2D Hybrid chaos map for image security transform based on framelet and cellular automata

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Abstract

In this paper, we provide some safe ways to transfer images securely by using cryptography and steganography methods. In order to enhance the security of the image transmission, we introduce a new type of uniformly distributed 2D-hybrid chaos map based on Logistic, Sine and Tent maps, and use the cellular automata and discrete framelet transform in the proposed algorithms and also mix the position of the image pixels by apply kinds of shifts. To show that the proposed algorithms are able to resist various attacks, different types of simulation results and security analysis are used.

Keywords: Cryptography; Steganography; Color image; Chaotic system; Framelet; Cellular Automata.

1 Introduction and Preliminaries

In the last decade, with the development of the social networks as WatsApp, Facebook, Instagram and etc., the importance of information security has increased. Since images play an important role in the social networks, creating a safe environment for transferring images is important. Encryption is one of the methods for safe transfer of image. Chaotic system is one of the most important tools in cryptography. Early studies on chaotic systems for encryption has been given by [1], and in recent years, this method has been used and developed in many papers [2, 3, 4, 5]. Another tool used in cryptography is the cellular automata. This tool is introduced by Von Neumann [6, 7] and is used in various articles for cryptography [8, 9, 10]. Steganography is the other method for image security transform. This method is used in many papers for image security transform [11, 12]. The spatial domain and the

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transform domain are common methods in the steganography [13]. The Least Significant Bit (LSB) is considered as the main technique in spatial domain [14, 15, 16]. Another tool that is used in this article is the framelet transform. In 1952, the concept of frames for Hilbert spaces was defined by Duffin and Schaeffer [17]. Although the details of the framelet transform can be found in the [18], in the continue of this section, some basic relationships to this discussion are given. This type of transition has applications in various area of image processing such as deblurring, denoising [19, 20]. In the present paper, in the first step, a 2D hybrid chaotic system has been introduced, then this system and the cellular automata method have been used in encryption algorithm. In the last step, an image hidden method based on the hybrid chaotic system and the framelet transform is introduced. The following subsections provide a brief overview of the basic principles.

1.1 Hybrid Chaos Map

The basic chaotic systems have some weaknesses. The nonuniform distribution and area limit of chaos behavior are the most important of these weaknesses. Combining these types of systems is one of the best ways to minimize these weaknesses. The first attempt to combine the two systems can be referred to Logistic Tent system (LTS) [21]. Subsequently, various methods have been introduced in the articles for hybrid systems, as Tent-Sine system [22] and 2D Sine Logistic modulation map (2D SLMM) [23].

1.2 Frame Theory

Let \mathcal{H} be a separable Hilbert space. We call a sequence $F = \{f_i\}_{i \in I} \subseteq \mathcal{H}$ a frame for \mathcal{H} , if there exist two constant $A_F, B_F > 0$ such that

$$A_F \|f\|^2 \le \sum_{i \in I} |\langle f, f_i \rangle|^2 \le B_F \|f\|^2, \quad f \in \mathcal{H}.$$
 (1.1)

If in (1.1), $A_F = B_F = 1$ we say that $F = \{f_i\}_{i \in I}$ is a Parseval frame for \mathcal{H} . Let $F = \{f_i\}_{i \in I}$ be a frame for \mathcal{H} , then the operator

$$T_F: l_2(I) \to \mathcal{H}, \quad T_F(\{c_i\}_{i \in I}) = \sum_{i \in I} c_i f_i,$$

is well define and onto, also its adjoint is

$$T_F^*: \mathcal{H} \to l_2(I), \quad T_F^* f = \{\langle f, f_i \rangle\}_{i \in I}.$$

The operators T_F and T_F^* are called the synthesis and analysis operators of the frame F. By composing T_F and T_F^* , we obtain the frame operator

$$S_F: \mathcal{H} \to \mathcal{H}, \quad S_F f = \sum_{i=1}^{\infty} \langle f, f_i \rangle f_i, \quad f \in \mathcal{H}.$$

The operator S_F is a positive, self-adjoint invertible operator on \mathcal{H} with $A_F.I_{\mathcal{H}} \leq S_F \leq B_F.I_{\mathcal{H}}$, where I is the identity operator on \mathcal{H} and A_F and B_F are the lower and upper frame bounds, respectively [18]. For Parseval frame $F = \{f_i\}_{i \in I}$, we have $S_F = I_{\mathcal{H}}$.

Theorem 1.1. [18] Let $F = \{f_i\}_{i \in I}$ be a frame for \mathcal{H} with frame operator S_F . Then

$$f = \sum_{i=1}^{\infty} \langle f, S_F^{-1} f_i \rangle f_i, \quad f \in \mathcal{H},$$

and

$$f = \sum_{i=1}^{\infty} \langle f, f_i \rangle S_F^{-1} f_i, \quad f \in \mathcal{H}.$$

In the proposed algorithm for image transform, we apply particular Parseval framelet systems in $\mathcal{H}=L_2(\mathbb{R})$ that were constructed from B-spline whose refinement mask is $h_0=1/4[1,2,1]$, with two corresponding framelet masks $h_1=\sqrt{2}/4[1,0,-1]$ and $h_0=1/4[-1,2,-1]$. For the details of the image transform by using framelet, the reader can refer to [24,25].

1.3 Cellular automata

The cellular automata can be considered as a mathematical model for the discrete dynamical systems which composed of a number of cells. These cells together build a network, so that they are updated according to the special rules. In the following of this paper, the general principles used in the paper are explained and more details about this area can be found in [26, 27]. In the proposed algorithms, we use the combined cellular automata. The combined cellular automata performance is like the common cellular automata, with the difference that the combined cellular automata can assign different rule number for each cell. The different dimensions can be classified for the cellular automata as one-dimensional, two-dimensional and three-dimensional. In this paper, one-dimensional cellular automata has been used. Also for using the cellular automata, there is a need for boundary conditions. For the cellular automata, we can consider periodic, reflective, and fixed value boundaries as boundary conditions. In the proposed algorithms, periodic boundary condition as follows has been used.

$$S_{i,j}^t = S_{u,v}^t \Leftrightarrow i \equiv u \pmod{n} \text{ and } j \equiv v \pmod{m}, \tag{1.2}$$

where i, j, u and v denote cell coordinates in two-dimensional space of the size $n \times m$ and S is cell state at the time step t. Also, in this method, different neighbors are considered as the von Neumann neighborhood, the Moore neighborhood and the extended Moore neighborhood (see Fig. 1). In the proposed methods, the von Neumann neighborhood is used. The cellular automata are divided into two categories: reversible and irreversible. In the proposed key generation algorithm, irreversible cellular automata is used, and in encryption algorithm, reversible cellular automata is used.

2 Two dimensional hybrid chaotic system

In this section, we describe two-dimensional combination chaotic system based on Logistic, Sine and Tent maps. One-dimensional Logistic, Sine and Tent maps are

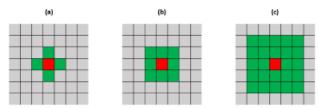


Figure 1: (a) Von Neumann neighborhood, (b) Moore neighborhood, (c) extended Moore neighborhood.

