloor \log_{2} n_{i} \rfloor + 1. \end{cases}$

Proof:

The proof is developed with the assumption that the $(x + 1)^{n_i}$ is the *only* factor annihilated by F. The generalization can be done where $(x + 1)^{n_i}$ is the *largest* factor annihilated by F. The characteristic polynomial f(x) of such a CA is denoted as

$$f(x) = (x+1)^{n_i} \times \tilde{\phi}(x); \quad \tilde{\phi}(x) = \phi_1(x)^{n_1} \cdot \phi_{i-1}(x)^{n_{i-1}} \cdot \phi_{i+1}(x)^{n_{i+1}} \cdot \phi_{\mathcal{N}}(x)^{n_{\mathcal{N}}}$$

The cycle structure $CS\tilde{\phi}(x)$ corresponding to $\tilde{\phi}(x)$ is same for both the *LCA* and *ACA*. Let *F* be the inversion vector which annihilates the factor $(x + 1)^{n_i}$.

As per the *theorem 5*, the cycle structures generated due to the factor $(x + 1)^{n_i}$ differ in C & C' and these are $CS'(x + 1)^{n_i}$ and $CS(x + 1)^{n_i}$ (from *corollary III.1*). From the eqs. (19) & (8), we have

$$CS'(x+1)^{n_i} = \mu'(2^{\mathcal{M}}), \quad \mu' = 2^{n_i - \mathcal{M}} \quad \& \quad \mathcal{M} = \lfloor \log_2(n_i) \rfloor + 1$$

and
$$CS(x+1)^{n_i} = [1(1) + \sum_{j=0}^m \tilde{\mu}_{2^j}(2^j)], \text{ where } m = \lceil \log_2(n_i) \rceil.$$

The cycle structure of $C\tilde{\phi}(x)$ follows from eq. (4) and forms a *linear cycle structure*

$$CS\tilde{\phi}(x) = [1(1) + \sum_{k_i} \sum_{j=0}^{m_{k_i}} \hat{\mu}_{2^j \cdot k_i} (2^j \cdot k_i)]$$

Therefore, the cycle structures CS and CS' are represented as

$$CS = CS\tilde{\phi}(x) \times CS(x+1)^{n_i} = [1(1) + \sum_{k_i} \sum_{j=0}^{m_{k_i}} \hat{\mu}_{2^j \cdot k_i}(2^j \cdot k_i)] \times [1(1) + \sum_{j=0}^m \tilde{\mu}_{2^j}(2^j)]$$
(22)

$$CS' = CS\tilde{\phi}(x) \times CS'(x+1)^{n_i} = [1(1) + \sum_{k_i} \sum_{j=0}^{m_{k_i}} \hat{\mu}_{2^j \cdot k_i}(2^j \cdot k_i)] \times [\mu'(2^{\mathcal{M}})]$$
(23)

The above two cross products² produce CS and CS' specified in the eqs. (20) and (21) respectively.

The results of theorem III.3 are illustrated below.

Example 5. Let us consider the T matrix of the 5-cell CA of Fig. 2 as shown below

Consider the three ACA with the characteristic matrix as T having these inversion vectors

	(0)			(1)			(0)	
	1			1			1	
$F_1 =$	1	,	$F_2 =$	1	,	and $F_3 =$	0	
	0			0			1	
	(0)			0)			$\left(1 \right)$	

- The characteristic polynomial f(x) of the example LCA/ACA, in invariant polynomial form, is $f(x) = (x + 1)(x + 1)^2(x^2 + x + 1)$.
- The inversion vector $F_1 = [01100]$ annihilates only $(x + 1)^2$. Therefore, $\mathcal{M} = \lfloor log_2(2) \rfloor + 1 = 2$ (*corollary III.1*). For this case,

