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A New Steganographic Algorithm Based on Multi Directional PVD and Modified LSB

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Steganographic techniques can be utilized to conceal data within digital images with small or invisible changes in the perceived appearance of the image. Generally, five main objectives are used to assess the performance of steganographic algorithms which include embedding capacity, imperceptibility, security, robustness and complexity. However, steganographic algorithms hardly take all of these factors into account. In this paper, a novel steganographic algorithm for digital images is proposed based on the pixel-value differencing (PVD) and modified least-significant bit (LSB) substitution (MDPVD-MLSB) techniques to address most of aforementioned objectives. Although there are many techniques for concealing data within pixels, the restricting factor is always the amount of bits adjusted in every pixel. Therefore, the main contributions of this paper aim to achieve a balance between the amount of embedded data, the level of acceptable distortion, as well as providing high level of security. The performance of this algorithm has been extensively evaluated in terms of embedding capacity, peak signal-to-noise ratio (PSNR) and structural similarity index measure (SSIM). Simulation results and comparisons with six relevant algorithms are presented to demonstrate the effectiveness of this proposed algorithm.

KEYWORDS: Tri-way pixel-value differencing (TPVD), Octa pixel-value differencing (Octa-PVD), least-significant bit (LSB), optimal pixel adjustment process (OPAP), embedding capacity, imperceptibility.

Introduction

Internet has become a vital part of everyone's lives, where anyone can send and receive data from any

place in the world at any time. Wireless connection is getting popular since it is more convenient than

wired Ethernet cables keeping in mind that wireless networks have numerous obstacles, among them low bandwidth, insecure links, and high error rate [1-9]. One of the primary issues of sending data over the Internet is security [10-11]. The ability to communicate securely is an important concern to organizations, people, and government. Therefore, it becomes very crucial to take data security into consideration as one of the most fundamental factors that requires attention during data transmission process. Generally, there are two popular approaches to secure data [12]. The first one is encryption, in which the original text is transformed into a cipher-text via certain algorithms. Thus, encryption changes the text into unreadable and incomprehensible text, which makes it suspicious enough to attract attackers' attention. Data Encryption Standard (DES), Advanced Encryption Standard (AES) and Rivest-Shamir-Adleman cryptosystem (RSA) are some of the common encryption methods [13]. The other approach is data hiding which is basically classified into two parts, namely, watermarking and steganography [14-17]. The main difference between them is that the former secures the carrier-object along with copyright information, while the later secures the embedded message into the carrier-object only [18-24]. Interestingly, there are other differences between the two methods based on the required performance objectives. For instance, watermarking is more sensitive to robustness than steganography. On the other hand, the steganography is more sensitive to capacity than watermarking. The security in watermarking lies in the difficulty of removing the watermark while in steganography, it lies in the difficulty of detecting/extracting the embedded data. Additionally, watermarking techniques are made between a sender and numerous receivers, whereas steganographic techniques are made between a sender and only one receiver [14].

This paper proposes an efficient and robust dynamic data hiding algorithm which mainly aims to improve the embedding capacity of the secret data, enhance the visual quality of stego-image, and make steganalysis a very hard task. The proposed algorithm is based on multi-directional pixel-value differencing (MDPVD) and modified least-significant bit (MLSB). However, the rest of the paper is organized as follows. In Section 2, a literature review of the most related works in the field is presented. Section 3 details the proposed algorithm. Experimental results and discussions are

provided in Section 4. Section 5 concludes the work and provides future possible directions.

Related work

Image steganographic algorithms, which have been discussed in literature, can be classified based on the embedding domain into two main classes: spatial domain and transform domain [14, 25-27]. The spatial domain algorithms are most frequently used because of their good concealment, great capability to hide information, and ease of realization [28]. In spatial domain methods, a steganographer modifies the pixel values of the host image directly [12]. LSB and PVD are the most common algorithms that essentially belong to this class [29-30]. In the transform domain algorithms, a steganographer embeds information into the coefficients of some transformed version of the host image [31]. The cover image can be converted into its transform domain by using one of the wavelet transformations such as: Discrete Cosine Transform (DCT), Discrete Wavelet Transform (DWT), and Fast Fourier Transform (FFT). Data insertion performed in the transform domain is greatly used for robust watermarking [25]. JSteg, F3, F4, F5, and Outguess methods are known steganographic algorithms that operate in the transform domain [25, 32].

