Chaotic Random Sequence Generated from Tent Map on Variant Maps

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ABSTRACT

Chaotic sequences, have being widely used in mobile communications and cyberspace security. Using the sequences, various stochastic analysis schemes are developed. The Tent map is one of the widely used chaotic maps with good ergodic uniformity. In this paper, the variant maps are used to illustrate the Tent chaotic sequence under different lengths and control parameters to show the statistical characteristics of sequences. Results are shown that when the control parameter of Tent map is close to 0.5, the generated sequences have the symmetrical distribution, that is more stable if the sequence is longer.

Keywords: tent map, chaotic random sequence, variant map, visual testing.

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1. Introduction

Pseudo-random bit sequences have a wide range of applications, in the field of cryptography and communications[1]. Chaotic maps have the features of ergodicity, sensitivity to initial conditions and sensitivity to control parameters[2], can be implemented in higher speed for various algorithms. Tent map is the simplest chaotic maps, one of the best-known discrete dynamical systems[3][4]. It can be used in optimization algorithms[5], key stream generation[6], chaotic block cryptosystem[7] and digital noise production[8].

From a distribution viewpoint, the randomness of Tent chaotic sequences may not be ideal[9]. Such measures do not reflect the various statistical distributions of Tent maps under different parameters. In this paper, chaotic sequences generated by different parameters and iteration times are tested and observed on variant maps, to illustrate the statistical distribution of Tent map systematically.

2. Tent Map

Lemma 2.1. Tent map is an one-dimensional piecewise linear map. The mapping form of Tent map is as follows:

\[ x_{n+1} = \begin{cases} \frac{x_n}{a} & 0 \leq x_n < a \\ \frac{1-x_n}{1-a} & a \leq x_n \leq 1 \end{cases} \]  

(2.1)

of DNA data[11], bat echolocation call[12] and stream ciphers[13].

Various variant maps are shown in Figure 1, four maps generated from one sample sequence as (a)-(d). In map (a), a 2DPQ map is shown in 3D distribution; in map (b), a 2DPQ map is shown as a 2D color map; in map (c), the 2DPQ map is projected in the horizontal direction as a 1DP map in a normal bell shape; in map (d), the 2DPQ map is projected in the horizontal direction as a 1DQ map in a narrow bell shape respectively.

3. Variant Map

Variant map is a visual technology using multiple statistical probability distributions as visual maps to handle multiple 0-1 vectors in its phase space on variant framework[10]. Different applications are explored using variant maps on sequences and digital noise production[8].

The basic symbols are used in this paper as follows:

(1)\(S\): an input 0-1 sequence
(2)\(s_i\): the i-th element of S
(3)\(S_j\): the j-th segment of S
(4)\(N\): the length of S
(5)\(m\): the length of a segment
(6)\(M\): the number of segments
(7)p: the number of 1s elements in the segment
(8)q: the number of 01s patterns in the segment

4. Testing Procedure

This testing procedure is shown in Figure 2 and composed of five steps:

Step 1: Initialization;
Step 2: Generation; Step 3: Segmentation; Step 4: Measurements; Step 5: Visual Maps.

Figure 2: Testing Procedure

Each step can be described as follows:

**Step 1: Initialization**
Using the condition shown in Lemma 2.1, different initial parameters are selected: the initial value $x_0$, controls parameter $a$, iteration number $n$, generating a chaotic sequence with length $N$.

**Step 2: Generation**
Applying the initial parameters, a 0-1 random sequence $S$ is generated with length $N$.

$$S = (s_0, s_1, ..., s_i, ..., s_{N-1}), s_i \in [0, 1], 0 \leq i < N$$

**Step 3: Segmentation**
Make $m$ length segmentation on the 0-1 flow of $S$, to divide it into $M$ segments and each segment has $m$ elements.

$$M = \left\lfloor \frac{N}{m} \right\rfloor$$

$$S = (S_0, S_1, ..., S_j, ..., S_{M-1}), 0 \leq j < M$$

$$S_j = (s_j \times m, s_j \times m + 1, ..., s_j \times m + k, ..., s_j \times m + m - 1), 0 \leq k < m, 0 \leq j \times m + k < N$$
Step 4: Measurements
Two measurements \( p_j \) and \( q_j \) are calculated from each segment \( S_j \) of \( S \), that is the number of 1 and 01. Pay attention to the 01 on tail head connecting. For example, a segment \( S_1 = 10010110 \), the measurements are \( p_1 = 4 \) and \( q_1 = 3 \). Calculating all segments of \( S \), a set of \( p \) and \( q \) measurements are figured out:

\[
\begin{align*}
 p_{\text{set}} &= \{p_0, p_1, ..., p_j, ..., p_{M-1}\} = \{p_j\}^{M-1}_{j=0}, 0 \leq j < M \\
 q_{\text{set}} &= \{q_0, q_1, ..., q_j, ..., q_{M-1}\} = \{q_j\}^{M-1}_{j=0}, 0 \leq j < M
\end{align*}
\]

Step 5: Visual Maps
Order segments by its value of \( p \) and \( q \), types of diagrams can be created: Firstly, 1DP and 1DQ are one-dimension maps, which is sorted from \( \{p_j\}^{M-1}_{j=0} \) and \( \{q_j\}^{M-1}_{j=0} \). Secondly, two-dimension 2DPQ is sorted from a pair of measurements \( \{(p_j, q_j)\}^{M-1}_{j=0} \), created from \( \{p_j\}^{M-1}_{j=0} \) and \( \{q_j\}^{M-1}_{j=0} \).