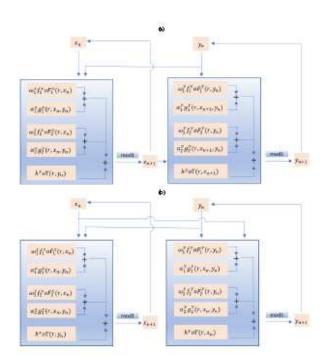


Figure 2: Proposed hybrid chaotic systems.

defined as follows [28]:

$$x_{i+1} = L(r, x_i) := r x_i (1 - x_i),$$
 (2.1)

$$x_{i+1} = S(r, x_i) := r \sin(\pi x_i)/4,$$
 (2.2)

$$x_{i+1} = T(r, x_i) := \begin{cases} rx_i/2, & \text{when } x_i < 0.5, \\ r(1 - x_i)/2, & \text{when } x_i \ge 0.5. \end{cases}$$
 (2.3)

One of the two-dimensional chaos map is the Logistic map, which has the following form [29]

$$x_{i+1} = r(3y_i + 1)x_i(1 - x_i),$$

$$y_{i+1} = r(3x_{i+1} + 1)y_i(1 - y_i).$$
(2.4)

Also in [23], 2D Sine Logistic modulation map (2D SLMM) is introduced as follows

$$x_{i+1} = \alpha(\sin(\pi y_i) + \beta)x_i(1 - x_i),$$

$$y_{i+1} = \alpha(\sin(\pi x_{i+1}) + \beta)y_i(1 - y_i).$$
(2.5)

Each of the above maps has weaknesses that we will review. Flat histogram and uniform distribution are two most important features of the ideal chaos map for encryption process. Nonuniform distribution over output series can cause weakness in the decryption attacks as statistical attack. The most important and basic chaos maps as Logistic or Tent are not well enough to estimate these conditions. However, a good combination and transformation of these maps can satisfy these conditions. In the proposed hybrid chaos map, combining, transferring and adding weights are used. In the following, two combinational algorithms are used to obtain the hybrid chaos maps. The general outline of the combinational algorithms are presented in Fig. 2. In the first step of the combinational algorithms (a) and (b), x_n and y_n are used as inputs. Regarding the combinational algorithms (see Fig. 2), it can be seen that the difference between the two algorithms is in the input of the second step. In the first algorithm, the output value of the first step (i.e. x_{n+1}) and y_n are used as input in the second step, but in the second algorithm, x_n and y_n are used as input values for the next step. The mathematical formulae for the algorithms can be expressed as follows

$$x_{i+1} := \begin{cases} \omega_1^x f_1^x \circ F_1^x(r, x_i) + \alpha_1^x g_1^x(r, x_i, y_i) \\ + h_1^x \left(\frac{(\beta_1^x - r)x_i}{2}\right) \mod 1, & \text{when } y_i < 0.5, \\ \omega_2^x f_2^x \circ F_2^x(r, x_i) + \alpha_2^x g_2^x(r, x_i, y_i) \\ + h_2^x \left(\frac{(\beta_2^x - r)(1 - x_i)}{2}\right) \mod 1, & \text{when } y_i \ge 0.5, \end{cases}$$

$$(2.6)$$

$$y_{i+1} := \begin{cases} \omega_{1}^{y} f_{1}^{y} \circ F_{1}^{y}(r, y_{i}) + \alpha_{1}^{y} g_{1}^{y}(r, \zeta, y_{i}) \\ + h_{1}^{y} (\frac{(\beta_{1}^{y} - r)\zeta}{2}) \mod 1, \text{ when } \zeta < 0.5, \\ \omega_{2}^{y} f_{2}^{y} \circ F_{2}^{y}(r, y_{i}) + \alpha_{2}^{y} g_{2}^{y}(r, \zeta, y_{i}) \\ + h_{2}^{y} (\frac{(\beta_{2}^{y} - r)(1 - \zeta)}{2}) \mod 1, \text{ when } \zeta \ge 0.5, \end{cases}$$

$$(2.7)$$

where

$$\zeta := \begin{cases} x_i, & \text{for Algorithm (a),} \\ x_{i+1}, & \text{for Algorithm (b).} \end{cases}$$
 (2.8)

In the above formula, $\omega_l^k, \alpha_l^k, \beta_l^k$ (k=x,y&l=1,2) are considered as weights, f_l^k, h_l^k (k=x,y&l=1,2) are used as combination maps and g_l^k (k=x,y&l=1,2) act as transfer maps. Also $F_l^k(k=x,y\&l=1,2)$ are chosen as an arbitrary map between Logistic and Sine maps. By choosing different values and maps for weights, combination and transfer maps, different chaotic systems are created. To investigate the properties of the proposed hybrid chaos map, we study several examples of these maps.

Consider the following cases:

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Case (i): \omega_1^x = \omega_1^y = 10, \omega_2^x = \omega_2^y = 20, \alpha_1^x = \alpha_1^y = \alpha_2^y = 2, \alpha_2^x = 7, \zeta = x_{i+1},
f_1^x(p) = \tan(p), f_2^x(p) = f_2^y(p) = \sin(p), f_2^y(p) = p,
g_1^x(r,q,p) = \exp(rp) + \exp(rq), g_2^x(r,q,p) = rp + \frac{2}{7}\exp(\pi rq),
g_1^y(r,q,p) = \tan(rq+p), g_2^y(r,q,p) = \exp(20rq),
\beta_1^x = 80, \beta_2^x = 20, \beta_1^y = 50, \beta_2^y = 30,
h_1^x(p) = \sin(2p), h_2^x(p) = \sin(4p), h_1^y(p) = \exp(2p), h_2^y(p) = \cos(4p),
F_l^k(k = x, y \& l = 1, 2) are considered as Logistic map.
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$$\begin{aligned} \textbf{Case (ii):} \ & \omega_1^x = 1, \omega_1^y = 16, \omega_2^x = 10, \omega_2^y = 20, \alpha_1^x = 15, \alpha_2^x = 7, \alpha_1^y = 2, \alpha_2^y = 14,, \zeta = x_{i+1}, \\ & f_1^x(p) = \cos(p), f_2^x(p) = \cot(p), f_2^y(p) = p, f_2^y(p) = \sin(\pi p), \\ & g_1^x(r,q,p) = rp + \frac{12}{15}\cos(rq), g_2^x(r,q,p) = -rp + \log(\pi rq), \\ & g_1^y(r,q,p) = \tan(rq+p), g_2^y(r,q,p) = \exp(20rq), \\ & \beta_1^x = 26, \beta_2^x = 2, \beta_1^y = 50, \beta_2^y = 30, \\ & h_1^x(p) = \sin(2p), h_2^x(p) = \exp(4p), h_1^y(p) = \exp(2p), h_2^y(p) = \cot(4p). \end{aligned}$$

$$\begin{aligned} \textbf{Case (iii):} \ & \omega_1^x = 10, \omega_1^y = 20, \omega_2^x = 5, \omega_2^y = 30, \alpha_1^x = 2, \alpha_2^x = 7, \alpha_1^y = 2, \alpha_2^y = 4,, \zeta = x_i, \\ & f_1^x(p) = p, f_2^x(p) = coth(p), f_2^y(p) = sin(p), f_2^y(p) = p, \\ & g_1^x(r,q,p) = \sin(rp) + 2\exp(rq), g_2^x(r,q,p) = \exp(20rq) + \sin(\pi r x), \\ & g_1^y(r,q,p) = p\tan(rq), g_2^y(r,q,p) = \cos(rq), \\ & \beta_1^x = 80, \beta_2^x = 20, \beta_1^y = 50, \beta_2^y = 3, \\ & h_1^x(p) = \sin(2p), h_2^x(p) = \cos(4p), h_1^y(p) = \sinh(2p), h_2^y(p) = 4p. \end{aligned}$$