²note the appendix for details

- f(x) can be expressed as $\tilde{\phi}(x) \times (x+1)^2$, where $\tilde{\phi}(x) = (1+x) \cdot (x^2 + x + 1)$.
- The cycle structure $CS\tilde{\phi}(x) = CS(1+x) \times CS(x^2+x+1) = [1(1), 1(1)] \times [1(1), 1(3)] = [2(1), 2(3)].$
- The cycle structure $CS(1+x)^2 = [1(1), 1(1), 1(2)]$, whereas $CS'(1+x)^2 = [1(4)]$.
- The resultant cycle structure $CS = CS\tilde{\phi}(x) \times CS(1+x)^2 = [[4(1), 2(2)], [4(3), 2(6)]]$ while $CS' = CS\tilde{\phi}(x) \times CS'(1+x)^2 = [2(4), 2(12)].$
- Similarly, F₂ = [11100] annihilates both the invariant polynomials (x + 1) and (x + 1)². This is, the case where (x + 1)² is the largest factor annihilated by F₂. Therefore, CS' = CS'(x + 1) × CS'(x + 1)² × CS(x² + x + 1) = [1(4)] × [1(2)] × [1(1), 1(3)] = [2(4), 2(12)]. It is same as that of the cycle structure generated by F₁.
- The changes in cycle structure of each cycle family k_i in CS and CS' respectively are as per eq. (21) of *theorem III.3*. For example, 2(4) where k₁ = 1 and 2^M = 4, the corresponding cycle structure in C which have got merged in C' is [4(1),2(2)]. The cyclic component of μ_{2².1} has been formed as per eq. 21. Here μ_{2².1} = ^{4 × 1 + 2 × 2}/₄ = 2.
- Consider $F_3 = [01000]$. It doesn't annihilate either (x + 1) or $(x + 1)^2$. Therefore, the cycle structure of C and C' are the same [4(1), 2(2), 4(3), 2(6)].

Theorem III.3 and the earlier example identify the nature of cycle structure of ACA. The complete algorithm to compute cycle structure of an ACA is described next.

III.4. Enumerating cycle structure of an ACA

The scheme to enumerate cycle structure of an ACA is developed based on the theory reported in the earlier subsections. *Theorem 3* reports that the cycle structures of ACA and the corresponding LCA are identical if the characteristic polynomial of the linear operator T doesn't contain the factor (x + 1). On the other hand, if the characteristic polynomial is having the factor (x + 1), then the cycle structure of the ACA is to be computed from eq. (21). However, it requires evaluation of \mathcal{M} - the minimum additive factor.

Definition 5. The least cycle length of any primary cycle (k) in CS' (cycle structure of ACA) is given by $2^{\mathcal{M}} \cdot k$; where \mathcal{M} is denoted as minimum additive factor.

The value of \mathcal{M} can be deduced from *theorem* 2. The rank of $[T^k + I]$ and $[T^k + I, \mathcal{F}]$ is successively compared for all $k = 2^m \cdot k_1$, $m \ge 0$, until the ranks become equal. The value of k at which both the ranks become equal provides the value of \mathcal{M} . The algorithm to deduce \mathcal{M} is next elaborated.

Algorithm 1. $Enum_{\mathcal{M}}(T, F)$ Input : T matrix, F Vector Output : \mathcal{M} $\mathcal{M} = -1$ I, \mathcal{F}] do {
$$\begin{split} \mathcal{M} &= \mathcal{M} + 1 \\ \text{Evaluate } \mathcal{F} &= [I + T + T^2 + \ldots + T^{2^{\mathcal{M}} - 1}]F \\ \text{Evaluate } \mathsf{R1} &= \mathsf{rank}[T^{2^{\mathcal{M}} - 1} + 1] \text{ and } \mathsf{R2} = \mathsf{rank}[T^{2^{\mathcal{M}} - 1} + I, \mathcal{F}] \\ \} \text{while (} R1 < R2\text{)} \\ \text{return } \mathcal{M} \end{split}$$

Once the \mathcal{M} is enumerated, the cycle structure of ACA - C' for each k_i -cycle family constituting CS' is evaluated from the CS of LCA - C (theorem III.3).

Algorithm 2. Enum_CS'_of_ACA(T, F, CS')Input : Characteristic matrix T and inversion vector FOutput : The cycle structure of ACA - C'Enumerate cycle structure CS of LCA, where

$$CS = [1(1) + \sum_{k_i} \sum_{j=0}^{m_{k_i}} \mu_{2^j \cdot k_i} (2^j \cdot k_i)]; \ k_i \ is \ odd \ and \ [CL] \ is \ the \ set \ of \ odd \ cycles \ of \ length \ k_i.$$

If characteristic polynomial of T doesn't have a factor of (x + 1), then

Output the cycle structure CS of C as CS' of C'.

else

$$\begin{split} \mathcal{M} &= Enum\mathcal{M}(T,F) \\ \text{for each } k_i \in [CL] \\ \{ \\ \text{Evaluate } \mu'_{2^{\mathcal{M}} \cdot k_i} \text{ for each } k_i \text{ following eq. (21)} \\ \mu'_{2^j \cdot k_i} &= \mu_{2^j \cdot k_i} \text{ for } \forall j > \mathcal{M} \\ \} \\ \text{Output } \mathbf{CS}' &= \sum_{k_i} \sum_{j=\mathcal{M}}^{m_{k_i}} \mu_{2^j \cdot k_i} (2^j \cdot k_i), \quad k_1 = 1 \end{split}$$

The next example illustrates the execution steps of the algorithm. The ACA is represented by its T matrix and F vector; the T matrix of this example is the one used in the *example 2*.