The LSB substitution utilizes the least bits of a pixel value in the cover-image for embedding secret data. Many steganographic algorithms hide a large amount of secret data in the first least significant bits of the cover-image pixels. Because of the weak sensitivity of the human visual system (HVS), the existence of the hidden secret information can be imperceptible. The quality of the stego-image produced by simple LSB substitution may not be acceptable if a large amount of LSBs is used for data embedding. As an example, a stego-image can achieve a peak signal-to-noise ratio (PSNR) as low as 31.78 dB by using a simple LSB-4 replacement [29, 32-34]. The use of the optimal pixel adjustment process (OPAP) improves the perceptual quality of the stego-image when compared to using the simple LSB substitution method alone. The OPAP method was proposed in [29] and its idea has been described and utilized in many research papers [35-39].

It is worth stating that one of the most widespread approaches that become a base for a large amount

of modern algorithms for secret data hiding is PVD, which can provide high embedding capacity and extremely good stego-image visual quality. For instance, the PVD method can achieve the capacity of 50,960 bytes (1.555 bpp) for Lena test image and its stego-image has a PSNR of 41.79 dB [34]. The main idea behind PVD is to use the difference between two consecutive pixels of a gray scale image to hide data. In [40], pixel-value differencing is used to distinguish between edge areas and smooth areas. Consequently, the capacity of embedded data in edge areas is higher than that of smooth areas. Wu and Tsai [40] proposed two types of quantization range table based on the range width of the power of two. The first range has the widths of {2, 2, 4, 4, 4, 8, 8, 32, 32, 64, 64} and is used to provide high imperceptibility or higher values of PSNR. The second range has the widths of {8, 8, 16, 32, 64, 128} and is used to provide high embedding capacity. It is worth mentioning that there are rare studies that aim to design new range tables. Tseng and Leng [41] proposed a PVD-based algorithm that mainly includes a new quantization range table based on the perfect square number, in order to obtain the secret data bits by using the difference value between consecutive pixels. They partition each range into two subranges for embedding variable number of secret data bits. After determining the perfect square number which belongs to the interval [1, 16], the ranges produced by this method are as follows {[0,1], [2,5], [6,11], [12,19], [20,29], [30,41], [42,55], [56,71], [72,89], [90,109], [110,131], [132,155], [156,181], [182,209], [210,239], [240,255]}.

In spite of PVD simplicity (i.e., being efficient in achieving large embedding capacity and extremely good stego-image visual quality) it has a limitation with respect to the capacity. In particular, the PVD does not make full use of edge areas. Motivated by the original PVD in [40], Chang et al [30] proposed a PVD-based scheme called tri-way pixel-value differencing (TPVD), which uses three different directional edges instead of one directional edge to eliminate the capacity limitation of the PVD method. Furthermore, the authors used the reference point selection and adaptive conditions in order to improve the quality of the stego-image. Based on TPVD, a data hiding algorithm was proposed in [42] in which three-directional PVD method for gray images is used. Unlike PVD and TPVD, the position of base pixel in this method is variable and depends on the pixel value of each group. Based on an index function, Jung and Yoo [43] pro-

posed a high-capacity steganographic approach. In this method, the cover-image is partitioned into $B \times B$ sub-blocks and the base pixel is computed according to the index function. Rather than fixing the position of the base pixel, the position in this method depends on an index. The indexes must satisfy the condition: $0 \leq x, y \leq B-1$. After applying the index function to select the base pixel, the pixel values are sorted in ascending/descending order according to the results of the index function. Exploiting Modification Direction (EMD) method is devised in [44]. The importance of the EMD scheme lies in providing a good quality of stego-image with PSNR of more than 52 dB, since at most only one cover pixel in each pixel group needs to be incremented or decremented by one. Taking into account the EMD method, Shen and Huang [45] proposed a new scheme to enhance the embedding capacity and hide digits in any-ary notation adaptively by using the absolute difference value of pixel pairs.

Similar to TPVD, Thaneker and Pawar [46] proposed a new PVD scheme with eight directional edges, which is called Octa-PVD, to achieve a better capacity than that in PVD and TPVD. Unfortunately, it is noticed that even though this research used more directional edges, the image distortion increases dramatically. Therefore, the big challenge in PVD-based steganographic schemes, which have multi directional edges, is to achieve a balance between the amount of both the embedded data and the acceptable distortion, as well as provide high level of security.