In Case (ii) and Case (iii), F_1^x, F_2^y are considered as Logistic maps and F_2^x, F_1^y are considered as Sine maps. Results for the above cases are presented in Fig.s 3-6. In the Fig. 3, for the proposed hybrid chaos map, 2D Logistic map and 2D SLMM distribution patterns for 100 points are given. With regard to the results, it can be seen that the proposed maps have a uniform distribution compared to the other two maps. The histograms of output data are shown in Fig. 4. As seen from this figure, the histograms of 2D Logistic map and 2D SLMM are not flat, while the histograms of the proposed maps in comparison with two maps have a flat distribution. Also, cobweb plots in Fig. 5 show chaotic behavior for Case (i) and Case (ii). In this figure, 100 points generated by the chaos maps are used. Lyapunov Exponents is one of the most important value in the study of the behavior of the chaotic systems. Sensitivity to initial conditions can be studied by using this value. Different methods have been introduced in some articles to calculate this value, for example, the reader can see [30, 31, 32]. In this section, the method in [31] is used to find the Lyapunov Exponent. The relationship between Lyapunov exponent and chaotic system has been studied in various articles [33, 34]. The following sentence is stated in [34]:

"In an n dimensional dynamical system, we have n Lyapunov exponents. Each λ_k represents the divergence of k-volume (k = 1: length,k=2: area, etc.). The sign of the Lyapunov exponents indicates the behavior of nearby trajectories. A negative exponent indicates that neighboring trajectories converge to the same trajectory. A positive exponent indicates that neighboring trajectories diverge [34]".

In Fig.6, 2000 points are generated by the chaos maps are used to calculate the Lyapunov exponents. According to the results in Fig. 6, it is seen that in the larger interval, in comparison with 2D Logistic map, the proposed hybrid map has chaotic behavior. Another convenient tool for chaos study is bifurcation diagram. The bifurcation diagram can be used to show chaotic attractors. The results for bifurcation diagrams are given in Fig. 6. Considering the above discussions, it can be concluded that the proposed chaos maps have good chaotic behavior such as flat distribution, sensitivity to initial condition and unpredictability of the output points.

3 Encryption and Decryption Algorithms

In this section, the details of the encryption and decryption algorithms are given.

Remark 3.1. In the reminder of this paper, $\varpi(x_0, y_0, r, n)$ denotes the matrix $(a_{i,j}) \in \mathscr{R}^{2 \times n}$, where

$$a_{i,j} := \begin{cases} x_j, & i = 1, \\ y_j, & i = 2, \end{cases}$$

and x_j, y_j (j = 1, 2, ..., n) are chaos map outputs for initial values x_0 and y_0 . Also, $\varpi_{\tau_1,\tau_2}(x_0, y_0, r, n)$ denotes $\varpi(x_0, y_0, r, n)$, which the decimal part of the numbers of the first row is cut from the τ_1 -th decimal number to the next digits and the decimal part of the numbers of the second row is cut from the τ_2 -th decimal number to the next digits.

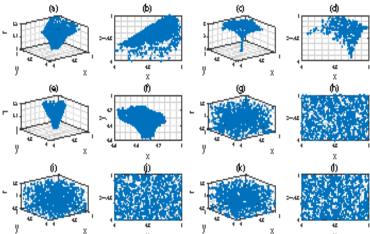


Figure 3: Distribution patterns: (a-b) 2D Logistic map, (c-d) 2D SLMM for different value of β with $\alpha = 1$, (e-f) 2D-SLMM for different value of α with $\beta = 2$, (g-h) Case (i) with r = 1.19, (i-j) Case (ii) with r = 1.19 (k-l) Case (iii) with r = 1.19.

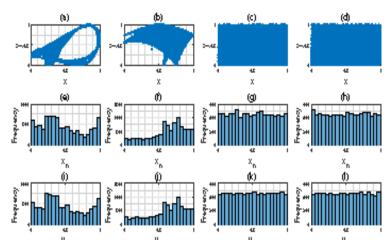
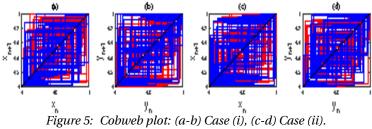


Figure 4: Distribution and histogram plots: (a-e-i) 2D-Logistic map with 1.19, (bf-j) 2D-SLMM with $\alpha=1,\beta=3$, (c-g-k) Case (i) with r=1.19, (d-h-l) Case (iii) with r = 1.19.



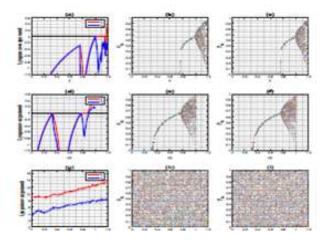


Figure 6: Lyapunov exponent and bifurcation diagram results for (a-c) 2D-Logistic map, (d-f) 2D-SLMM for $\beta = 3$, (g-i) proposed hybrid chaotic system case (iii).

Remark 3.2. The outputs of the decryption by reversible cellular automata for \mathbf{x} and \mathbf{y} as inputs in the t-1 and t times, respectively, rule number "rule" and after "rep" repetitions will be shown by $[\mathbf{x}', \mathbf{y}'] := \Phi(\mathbf{x}, \mathbf{y}, rule, rep)$. Also, the output of the decryption by irreversible cellular automata for \mathbf{x} as inputs by rule number "rule" and after "rep" repetitions will be shown by $\mathbf{x}' := \overline{\Phi}(\mathbf{x}, rule, rep)$.

3.1 Key generation function

One of the most important parts of an encryption algorithm is the key space. In this subsection, we describe the method of choosing the keys used in the algorithm. The keys used in the proposed algorithm are divided into three categories. A number of keys are selected as inputs with a fixed value, some of the keys are dependent on the input image and a number of keys are changed for each run of the algorithm. Therefore, the key space for the different images will be different. Also, for a fixed image, the key space for each implementation of the algorithm will be different. The following key space is used in the proposed algorithm.

$$key = \{r_1, r_2, x_0^1, y_0^1, x_0^2, y_0^2\},$$
 (3.1)

where parameters r_1 , r_2 are consider as the parameter r, and x_0^i , y_0^i (i=1,2) are considered as starting value in the chaotic systems. In the encryption algorithm, r_1 and r_2 are considered as fixed key, and the rest of the key space are generated by using key generation function. The following algorithm is used as key generation function (It should be noted that matlab notation formats are used in the algorithms).

Step 1. Consider a color image *I* of the size $n \times m \times 3$, a text "Text" and inputs r_1, r_2 .

If n is an odd number, then the last row of the plain image is repeated twice. In the next steps, the number of rows (i.e., n) is considered as an even number.