5. Test Results
In this paper, we use 2DPQ to show the distribution characteristic of Tent map. The test sample is a binary sequence cipher stream generated by Tent map. The initial value \( x_0 \) is fixed at 0.314 and the control parameter \( a \) is from 0 to 1, which satisfies the chaotic condition. The sample size ranges from 1KB to 100MB.

5.1. Test Results on Different Segment Lengths
The parameters and the results are shown in Table 1 and Figure 3 respectively:

| Segment Length \( m \) | Initial Value \( x_0 \) | Control Parameter \( a \) | Length \( N \)      | Figure 3 |
|------------------------|------------------------|------------------------|---------------------|---------|
| 8                      | 0.314                  | 0.499                  | 10,000,000          | (a)     |
| 16                     |                         |                        |                     | (b)     |
| 32                     |                         |                        |                     | (c)     |
| 64                     |                         |                        |                     | (d)     |
| 128                    |                         |                        |                     | (e)     |
| 256                    |                         |                        |                     | (f)     |

Figure 3: Test Results on (a)m=8 (b)m=16 (c)m=32 (d)m=64 (e)m=128 (f)m=256
5.2. Test Results on Different Sequence Lengths

The parameters and the results are shown in Table 2 and Figure 4 respectively:

Table 2: Relevant Parameters for Sequence Lengths

| Segment Length m | Initial Value $x_0$ | Control Parameter $\alpha$ | Length N  | Figure 4 |
|------------------|---------------------|----------------------------|-----------|----------|
| 16               | 0.314               | 0.499                      | 1,000     | (a)      |
|                  |                     |                            | 10,000    | (b)      |
|                  |                     |                            | 100,000   | (c)      |
|                  |                     |                            | 1,000,000 | (d)      |
|                  |                     |                            | 10,000,000| (e)      |
|                  |                     |                            | 100,000,000| (f)    |

Figure 4: Test Results on (a)N=1,000 (b)N=10,000 (c)N=100,000 (d)N=1,000,000 (e)N=10,000,000 (f)N=100,000,000

5.3. Test Results on Different Control Parameters

The parameters and the results are shown in Table 3 and Figure 5 respectively:

Table 3: Relevant Parameters for Control Parameters

| Segment Length m | Initial Value $x_0$ | Control Parameter $\alpha$ | Length N  | Figure 5 |
|------------------|---------------------|----------------------------|-----------|----------|
| 32               | 0.314               | 0.499                      | 10,000,000| (f)      |
|                  |                     |                            | 0.6       | (g)      |
|                  |                     |                            | 0.7       | (h)      |
|                  |                     |                            | 0.8       | (i)      |
|                  |                     |                            | 0.9       | (j)      |
|                  |                     |                            | 0.999     | (k)      |

Figure 5: Test Results on (a)$\alpha=0.001$ (b)$\alpha=0.1$ (c)$\alpha=0.2$ (d)$\alpha=0.3$ (e)$\alpha=0.4$ (f)$\alpha=0.499$ (g)$\alpha=0.6$ (h)$\alpha=0.7$ (i)$\alpha=0.8$ (j)$\alpha=0.9$ (k)$\alpha=0.999$
6. Result Analysis

When only m is changed, all 2DPQ maps are shown in left-right reflection symmetry and most of the segments are concentrated at the center, regardless of the value of m. For a larger m, more concentrated distributions are shown in the central region.

When input sequences are changed, a longer sequence is shown in a more stable distribution. It seems that when N reaches millions, variant maps Figure 4 (d)-(f) are shown in similar as one distribution.

When only a is changed, three convergent points \( a = 0, a = 0.5, a = 1 \) are interesting, three closed parameter values \( a = 0.001, a = 0.499, a = 0.999 \) are chosen. It can be seen that when a approaches 0, the distribution map converges to a single point shown in Figure 5(a).

When a changes from 0 to 0.5, shown in Figure 5(a)-(f), the convergent point gradually diverges and the distribution tends to be stable, which shows that as the control parameter is close to 0.5, the map is shown the best stability of normal distributions in Figure 5(f).

As a changes from 0.5 to 1, relevant maps are changed gradually from left-right symmetric distributions to be two separated concentrations to identified two points on both 0 and 1 positions shown in Figure 5(k).

7. Conclusion

In this paper, variant maps are applied to the chaotic random sequence based on Tent map, to show the statistical distributions of Tent map systematically. The distribution is stable when the control parameter is close to 0.5 if the sequence is long enough. The convergent points of the Tent map are identified and verified with the numeric condition, which can be shown from variant maps. From what has been discussed, the variant maps can be applied to Tent chaotic map. Since this is only an initial testing work, further systematic testing and evaluations are required to explore complex behaviors of Tent chaotic sequences.

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