In summary, an efficient and robust dynamic algorithm of data hiding is proposed in this paper. This algorithm aims to improve the embedding capacity of the secret data, enhance the visual quality of stego-image, and make steganalysis a very hard task. The proposed algorithm is based on multi-directional PVD and modified LSB. The idea of MLSB is to increase or decrease the pixel gray value after embedding, using simple LSB replacement by 2^k (where k is the number of secret data bits that have to be embedded in LSBs of each pixel in a cover image) in order to enhance the image quality [29, 42].

The proposed algorithm

Since this study is based on MDPVD, the algorithm is divided into two separate schemes. These schemes

are called (Quinary-PVD-MLSB) and (Octa-PVD-MLSB) because the cover-image is partitioned either into 2×3 pixel blocks, each of which has 5 pairs to embed secret data, or into 3×3 pixel blocks, each of which has 8 pairs to embed secret data. On the other hand, three branch conditions are proposed that permit automatic switching between MDPVD and MLSB (one of them can be selected only in the embedding and extracting procedures). In fact, these branch conditions can certainly reduce distortion caused by the offsetting of pairs process. If the selected branch condition is satisfied, then the current block can raise the distortion. Thus, this block must be embedded using MLSB to avoid any degradation in the visual quality of stego-image. The details about the proposed branch conditions are described in Section 3.1.

To protect the embedded secret data when using MLSB embedding, a secret key (*SK*) is used to insert secret data bits in different pixel indices according to the generated integer set N_s . N_s is generated using the set-generation function $H_s(SK, N_{bp})$, where N_{bp} is the number of block pixels, $N_s = \{N_{si} \mid i = 1, 2, \dots, N_{bp}\}$ and $SK \in [1, N_{bp}!]$ (i.e. $SK \in [1, 720]$ in Quinary-PVD-MLSB and $SK \in [1, 362880]$ in Octa-PVD-MLSB). In other words, based on N_{bp} , H_s set function will generate ($N_{bp}!$) possible permutations and according to the selected *SK* value, H_s will generate the required N_s . Each element in the N_s set is unique and its value falls within the range $[1, N_{bp}]$.

Generally, the proposed algorithm has three main phases: the partition phase, the embedding phase, and the extracting phase, which will be described in details in Sections 3.2, 3.3 and 3.4, respectively.

Branch conditions to reduce distortion

Although both Quinary-PVD and Octa-PVD can hide large amount of secret data if they are utilized without using MLSB, embedding such large amount of data can clearly cause a very poor visual quality due to the offsetting of pairs process. Therefore, there is a need to determine some branch conditions that allow automatic switching between Quinary-PVD and MLSB or between Octa-PVD and MLSB. Three branch conditions are designed and only one of them can be used at a time according to the requirements of the practical applications such as the embedding capacity and the visual quality of the stego-image. These conditions are called BC_1 , BC_2 and BC_3 . They are as the following:

- 1 At least one of the difference values is equal or greater than 8 (i.e. $\exists D \geq 8$).
- 2 At least one of the difference values is equal or greater than 16 (i.e. $\exists D \geq 16$).
- 3 At least one of the difference values is equal or greater than 32 (i.e. $\exists D \geq 32$).

Where $D = \{d_i \mid i = 0, \dots, N_{bp} - 2\}$. If the selected branch condition from the above three conditions is satisfied, the current block results in higher distortion if MDPVD is used. Therefore, MLSB is used to individually embed each pixel in the block.

Partition phase

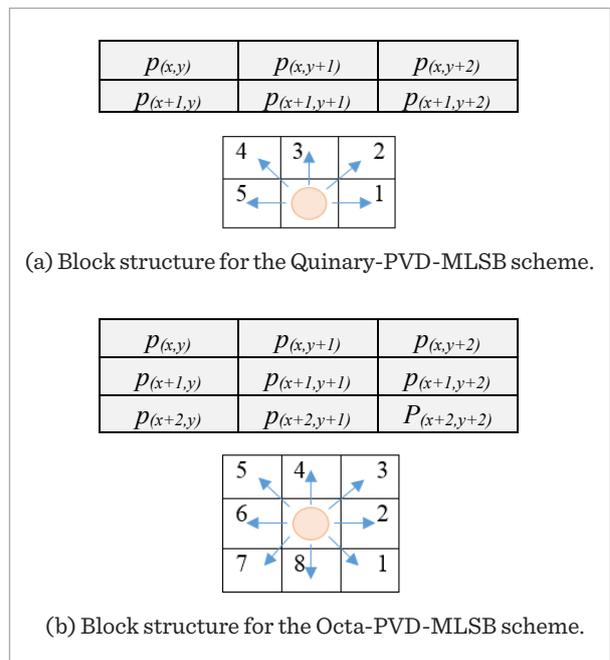
In the proposed algorithm, the cover-image (*CI*) is partitioned into 2×3 or 3×3 non-overlapping pixel blocks B_i , each of which has six or nine pixels in raster scan order or zig-zag scan order such that:

$$CI = \left\{ B_i \mid i = 1, 2, \dots, \frac{M \times N}{6} \right\} \text{ Or } CI = \left\{ B_i \mid i = 1, 2, \dots, \frac{M \times N}{9} \right\}. \quad (1)$$

Fig. 1 (a, b) shows the block structures for the two schemes of the proposed algorithm.

Figure 1

(a, b). Block structures in MDPVD-MLSB algorithm



Embedding phase

The embedding phase of our algorithm when the Octa-PVD-MLSB scheme is considered will be explained. As shown in Fig. 1. (b), each 3×3 block includes nine pixels $\{p_{(x,y)}, p_{(x,y+1)}, p_{(x,y+2)}, p_{(x+1,y)}, p_{(x+1,y+1)}, p_{(x+1,y+2)}, p_{(x+2,y)}, p_{(x+2,y+1)}, p_{(x+2,y+2)}\}$ where x and y are the pixel location in the cover image. Let $p_{(x+1,y+1)}$ be the beginning point, then eight pixel pairs can be formed by pairing the beginning point $p_{(x+1,y+1)}$ with each of its eight neighbors in the block. These eight pixel pairs are denoted by $\{P_0, P_1, P_2, P_3, P_4, P_5, P_6, P_7\}$. When using Octa-PVD to embed secret data bits, each pair has a new difference value d_i ($0 \leq i \leq 7$) and modified pixel pair P'_i ($0 \leq i \leq 7$). Consequently, each pair has new pixel values which are not the same as their original ones. That is, there are eight different values for the beginning point $p_{(x+1,y+1)}$. However, after finishing the Octa-PVD embedding process, only one value for the beginning point can be used. Thus, one of the eight pairs is selected as the optimal reference point to offset the other seven pixel values. Therefore, an approach to select the optimal reference point in the proposed algorithm is introduced. In addition, the shifting function is added to resolve the falling-off-boundary problem resulting from Octa-PVD embedding process. The details about both the optimal reference point selection approach and shifting function are left to be described later in subsections 3.3.1 and 3.3.2, respectively.

The embedding phase for the proposed algorithm uses the same equations that are employed in [29, 40, 46] with some modifications to suit the proposed algorithm. These modifications include the directions of pixel pairs. From the implementation of Octa-PVD, it is found that the best directions are those shown in Fig. 1 (b) in order to avoid excessive degradation in the stego-image. The details for embedding the secret data bits dynamically in each block B_i that contains nine pixels from a cover image, are described in the following steps:

Input: A $W \times H$ grayscale cover-image CI , secret data S , range table R and secret key SK .

Output: A $W \times H$ stego-image SI .

Step 1. The secret data S is converted to form binary bitstream S' .

Step 2. As explained beforehand, $H_s(SK, N_{bp})$ function is used to generate integer set N_s . According to the pixel indices in N_s , the secret data bits are embed-

ded in block pixels. For instance, if it is assumed that $N_{bp} = 9$ pixels, then all possible permutations for N_s will be as shown in Table 1. Likewise, if it is assumed that $SK = 40317$, then $H_s(40317, 9)$ function will generate the following $N_s = [7, 4, 8, 6, 5, 1, 9, 3, 2]$. Therefore, the MLSB embedding procedure begins with the pixel that has the index of 7 and ends with the pixel that has the index of 2. The index of pixels is assumed as shown in Fig. 2.

Table 1

All possible permutations for N_s

SK	Permutations
1	{1, 2, 3, 4, 5, 6, 7, 8, 9}
2	{2, 1, 3, 4, 5, 6, 7, 8, 9}
.	.
.	.
40317	{7, 4, 8, 6, 5, 1, 9, 3, 2}
.	.
.	.
362880	{9, 8, 7, 6, 5, 4, 3, 2, 1}

Figure 2

The index of pixels in block B_i of Octa-PVD-MLSB scheme

1	2	3
4	5	6
7	8	9

Step 3. Select one of the branch conditions BC_i ($1 \leq i \leq 3$), which we described previously in section 3.1.