Step 2. Matrix Ψ of the size $n \times m \times 3$ based on the chaos map and image I are obtained as follows:

```
\begin{array}{l} \mbox{ for } k=1:3 \ \mbox{do} \\ \mbox{ for } i=1:n \ \mbox{do} \\ \mbox{ } \Psi(i,1,k)=\sum I(i,:,1)/(256\times m); \\ \mbox{ end for} \\ \mbox{ for } j=1:n/2 \ \mbox{do} \\ \mbox{ } \Psi(2j-1,2:m,k)=\varpi\big(\Psi(2j-1,1,k),\Psi(2j,1,k),r_1,m\big)(1,:); \\ \mbox{ } \Psi(2j,2:m,k)=\varpi\big(\Psi(2j-1,1,k),\Psi(2j,1,k),r_1,m\big)(2,:); \\ \mbox{ end for} \\ \mbox{ end for} \end{array}
```

Step 3. By using the BITXOR operation, vector ξ_1 and ξ_2 are found by using the following algorithm.

```
for k = 1:3 do  \Upsilon(:,:,k) = I(:,:,k) \oplus \Psi(:,:,k);  for j = 1:n-1 do  \Upsilon(j+1,:,k) = \Upsilon(j,:,k) \oplus \Upsilon(j+1,:,k);  end for for j = 1:m-1 do  \overline{\Upsilon}(:,j+1,k) = \overline{\Upsilon}(:,j,k) \oplus \overline{\Upsilon}(:,j+1,k);  end for end for  \xi_1 = \bigoplus_{i=1}^3 \Upsilon(n,:,:);   \xi_2 = \bigoplus_{i=1}^3 \overline{\Upsilon}(:,m,:);
```

Step 4. The final vectors v_1 and v_2 are calculated as follows:

$$v_1 = w(1,:), \ v_2 = w(2,:),$$

where

$$w = \omega \Big(\sum \xi_1 / (n \times 256 \times 3), \sum \xi_2 / (m \times 256 \times 3), r_2, n \Big).$$

Step 5. Key x_0^1 is considered as $(v_1 + v_2)(1, n/2)/2$.

Step 6. By using x_0^1 and input text, the vector T is made by the following conditions. First, y_0 is defined as

$$y_0 = mod \Big(\bigoplus_{i=1}^{text \ size} Text, 1 \Big),$$

– If the input text size is m, w is defined as

$$w(1,i) := \left\{ \begin{array}{ll} \varpi \left(x_0^1, y_0, r_1, [m/2] + 1 \right) (1, [i/2]), & \text{if i is odd,} \\ \\ \varpi \left(x_0^1, y_0, r_1, [m/2] + 1 \right) (2, i/2), & \text{if i is even,} \end{array} \right.$$

then, we define

$$T = \text{Text} \oplus w(1, 1:m).$$

– If the input text size is smaller than m, then remaining parts will be filled using the following vector elements:

$$w(1,i) := \left\{ \begin{array}{ll} \varpi \big(x_0^1, y_0, r_1, [text \, size - m/2] + 1 \big) (1, [i/2]), & \text{if } i \text{ is odd,} \\ \\ \varpi \big(x_0^1, y_0, r_1, [text \, size - m/2] + 1 \big) (2, i/2), & \text{if } i \text{ is even.} \end{array} \right.$$

– If the input text size is bigger than m, then T is found by

$$T = \text{Text}(1, 1: m) \oplus \varpi(x_0^1, y_0, r_1, m)(1, :) \oplus \varpi(x_0^1, y_0, r_1, m)(2, :).$$

Step 7. The binary vector *T* has been divided into 8 section and s_i , δ_i (i = 1, ..., 8) are found as:

for
$$i = 1:8$$
 do $s_i = circshift(T(1,9(i-1):8i),[1,5]);$ $\delta_i = \bigoplus_j^m s_i(1,j), (i = 1,...,8);$ end for

Step 8. The vector v_3 is found by using the following algorithm.

for
$$i=1:8$$
 do
$$rule=mod(\varpi_{10,10}(\delta_i,\delta_{i+1},r_2,1),m)(1,1);$$
 $rep=mod(\varpi_{10,10}(\delta_i,\delta_{i+1},r_2,1),m)(1,1);$ $s_i'=\overline{\Phi}(s_i,rule,rep);$ end for $v_3=(s_1',s_2',s_3',s_4',s_5',s_6',s_7',s_8')./256;$

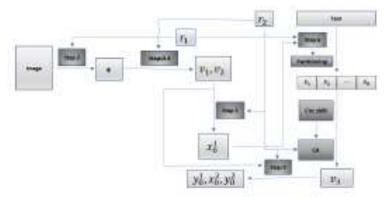


Figure 7: Proposed key generation algorithm.

In the above algorithm, we consider $\delta_9 = \delta_1$.

Step 9. y_0^2 is defined as

$$y_0^2 = mod\left(\sum_{i=1}^3 v_i, 1\right).$$

Also x_0^2 and y_0^1 are obtained as follows

$$\begin{aligned} x_0^2 &= mod \big(x_0^1 + \sum_i \varpi(x_r, y_r, r_2, m)(1, i), 1 \big), \\ y_0^1 &= mod \big(y_0^2 + \sum_i \varpi(x_r, y_r, r_2, m)(2, i), 1 \big), \end{aligned}$$

where x_r and y_r denote random numbers in the interval (0,1).

The flowchart of the proposed key function is given in Fig. 7. The above algorithm can also be performed for grayscale and binary images by making some changes. For grayscale and binary images, in steps 2-3, k is considered equal to 1. Also, in step 4, w is changed as

$$w = \omega \Big(\sum \xi_1 / (n \times 256), \sum \xi_2 / (m \times 256), r_2, n \Big).$$

To check the key space sensitivity, numerical results for the above algorithm are presented in Table 1.

Table 1:	Key space generated	by the proposed	l algorithm for v	arious images and texts.
Tubic 1.	ice opace generated	by the proposed	uigoriumii ioi v	arrous milages and texts.

	, ,	, , ,			0				
					Keys				
Image	Number of key space		Text	r_1	r_2	x_0^1	y_0^1	x_0^2	y_0^2
girl	1	Original	Hello, world	1.2000	0.7500	0.5279	0.5512	0.4427	0.2714
$(256 \times 256 \times 3)$	2	One bit changed	Hello, world	1.2000	0.7500	0.8815	0.1107	0.2561	0.7083
	3	Original	Hello world	1.2000	0.7500	0.5279	0.5127	0.7928	0.8495
	4	Original	Hello world	1.0000	0.5000	0.6080	0.0216	0.4172	0.6206
lena	5	First time run	New world	1.0000	0.5000	0.6019	0.4763	0.6122	0.995
$(256 \times 256 \times 3)$	6	Second time run	New world	1.0000	0.5000	0.6019	0.4488	0.8778	0.995
	7	Third time run	New world	1.0000	0.5000	0.6019	0.0902	0.5927	0.995
	8	One bit changed	New world	1.0000	0.5000	0.1974	0.5202	0.1163	0.6208
	9	Original	new world	1.0000	0.5000	0.6019	0.8281	0.4053	0.4990
	10	Original	newworld	1.0000	0.5000	0.6019	0.7584	0.6434	0.350
pirate	11	First time run	Hello world	0.9000	0.7000	0.5799	0.0129	0.0525	0.521
(256×256)	12	Second time run	hello world	0.9000	0.7000	0.5799	0.4723	0.9962	0.521
	13	Third time run	Hello world	0.9000	0.7000	0.5799	0.1153	0.8749	0.521
	14	One bit changed	Hello world	0.9000	0.7000	0.7090	0.2551	0.9203	0.4178
	15	Original	HellO world	0.9000	0.7000	0.5799	0.6393	0.8851	0.9122
deer	16	Original	Newworld	0.5000	1.2000	0.1543	0.9777	0.2099	0.680
(256×256)	17	Original	New worlD	0.5000	1.2000	0.1543	0.0793	0.4856	0.3792
, ,	18	Original	New world	0.5000	1.2000	0.1543	0.2571	0.9620	0.367
	19	One bit changed	New world	0.5000	1.2000	0.6818	0.5516	0.8778	0.981
	20	One bit changed	Newworld	0.5000	1.2000	0.6818	0.5968	0.3163	0.294
parthenon	21	First time run	Hello, world	0.7500	1.2000	0.6873	0.2154	0.0757	0.383
(256×256)	22	Second time run	Hello, world	0.7500	1.2000	0.6873	0.6651	0.0649	0.383