Example 6. Let the T matrix of a 5-cell ACA be

$$T = \begin{pmatrix} [1] & 0 & 0 & 0 & 0 \\ 0 & [1 & 1] & 0 & 0 \\ 0 & [0 & 1] & 0 & 0 \\ 0 & 0 & 0 & [0 & 1] \\ 0 & 0 & 0 & [1 & 1] \end{pmatrix} \quad \text{and } F = \begin{pmatrix} 1 \\ 1 \\ 0 \\ 1 \\ 1 \end{pmatrix}$$

The matrix $[T^k + I, \mathcal{F}]$ is referred to as *augmented matrix*.

Step 1 & 2 : Finding characteristic polynomial and cycle structure -

Characteristic polynomial $f(x) = (x + 1)^3 \cdot (x^2 + x + 1)$

The cycle structure of the *LCA* having the characteristic polynomial f(x) is [4(1), 2(2), 4(3), 2(6)]. Step 3 : Finding the values of \mathcal{M} .

- Rank of $(T^1 + I)$ is 3, while the rank of the augmented matrix is 4. Hence, cycle of length 1 does not exist.
- Rank of $(T^2 + I)$ is 2, while the rank of the augmented matrix is 3. Hence, cycle of length 2 also does not exist in the ACA.
- Rank of $(T^4 + I)$ is 2, while the rank of the augmented matrix is 2. So, the value of \mathcal{M} is 2.

Step 4 : Finding the components **1.** $k_i = 1$ $\mu_{2^2.1} = \frac{4 \times 1 + 2 \times 2 + 0 \times 4}{2^{2.1}} = 2$. **2.** $k_i = 3$ $\mu_{2^2.3} = \frac{4 \times 3 + 2 \times 6 + 0 \times 12}{2^{2.3}} = 2$. Hence, the cycle structure becomes [2(4), 2(12)].

IV. Conclusion

This paper presents the complete scheme to compute the cycle structure of additive cellular automata (ACA). The similarities between the cycle structure of an LCA and the cycle structure produced by the ACA is explored. An analytical scheme has been devised to extract the similarities for enumeration of the cycle structure of ACA. The scheme can be used further to efficiently utilize ACA for different purposes in the future.

Appendix

Elaboration of the steps to obtain the cross product of eq. (23)

The cycle structure (CS) of an LCA is evaluated from the expression referred in eq. (22). The expressions for CS noted in eq. (22) is reproduced below along with the final result noted in eq. (20). The intermediate computation steps are not shown as it can be directly obtained through simple cross product [20].

$$CS = CS\tilde{\phi}(x) \times CS(x+1)^{n_i} = [1(1) + \sum_{k_i} \sum_{j=0}^{m_{k_i}} \hat{\mu}_{2^j \cdot k_i}(2^j \cdot k_i)] \times [1(1) + \sum_{j=0}^m \tilde{\mu}_{2^j}(2^j)]$$
$$= [1(1) + \sum_{k_i} \sum_{j=0}^{m_{k_i}} \mu_{2^j \cdot k_i}(2^j \cdot k_i)]; \quad k_1 = 1;$$

The cycle structure (CS') of ACA is evaluated through eq. (23) reproduced below.

$$CS' = CS\tilde{\phi}(x) \times CS'(x+1)^{n_i} = [1(1) + \sum_{k_i} \sum_{j=0}^{m_{k_i}} \hat{\mu}_{2^j \cdot k_i}(2^j \cdot k_i)] \times [\mu'(2^{\mathcal{M}})]$$

The cross product of eq. (23) produces eq. (21) (page 1013). As can be seen from eq. (21), the cyclic components of CS', $\mu'_{2^j \cdot k_i}$ has some relationship with $\mu_{2^j \cdot k_i}$ - the cyclic components of CS. This section establishes the relationship defined in eq. (21). The relationship between $\mu_{2^j \cdot k_i}$ and $\mu'_{2^j \cdot k_i}$ comprises of two parts - (a) where $j > \mathcal{M}$ and (b) where $j \leq \mathcal{M}$

Let us consider a particular cycle family k_i . We (1) first show that the number of states encompassed by a the family is same for both CS and CS'. (2) Then we show $\mu_{2^j \cdot k_i} = \mu'_{2^j \cdot k_i}$ for $j > \mathcal{M}$ and finally (3) establish the relation between $\mu'_{2^{\mathcal{M}} \cdot k_i}$ and $\mu_{2^j \cdot k_i}$ for $j \leq \mathcal{M}$.