Step 4. Calculate the difference values for the eight pixel pairs $\{P_0, P_1, P_2, P_3, P_4, P_5, P_6, P_7\}$ where:

$$\begin{aligned}
 P_0 &= (p_{(x+1,y+1)}, p_{(x+2,y+2)}), \\
 P_1 &= (p_{(x+1,y+1)}, p_{(x+1,y+2)}), \\
 P_2 &= (p_{(x+1,y+1)}, p_{(x,y+2)}), \\
 P_3 &= (p_{(x+1,y+1)}, p_{(x,y+1)}), \\
 P_4 &= (p_{(x+1,y+1)}, p_{(x,y)}), \\
 P_5 &= (p_{(x+1,y+1)}, p_{(x+1,y)}), \\
 P_6 &= (p_{(x+1,y+1)}, p_{(x+2,y)}), \\
 P_7 &= (p_{(x+1,y+1)}, p_{(x+2,y+1)}).
 \end{aligned} \tag{2}$$

Let the difference value be d_i , where $0 \leq i \leq 7$, then the difference values for all pixel pairs in the block B_i can be calculated as follows:

$$\begin{aligned} d_0 &= p_{(x+2, y+2)} - p_{(x+1, y+1)} \\ d_1 &= p_{(x+1, y+2)} - p_{(x+1, y+1)} \\ d_2 &= p_{(x, y+2)} - p_{(x+1, y+1)} \\ d_3 &= p_{(x, y+1)} - p_{(x+1, y+1)} \\ d_4 &= p_{(x, y)} - p_{(x+1, y+1)} \\ d_5 &= p_{(x+1, y)} - p_{(x+1, y+1)} \\ d_6 &= p_{(x+2, y)} - p_{(x+1, y+1)} \\ d_7 &= p_{(x+2, y+1)} - p_{(x+1, y+1)}. \end{aligned} \quad (3)$$

Step 5. The range table that is shown in Fig. 3 is used. It consists of six contiguous sub-ranges R_j where ($j=1..6$). In other words, $R = \{R_j = [l_j, u_j]\} = \{[0, 7], [8, 15], [16, 31], [32, 63], [64, 127], [128, 255]\}$. Thus, for a given difference value $|d_i|$, locate the suitable sub-range R_j in the designed range table which has values from 0 to 255, where l_j and u_j are the lower and upper bounds, respectively.

Step 6. If the selected branch condition is satisfied, apply Steps from 7 to 10 to embed secret data using MLSB. Otherwise, go to Step 11 to embed secret data bits using MDPVD.

Figure 3

Range table proposed in [40]

$R_1 = [0, 7]$	$t_1 = 3$
$R_2 = [8, 15]$	$t_2 = 3$
$R_3 = [16, 31]$	$t_3 = 4$
$R_4 = [32, 63]$	$t_4 = 5$
$R_5 = [64, 127]$	$t_5 = 6$
$R_6 = [128, 255]$	$t_6 = 7$

Step 7. For each pixel p_i in the block and using the indices of pixels in N_s , replace the k -rightmost LSBs of the pixel p_i by the k -leftmost bits of the bitstream S' ($k \in \{3, 4\}$) to obtain stego pixel p_i' . In addition, transform these k bits-LSBs from p_i and k bits from the binary bitstream S' into their decimal values (call them LSB_i and s_i , respectively).

Step 8. Calculate the difference value a_i between the values of LSB_i and s_i as follows:

$$a_i = LSB_i - s_i. \quad (4)$$

Step 9. Use the OPAP and modify the value of p_i' with p_i''

$$p_i'' = \begin{cases} p_i' + 2^k, & \text{if } a_i > 2^{k-1} \text{ and } p_i' + 2^k \in [0, 255] \\ p_i' - 2^k, & \text{if } a_i < -2^{k-1} \text{ and } p_i' - 2^k \in [0, 255] \\ p_i', & \text{otherwise} \end{cases}. \quad (5)$$

Step 10. Move to the next block and return back to Step 4.

Step 11. For embedding secret data bits using Octa-PVD, use the obtained sub-ranges R_j from Step 4 to calculate the number of embedding secret bits (t_i) for each pixel pair in the current block B_i . The number of embedding secret data bits can be calculated using the following equation:

$$t_i = \log_2(u_i - l_i + 1). \quad (6)$$

Step 12. For each pixel pair, read t_i bits from the binary bitstream S' and convert them into its decimal value b_i .