3.2 Encryption process

In this section, a color image of the size $n \times m \times 3$ is considered as an input image and the proposed algorithm for image encryption based on the 2D-hybrid chaos map and cellular automata is written as follows:

- Step1. Input image is divided into its own three plates.
- **Step2.** First the odd columns of each plates are moved to the left hand and so the even columns are moved to the right hand, then the odd rows of each plates are moved up and so the even rows are moved down. By applying this processes twice, each plates are divide into sixteen segments and each parts can see as the small size of the input image. The output images of this step is seen as blocks matrix of the size 4×4 . We show these block matrices by $I_{\rm b}^k$, k = r, g, b.
- **Step3.** The mixed block matrix $I_b^{\mathfrak{m}}$ of the size 6×8 , is achieved by using the following algorithm:

for
$$j = 1:2$$
 do
 $I_{b}^{m}(:,2j-1) = I_{b}^{r}(:,j);$
 $I_{b}^{m}(:,2j) = I_{b}^{b}(:,j);$

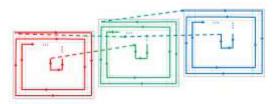


Figure 8: Proposed circular fashion shifting process.

$$I_{\rm b}^{\mathfrak{m}}(j+4,:) = \left(I_{\rm b}^{g}(2j-1,:),I_{\rm b}^{g}(2j,:)\right);$$
 end for

where $I_{\mathsf{b}}^k(i,j), (k=r,g,b,\mathfrak{m})$ denotes (i,j)-th block of the block matrix I_{b}^k .

Step4. First the odd and even columns of block matrix $I_b^{\mathfrak{m}}$, are shifted $\varpi_{1,1}(x_0,y_0,r,8)(1,j)$ units up and down, respectively, then the odd and even rows of $I_b^{\mathfrak{m}}$, are shifted $\varpi_{1,1}(x_0,y_0,r,8)(2,j)$ units right and left, respectively. We show this obtained shifted mixed matrix by $SI_b^{\mathfrak{m}}$.

Step5. By using the following algorithm, the block matrix $SI_b^{\mathfrak{m}}$ is separated into three matrices, named L, C and R:

```
\begin{split} L &= SI_{\rm b}^{\mathfrak{m}}(1:4,1:4); \\ R &= SI_{\rm b}^{\mathfrak{m}}(1:4,5:8); \\ C &= \left(SI_{\rm b}^{\mathfrak{m}}(5,1:4); SI_{\rm b}^{\mathfrak{m}}(5,5:8); SI_{\rm b}^{\mathfrak{m}}(6,1:4); SI_{\rm b}^{\mathfrak{m}}(6,5:8)\right); \end{split}
```

Step6. We take the matrices L, C, R, all together and carry out the shifting operation of the arrays of these matrices as follows:

We shift the matrix L in a circular fashion (clockwise) and starting from the array L(1,1). Note that in a shifting of the arrays of the matrix L, if we reach the last array in the shifting process, like the array $L(e_r,e_c)$, we will enter in the matrix C from the array $C(e_r,e_c)$, and continue shifting in a circular fashion (counter clockwise) and when we reach the array C(1,1), we will enter in the matrix R from the array R(1,1) and continue shifting in a circular fashion (clockwise) and if we reach the $R(e_r,e_c)$, we will enter in the matrix L from the array L(1,1) and continue shifting again clockwise (see Fig. 8). We show these three obtained matrices by L_s , C_s and R_s , respectively.

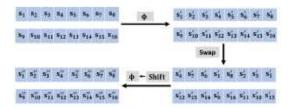


Figure 9: Process of Step 7.

- **Step7.** First of all, if n is odd number, we repeat the last row of each matrices L_s , C_s and R_s , one more time in the last of its, and so the number of the row will be even. Now, we continue our encryption algorithm by applying the cellular automata as follow (see Fig. 9):
 - (i) All of the matrices L_s , C_s and R_s are converted to the binary matrices.
 - (ii) The rows of each matrices are segmented to 8 parts and are considered as inputs of Φ in the following way. For K = L, C, R, we have

```
for i = 1: n/2 do

for k = 1: 8 do

\mathbf{x} = K_s (i, (k-1)m+1:km);

\mathbf{y} = K_s (n/2+i, (k-1)m+1:km);

rule = \omega_{1,2}(x_0, y_0, r, n/2)(1, i);

rep = \omega_{1,2}(x_0, y_0, r, n/2)(2, i);

[\mathbf{x}, \mathbf{y}] = \Phi(\mathbf{x}, \mathbf{y}, rule, rep);

end for

end for
```

(iii) The rows of each matrices was segmented to 8 parts and swaped in the following way:

```
for i = 1: n do

for k = 1: 8 do

seg_k = K_s(i, (k-1)m+1:km);

end for

[seg_1, seg_4] = swap(seg_1, seg_4);

[seg_2, seg_7] = swap(seg_2, seg_7);

[seg_3, seg_6] = swap(seg_3, seg_6);

[seg_5, seg_8] = swap(seg_5, seg_8);

end for
```

(iv) By using the function "*circshif t*", the array of the matrices is shifted to the right hand in the following way:

```
for i = 1: n/2 do

v = [K_s(i,:), K_s(n/2+i,:)];

t = \omega_{4,0}(x_0, y_0, r, n/2)(1, i);

v = circshift(v, t);

K_s(i,:) = v(1, 1: m);

K_s(n/2+i,:) = v(1, m+1: 2m);

end for
```

(v) The Step(7)-(a) is repeated one more time.

We show these three obtained matrices by L_1^r, L_1^g and L_1^b , respectively.

Step8. In this step, L_2^r, L_2^g and L_2^b are calculated by using the following algorithm

```
x_0 = (x_0^1 + x_0^2)/2; \ y_0 = (y_0^1 + y_0^2)/2; \ r = \max(r_1, r_2); Q^r = \varpi(x_0, y_0, r, n \times m)(1,:); Q^g = \varpi(x_0, y_0, r, n \times m)(2,:); Q^b = Q^r \oplus Q^g; for j = r, g, b do L_2^j = Q^j \oplus L_1^j; end for
```

Step9. At the last step, we put L_2^r, L_2^g and L_2^b on each other as plates of the final encrypted image.