1. Computing the number of states in a family k_i .

From eq. (5), product of $\mu_{k_{1i_1}}(k_{1i_1})$ and $\mu_{k_{2i_2}}(k_{2i_2})$ results in the cyclic component μ_k of length k, where $\mu_k = \mu_{k_{1i_1}} \cdot \mu_{k_{2i_2}} \cdot gcd(k_{1i_1}, k_{2i_2})$ and $k = lcm(k_{1i_1}, k_{2i_2})$. Therefore, the number of states involved in forming the cycle is given by

$$\mu \times k = (\mu_{k_{1i_1}} \cdot \mu_{k_{2i_2}} \cdot gcd(k_{1i_1}, k_{2i_2})) \times (lcm(k_{1i_1}, k_{2i_2})) = \mu_{k_{1i_1}} \cdot \mu_{k_{2i_2}} \cdot k_{1i_1} \cdot k_{2i_2}$$
(24)

The above result can be applied to compute the number of states of a particular family k_i in CS and CS'. Let refer cycle structure of k_i family as $CS(k_i)$ and $CS'(k_i)$ for LCA and ACA respectively. The number of states covered are considered as $S(k_i)$ and $S'(k_i)$ respectively in $CS(k_i)$ and $CS'(k_i)$. Therefore,

$$CS(k_i) = \sum_{j=0}^{m_{k_i}} \hat{\mu}_{2^j \cdot k_i} (2^j \cdot k_i) \times [1(1) + \sum_{j=0}^m \tilde{\mu}_{2^j} (2^j)] \Rightarrow S(k_i) = \sum_{j=0}^{m_{k_i}} (\hat{\mu}_{2^j \cdot k_i} \times 2^j \cdot k_i \times [1 \times 1 + \sum_{j=0}^m (\tilde{\mu}_{2^j} \times 2^j)])$$

Similarly,

$$CS'(k_i) = \sum_{j=0}^{m_{k_i}} \hat{\mu}_{2^j \cdot k_i} (2^j \cdot k_i) \times [\mu'(2^{\mathcal{M}})] \Rightarrow S'(k_i)] = \sum_{j=0}^{m_{k_i}} (\hat{\mu}_{2^j \cdot k_i} \times 2^j \cdot k_i \times \mu' \times 2^{\mathcal{M}})$$

From the eqs. (19) & (8), we have

$$\mu' \times 2^{\mathcal{M}} = 1 + \sum_{j=0}^{m} \tilde{\mu}_{2^{j}}(2^{j})$$
(25)

Therefore, $S(k_i) = S'(k_i)$ — Result (1). 2. The $\mu_{2^j \cdot k_i} = \mu'_{2^j \cdot k_i}$ for $j > \mathcal{M}$.

$$\begin{split} CS &= CS\tilde{\phi}(x) \times CS(x+1)^{n_i} = [1(1) + \sum_{k_i} \sum_{j=0}^{m_{k_i}} \hat{\mu}_{2^j \cdot k_i}(2^j \cdot k_i)] \times [1(1) + \sum_{j=0}^m \tilde{\mu}_{2^j}(2^j)] \\ &= [1(1) + \sum_{j=0}^m \tilde{\mu}_{2^j}(2^j) + \sum_{k_i} (\sum_{j=0}^{m_{k_i}} \hat{\mu}_{2^j \cdot k_i}(2^j \cdot k_i) \times [1(1) + \sum_{j=0}^m \tilde{\mu}_{2^j}(2^j)])] \\ &= [1(1) + \sum_{j=0}^m \tilde{\mu}_{2^j}(2^j) + \sum_{k_i} (\sum_{j=0}^{\mathcal{M}} \hat{\mu}_{2^j \cdot k_i}(2^j \cdot k_i) \times [1(1) + \sum_{j=0}^m \tilde{\mu}_{2^j}(2^j)] + \sum_{j=\mathcal{M}+1}^{m_{k_i}} \hat{\mu}_{2^j \cdot k_i}(2^j \cdot k_i) \times [1(1) + \sum_{j=0}^m \tilde{\mu}_{2^j}(2^j)] + \sum_{j=\mathcal{M}+1}^m \tilde{\mu}_{2^j \cdot k_i}(2^j \cdot k_i) \times [1(1) + \sum_{j=0}^m \tilde{\mu}_{2^j}(2^j)])] \end{split}$$