Remark 3.3. To encrypt a grayscale or a binary images, we can apply the steps 2, 6, 7 and 8. Just in the step 6, if we reach the last array in the shifting process, like the array $I(e_r, e_c)$, we will come back to the first array, i.e. I(1,1), and continue shifting in a circular fashion (clockwise). As an example for step 6, in Fig. 10.b-c, an image where almost the center of it is changed to white, is used to simulate, and Fig. 10.a shows circular fashion shifting process for grayscale image.

4 Steganography

In this section, *I* is a RGB image of the size $n \times m \times 3$ and *w* is considered as a secret image of the size $n' \times m'$, where $n'm' \le nm/3$.



Figure 10: Showing the results of the mixing of an image by circular fashion shifting: (a) proposed shift, (b) given the image, (c) circular fashion shifting after 190 shifts (d) circular fashion shifting after 150000 shifts.

- **Step1.** Input a cover image I and a secret image w.
- **Step2.** Image *I* is divided into its own three plates and named them I^k , k = r, g, b.
- **Step3.** To avoid exorbitant shifts in the arrays of the matrix *I*, we use the function mod. We take the matrices I^k , k = r, g, b all together and carry out the shifting operation of the arrays of these matrices as follows: First, we shift the matrix I^r in a spiral fashion and starting from the array $I^{r}(1,1)$ and continuing along the column of the matrix and the row of the matrix, respectively. Note that in a shifting of the arrays of the matrix I^r , if we reach the last array of this matrix in the shifting process, we will enter in the matrix I^g from the array $I^g(1,1)$ and continue shifting in the same way and when we reach the last array of this matrix in the shifting process, we will enter in the matrix I^b from the array $I^b(1,1)$ and continue shifting in a the same way and if we reach the last array of this matrix, we will come back to the matrix I^r from the array $I^r(1,1)$ and continue shifting. We called this shift I. (See Fig. 11.(a)). Then, we considered these three obtained matrices again all together and start shifting similar to the above way, but starting from the last array $I^r(n,m)$ and continuing along the row of the matrix and the column of the matrix, respectively. We called this shift II (see Fig. 11.(b)). We denote these three obtained matrices by I_s^k , k = r, g, b.
- **Step4.** (i) We get the matrices $I_{s,d}^k$, k = r, g, b by applying the discrete framelet transform on the matrices I_s^k , k = r, g, b, and we cut the LL part of $I_{s,d}^k$, k = r, g, b, and named them $\hat{I}_{s,d}^k$, k = r, g, b.
 - (ii) In this step n'' and m'' are considered as row and column numbers of the output matrices in the part (i). First the odd and even columns of matrices $\hat{I}_{s,d}^k$, k = r, g, b, are shifted $\omega_{3,3}(x_0, y_0, r, \max(n', m', n))(1, j)$ units up and down, respectively, then the odd and even rows of I_b^m are shifted $\omega_{3,3}(x_0, y_0, r, \max(n', m', n))(2, j)$ units right and left, respectively.
- **Step5.** By considering the arrays of the matrices $\hat{I}_{s,d}^k$, k = r, g, b and w as interger numbers, we converted these matrices to the binary matrices named $b\hat{I}_{s,d}^k$, k = r, g, b

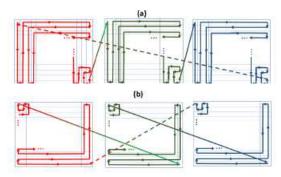


Figure 11: Proposed spiral fashion shifting process for a color image:(a) shift I, (b) shift II.

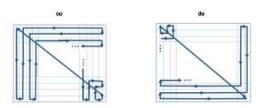


Figure 12: Proposed spiral fashion shifting process for a grayscale image.

r,g,b and bw, respectively. We set the most significant bits of the bw in the least significant bits of the $b\widehat{I}_{s,d}^k, k=r,g,b$ as the following algorithm and the obtained matrices $b\widehat{I}_{s,d}^k, k=r,g,b$, is converted to the matrices with integer arrays.

```
for i = 1: n' do

for j = 1: m' do

b\widehat{I}_{s,d}^{g}(i,8j) = bw(i,8*(j-1)+1);
b\widehat{I}_{s,d}^{g}(i,8j) = bw(i,8*(j-1)+2);
b\widehat{I}_{s,d}^{g}(i,8j-1) = bw(i,8*(j-1)+3);
b\widehat{I}_{s,d}^{b}(i,8j) = bw(i,8*(j-1)+4);
end for

end for
```

Step6. First, the LL of the $I_{s,d}^k$, k=r,g,b, is replaced by $b\widehat{I}_{s,d}^k$, k=r,g,b and then, the inverse discrete framelet transform is applied on the achieved matrices and we put these matrices as the plates of the final image.

Similarly, we can do all the steps relevant to the forward part in the opposite way to get the secret image.

Remark 4.1. For the steganography process of a grayscale or a binary images in a grayscale images, we can apply the steps 1, 3, 4, 5 and 6. Just in the steps 3 and 6, if we reach the last array in the shifting process, we will come back to the first array, i.e. I(1,1), and continue shifting in a spiral fashion (see Fig. 12). (For more examples, we refer to Appendix section). Also, step 5 for grayscale in grayscale and binary in grayscale are as follows:

- Grayscale in grayscale:

```
 \begin{aligned} & \textit{for } i = 1 : n' \; \textit{do} \\ & \textit{for } j = 1 : m' \; \textit{do} \\ & \textit{for } k = 1 : 4 \; \textit{do} \\ & b \widehat{I}_{s,d}(i,8j-4+k) = b w(i,8*(j-1)+k); \\ & \textit{end for} \\ & \textit{end for} \\ & \textit{end for} \end{aligned}
```

−Binary in grayscale:

```
egin{aligned} & 	extbf{for} \ i = 1 : n' \ 	extbf{do} \ & 	extbf{for} \ j = 1 : m' \ 	extbf{do} \ & 	extbf{b} \widehat{I}_{s,d}(i,8j) = w(i,j); \ & 	extbf{end} \ 	extbf{for} \ & 	extbf{end} \ 	extbf{for} \end{aligned}
```

5 Numerical experiments and security analysis

In this section, simulation results for the proposed algorithms are given. In the numerical results for space key, we use Table 1. Note that, case (iii) is used as the chaos map in the proposed algorithm. Also, in order to show the performance of the proposed algorithms, the results are compared with the results of other papers.

5.1 Simulation results of the encryption algorithm

The security analysis of the proposed algorithm are studied in this subsection. Also, the results for encryption proposed algorithm are given in Fig. 13.

5.1.1 Security key space

One of the most important components of the encryption algorithm is key space. The key space in the proposed algorithm has a dynamic structure i.e., when a bit of the input image is changed, the key space changes, also, a different key space is

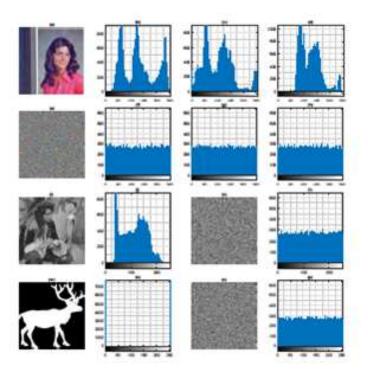


Figure 13: Original images and encryption results and their histograms: (a)-(d) girl original image, (e)-(h) encrypted girl image, (i)-(j) pirate original image, (k)-(l) encrypted pirate image, (m)-(n) deer original image, (o)-(p) encrypted deer image.

used for each algorithm implementation. In Fig. 14, the proposed algorithm is run twice on the parthenon image. In these results, the key spaces 21 and 22 from Table 1 are used. To show the difference between the two images, pixel-to-pixel difference is given in Fig. 14. The results show, the two images are very different. Therefore, it can be concluded that the proposed algorithm is able to resist the chosen-plain text attack. Also, key size is important item in key space. Key size plays an important role in attacks. The relationship between key size and brute force attack has been studied in [35]. According to this study, if key space is larger than 2^{100} , then the algorithm can withstand the brute force attack. The key space in the proposed algorithm consists of six keys. Therefore, if in the image encryption algorithm, the precision of the key size is equal to 10^{-15} , then the key space is $10^{90} = 2^{90\log_2 10} \approx 2^{298}$, therefore the key space for resistance against the brute force attack will be sufficient enough. As last study in this subsection, the key space changes is studied in the decryption process. By applying small changes on the key space 5 from Table 1, two keys are created as follows

$$\overline{K}_5 = \{1, 0.5, 0.6019 + 10^{-15}, 0.4763, 0.6122, 0.9951\},\$$

$$\widehat{K}_5 = \{1, 0.5, 0.6019, 0.4763, 0.6122 + 10^{-15}, 0.9951\},\$$

In Fig. 15.a, lena image is encrypted by K_5 (key space number 5, from Table 1), in Fig. 15.b, the decryption process is run by K_5 , also in the Fig.s15.c-d, the decryption process are run by \overline{K}_5 and \widehat{K}_5 , respectively. Therefore by small changes in the keys, we can not decrypt the image. Table 7 shows that the key space size of the proposed algorithm is bigger than the key space size of the compared algorithms.

Table 2: Comparison of the key space size.

	Proposed algorithm	Ref.[36]	Ref.[37]	Ref.[38]
Key space size	$10^{90} \approx 2^{298}$	1084	2^{120}	10^{56}

5.1.2 Statistical analysis

In this subsection, statistical analysis including correlation values, information entropy and histogram analysis are studied. The correlation values for image can be found by using following formula from [39]

$$C_{x,y} = \frac{E(x - \mu_x)(y - \mu_y)}{\sigma_x \sigma_y},$$
(5.1)

where E,μ and σ denote expectation, mean values and standard deviation, respectively. The correlation values for different images are given in Table 3. By using results in this table, we can see that the original images have high correlation values, while encrypted images have low correlation values. Also, Fig. 16 shows correlation

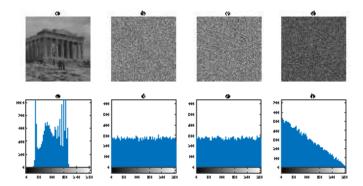


Figure 14: Results for image and histogram: (a-e) the plain image, (b-f) the first encrypted image, (c-g) the second encrypted image, (d-h) the pixel-to-pixel image.

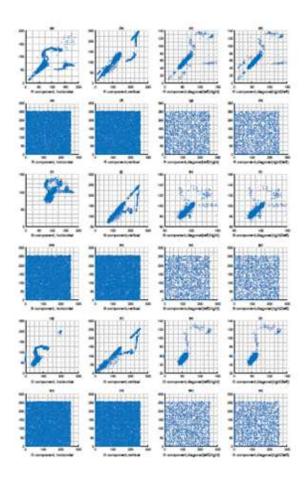


Figure 15: Decryption process for different keys.

distributions for the original images and encrypted images. Comparison results of the proposed algorithm with the algorithm in [40], that are given in Table 4, illustrates the advantage of the proposed algorithm. The information entropy is calculated by

$$H(k) = -\sum_{i=0}^{w-1} P(k_i) \log_2 P(k_i),$$
 (5.2)

where w and P represent the gray level and the probability, respectively. The values of this expression are in the interval [0,8], and the ideal value for the encrypted image is equal to 8. Numerical results of the information entropy for different images are shown in Table 3. By using these results, we can say that the information entropy for encrypted images are close to the ideal value. Also, the comparison results for the information entropy are presented in Table 5. Histogram diagram for original and encrypted images are given in Fig. 13. In this figure, we can see that encrypted images have a flat distribution compared to the original images.



 $Figure\ 16:\ Correlation\ of\ neighbourhood\ pixels\ at\ different\ directions\ before\ and\ after\ encryption\ of\ girl\ image.$

Table 3: Numerical results for UACI, NPCR, Information entropies, Correlation coefficients for different images.

images.					Correlation coefficients			
Image		UACI	NPCR	Information entropies	Horizontal	Vertical	Diagonal (lower left to top right)	Diagonal (lower right to top left)
	R			7.2549	0.9786	0.9879	0.9688	0.9684
girl-original	G			7.2704	0.9660	0.9820	0.9515	0.9507
	В			6.7825	0.9523	0.9718	0.9307	0.9306
	R	33.5542	99.5850	7.9971	0.0028	0.0088	-0.0071	0.0048
girl-encrypted	G	33.3360	99.5804	7.9973	-0.0014	-0.0009	-0.0006	-0.0047
	В	33.2780	99.5895	7.9972	-0.0004	-0.0034	-0.0086	-0.0050
pirate-original				7.2887	0.9434	0.9564	0.9201	0.9134
pirate-encrypted		33.5110	99.5636	7.9971	-0.0014	0.0006	-0.0014	-0.0036
deer-original				0.9102	0.9463	0.9538	0.9274	0.9317
deer-encrypted		33.1739	99.6170	7.9975	0.0055	-0.0009	-0.0004	-0.0068

 $Table~4:~Correlation~coefficients~for~lena~(256\times256\times3)~by~10,000~pixels~pairs~of~adjacent~positions.$

		Horizontal	Vertical	Diagonal
Proposed algorithm	R	0.0038	0.0094	0.002
	G	-0.0062	-0.0042	-0.0016
	В	0.0109	-0.0054	-0.00695
Algorithm in [40]	R	-0.0127	0.0067	0.0060
	G	-0.0075	-0.0068	-0.0078
	В	-0.0007	0.0042	0.0026

Table 5: Information entropy of the encrypted color image for lena $(256 \times 256 \times 3)$.