(i) Considering $2^{j} = 2^{j} \cdot k_{i}$, where $k_{1} = 1$ & since $m \leq \mathcal{M}$ (eqs. (19) & (8))

$$CS = [1(1) + \sum_{k_i} (\sum_{j=0}^{\mathcal{M}} \mu_{2^j \cdot k_i}(2^j \cdot k_i) + \sum_{j=\mathcal{M}+1}^{m_{k_i}} \hat{\mu}_{2^j \cdot k_i} \times [1(1) + \sum_{j=0}^{m} \tilde{\mu}_{2^j}(2^j)](2^j \cdot k_i))]$$

$$\begin{split} CS' &= CS\tilde{\phi}(x) \times CS'(x+1)^{n_i} = [1(1) + \sum_{k_i} \sum_{j=0}^{m_{k_i}} \hat{\mu}_{2^j \cdot k_i} (2^j \cdot k_i)] \times [\mu'(2^{\mathcal{M}})] \\ &= [\mu'(2^{\mathcal{M}})] + \sum_{k_i} [\sum_{j=0}^{m_{k_i}} \hat{\mu}_{2^j \cdot k_i} (2^j \cdot k_i) \times \mu'(2^{\mathcal{M}})] \\ &= [\mu'(2^{\mathcal{M}})] + \sum_{k_i} [(\sum_{j=0}^{\mathcal{M}} \hat{\mu}_{2^j \cdot k_i} (2^j \cdot k_i) \times \mu'(2^{\mathcal{M}})) + (\sum_{j=\mathcal{M}+1}^{m_{k_i}} \hat{\mu}_{2^j \cdot k_i} (2^j \cdot k_i) \times \mu'(2^{\mathcal{M}}))] \end{split}$$

(ii) Considering $2^{\mathcal{M}} = 2^{\mathcal{M}} \cdot k_i$, where $k_1 = 1$.

$$CS' = \sum_{k_i} \left[\left(\sum_{j=0}^{\mathcal{M}} (\mu_{2^{\mathcal{M}} \cdot k_i} (2^{\mathcal{M}} \cdot k_i)) + \sum_{j=\mathcal{M}+1}^{m_{k_i}} \mu' \times 2^{\mathcal{M}} \times \hat{\mu}_{2^j \cdot k_i} (2^j \cdot k_i) \right] \right]$$

For $j > \mathcal{M}$

$$\mu_{2^{j} \cdot k_{i}} = \hat{\mu}_{2^{j} \cdot k_{i}} \times [1 + \sum_{j=0}^{m} \tilde{\mu}_{2^{j}}(2^{j})]$$
$$\mu'_{2^{j} \cdot k_{i}} = \mu' \times 2^{\mathcal{M}} \times \hat{\mu}_{2^{j} \cdot k_{i}}$$

Therefore, $\mu_{2^j \cdot k_i} = \mu'_{2^j \cdot k_i}$ for $j > \mathcal{M}$ (eq. (25)) – Result (2). 3. Computation of $\mu'_{2^{\mathcal{M}} \cdot k_i}$

Combining the results (1) and (2), the number of states covered by the cycles of length $\leq 2^{\mathcal{M}} \cdot k_i$ are same for both *LCA* and *ACA*.

The states covered by ACA are $\mu'_{2^{\mathcal{M}}\cdot k_i} \times 2^{\mathcal{M}} \cdot k_i$.

Similarly, the states covered by an *LCA* for any $k_i \neq 1$ are $\sum_{j=0}^{\mathcal{M}} \mu_{2^j \cdot k_i} \times 2^j \cdot k_i$. Therefore

Therefore,

$$\mu'_{2\mathcal{M}\cdot k_i} = \frac{\sum_{j=0}^{\mathcal{M}} \mu_{2^j \cdot k_i} \times 2^j \cdot k_i}{2^{\mathcal{M}} \cdot k_i}; \quad k_i \neq 1$$

However, the states covered by the LCA for $k_i = 1$ are $\sum_{j=0}^{\mathcal{M}} (\mu_{2^j \cdot k_i} \times 2^j \cdot k_i) + 1$ and, therefore,

$$\mu'_{2^{\mathcal{M}}\cdot k_i} = \frac{1 + \sum_{j=0}^{\mathcal{M}} \mu_{2^j \cdot k_i} \times 2^j \cdot k_i}{2^{\mathcal{M}} \cdot k_i}; \quad k_i = 1$$

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