	71 D		<u> </u>
	R	G	В
Proposed algorithm	7.9970	7.9973	7.9973
Algorithm in [41]	7.9891	7.9898	7.9899
Algorithm in [42]	7.9896	7.9893	7.9896
Algorithm in [43]	7.9893	7.9896	7.9903
Algorithm in [44]	7.9874	7.9872	7.9866
Algorithm in [45]	7.9278	7.9744	7.9705

5.1.3 Sensitivity analysis

NPCR and UACI are two important tools in sensitivity analysis. NPCR (i.e., number of pixels change rate) represents the number of pixels change rate while one pixel of plain image changed and UACI (i.e., unified average changing intensity) denotes the average intensity of difference between the plain image and encrypted image. A good study based on statistical discussions for these values is presented in [46]. The ideal value for NPCR and UACI are considered as 100% and $33.\overline{33}$ %, respectively. Using these values, the resistance against differential and plaint text attacks can be studied. When these values are close to the ideal values, we can say that the encryption algorithm is more sensitive to the changing of plain image, therefore the

encryption algorithm can effectively resist plain text and differential attacks. By using numerical results in Table 3, we can see that results are close to the ideal value. As well as, the comparison results for NPCR and UACI values have been tabulated in Table 6. In computing, NPCR and UACI are considered as follows

$$NPCR = \frac{\sum_{i,j} D(i,j)}{m \times n} \times 100\%,$$

$$UACI = \frac{\sum_{i,j} |C_1(i,j) - C_2(i,j)|}{255 \times m \times n} \times 100\%,$$
(5.4)

$$UACI = \frac{\sum_{i,j} |C_1(i,j) - C_2(i,j)|}{255 \times m \times n} \times 100\%,$$
(5.4)

with

$$D(i,j) := \begin{cases} 1, & \text{when } C_1(i,j) \neq C_2(i,j), \\ 0, & \text{when } C_1(i,j) = C_2(i,j), \end{cases}$$
 (5.5)

where, C_1 and C_2 are considered as the encrypted image before and after the original image is changed.

Table 6: UACI and NPCR of the encrypted color image for lena $(256 \times 256 \times 3)$.

		R	G	В
NPCR	Proposed algorithm	99.59260	99.63990	99.60020
	Algorithm in [47]	99.58649	99.21722	98.84796
	Algorithm in [48]	99.42	99.60	99.54
	Algorithm in [49]	99.26	99.45	99.13
UACI	Proposed Algorithm	33.49960	33.41470	33.41000
	Algorithm in [47]	33.48347	33.46399	33.26891
	Algorithm in [48]	27.78	27.66	24.94
	Algorithm in [49]	21.41	23.42	15.08

5.1.4 Noise and data loss attacks

The purpose of some attacks is not decryption, and the purpose is to destroy information, like noise and data loss attacks. In these attacks, parts of the information may be lost. In the proposed algorithm, the location of pixels are changed, and this can be a response to these attacks. The simulation results for these attacks are given in Fig.s 17-18. In the Fig. 17, the results show that the proposed algorithm has better results than the algorithm in [50]. Also, in Fig. 18, for different values of noise density, results are presented. By using the results, it is seen that the encrypted images are recognizable. Therefore, we can say that the proposed algorithm can resist noise and data loss attacks.



Figure 17: Results of the data loss attacks: (a),(d) attacked images, (b),(e) results of the algorithm in [50], (c),(f) results of the proposed algorithm.

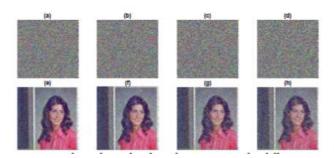


Figure 18: Noise attack results with salt and pepper noise for different noise density: (a)-(d) noise attacks for encrypted image with noise density = 0.08, 0.1, 0.15, 0.2, respectively. (e)-(h) decryption results for noise attacks with noise density = 0.08, 0.1, 0.15, 0.2, respectively.

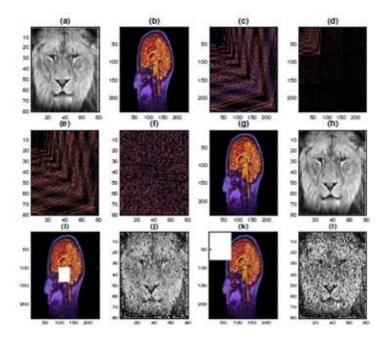


Figure 19: Results for steganography algorithm: (a) secret image, (b) cover image, (c) output image of the step 3, (d)-(f) output images of the step 4, (g) stego image, (h) decrypted secret image, (i), (k) data loss attack images, (j), (l)decrypted secret images.

5.2 Simulation results of the steganography algorithm

In this section, the results of the proposed steganography algorithm are given. In Fig. 19, we apply the proposed algorithm on $lion(80 \times 80)$ and MRI(250 × 250) images as secret and cover images, respectively. According to the proposed algorithm, first of all, we shift the cover image by using two types of the spiral fashion shifting (I and II) after 1000 shifts (Fig. 19.c). Next, we use discrete framlete transform (DFT)(Fig. 19.d). Next, we cut the LL part of DFT (Fig. 19.e). Then in Fig. 19.f, the odd and even columns are shifted up and down and the odd and even rows are shifted right and left by applying $x_0 = 0.1$, $y_0 = 0.3$ and r = 0.2 as input values for the proposed chaos map (for details, we refer to step4 (ii)). Note that we can use the proposed key generation function to get x_0 , y_0 and r. Finally, we set the most significant bits of the secret image in the least significant bits of the obtained image from the previous step (see step5). By doing the above processes conversely, stego image (Fig. 19.g) is obtained. To show the resistance of the proposed algorithm against the data loss attacks, the simulation results for data loss attacks are given in Fig.s 19.i-l, where once the 40×40 size of the central part and once again in the size of the secret image, i.e. 80×80 , of the corner part of the stego image are removed.

Tab	Table 7: PSNR values for steganography algorithm					
	Fig. 19.g	Fig. 19.h	Fig. 19.j	Fig. 19.l		
PSNR	48.1594	29.2125	16.5766	14.6422		

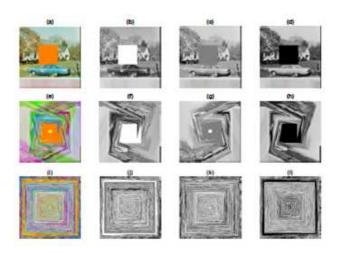


Figure 20: Showing the results of the mixing of an image by circular fashion shifting: (a-d) original image, (e-h) circular fashion shifting after 190 shifts (i-l) circular fashion shifting after 150000 shifts.

6 Appendix

In this section, to make the visualization of the proposed steps easier, we present some examples.

Fig. 20 represents the circular fashion shifting in the proposed encryption process. For better understanding, we apply this algorithm on an image of the size 256×256 , that almost the center of this changed to orange color.

7 Conclusion

In this paper, by introducing a 2D-hybride chaos map, safe ways to transfer images securely by using cryptography and steganography methods based on this chaos map are presented. Also, in the proposed algorithm, framelet, cellular automata and kinds of shifts have been used to increase the resistance against different attacks. By using simulation, results, we can say that the proposed algorithms can effectively resist differential, statistical, noise, data loss and chosen-plain text attacks.

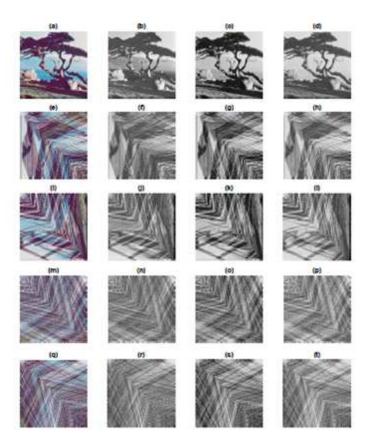


Figure 21: Showing the results of the mixing of the image by spiral fashion shifting: (a-d) original image and its own component, (e-h) shifting by I type, (i-l) shifting by II type, (m-p) shifting by I and II types, respectively, (q-t) shifting by II and I types, respectively.

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