Step 13. Calculate eight new difference values d_i' ($i=0, \dots, 7$) to replace the eight original difference values d_i ($i=0, \dots, 7$)

$$d_i' = \begin{cases} l_i + b_i, & \text{if } d_i \geq 0 \\ -(l_i + b_i), & \text{otherwise} \end{cases}. \quad (7)$$

Step 14. Compute z_i as follows:

$$z_i = (d_i' - d_i) / 2, \quad (i=0, \dots, 7). \quad (8)$$

Step 15. Calculate new pixel pair values (p_i', p_{i+1}') using

$$(p_i', p_{i+1}') = (p_i - \lceil z_i \rceil, p_{i+1} + \lfloor z_i \rfloor). \quad (9)$$

Step 16. Use the optimal reference point selection approach (ORPSA) that will be explained later to select the optimal reference point from the resulting eight pixel pairs and then use it to modify the other seven pixel pairs.

Step 17. If there is any falling-off-boundary cases happened, then use the shifting function and return back to Step 6 to re-embed this block using MLSB.

Step 18. Move to the next block and go to Step 4 until all secret data bits are embedded in the cover-image.

Notice that by using the shifting function, all blocks in the cover-image will be exploited to hide secret data bits, and thus giving higher embedding capacity. It is worth stating that the complexity time produced by our method is about $O(n^2)$ which is less than or equal to 59s. The embedding process is illustrated in Fig. 4.

Optimal Reference Point Selection Approach (ORPSA)

To achieve a minimum mean square error (MSE) and thus minimize stego-image distortion, the optimal reference point must be carefully obtained to adjust the other remaining gray values of the pixel pairs in each block. The authors of the TPVD algorithm [30] proposed optimal selection rules for obtaining the reference point according to m_i values where $m_i = d_i - d'_i$. However, their rules for selecting the optimal reference point depend only on three values of ' m_i ' since there are three directional pixel pairs. Unfortunately, these rules are not applicable in this algorithm since there exist five or eight directional pixel pairs. Therefore, in this algorithm, a new function is proposed for selecting the optimal reference point for each block, depending on the calculations of the PSNR. Consequently, the optimal reference point can be obtained dynamically to ensure that the best selection is acquired. The overall computational complexities, induced by this method and TPVD method, are approximately the same.

Shifting function

One of the deficiencies of PVD method is the falling-off-boundary problem. Therefore, a need to use a shifting function arises to modify one pixel value in the block B_i to re-embed data in this block using MLSB method. Using shifting function in the proposed algorithm will somehow increase the embedding capacity.

Moreover, this modification for one pixel value rarely affects the overall visual quality of the stego-image significantly. The following steps describe how the shifting function works considering the following variables:

- _ Mp is the maximum value in block B_i .
- _ Mip is the middle pixel in block B_i .
- _ BC_i is the selected branch condition.

Step 1. Find the maximum pixel (Mp) among pixels in the block B_i , excluding the middle pixel Mip .

Step 2. Let $Omp = Mp$.

Step 3. Increase Mp by 1 (i.e. $Mp = Mp + 1$). Then, compute $dif = Mp - Mip$.

Step 4. If only one difference value satisfies the selected BC_i (i.e., if the selected branch condition is BC_i , that means the condition will be "if the dif is equal or greater than 8"), then repeat Step 3 until this "if condition" is satisfied.

Step 5. If the Mp is not in the range of 0-255 (i.e. $Mp < 0$ or $255 < Mp$), then make $Mp = Omp$. Otherwise, go to Step 8.

Step 6. Increment Mp by 1 (i.e. $Mp = Mp - 1$). Then, compute $dif = Mp - Mip$.

Step 7. If dif does not satisfy the selected BC_i , then repeat Step 6 until the "if condition" is satisfied.

Step 8. Modify the original value of the max pixel in the block B_i with Mp .

Embedding phase example

Assuming the bitstream of secret data is given as described in Fig. 5(a) and the sample sub-block for nine neighboring pixels is given as found in Fig. 5(b) where the sub-block has the gray-values of (192, 190, 192, 192, 193, 189, 192, 189, and 187). Consequently, the following are the outcomes (in sequence):

- _ The difference values are generated as shown in Fig. 5(c).
- _ Assuming branch condition 1 is used, all the difference values, in this sub-block, are less than 8. Thus, this block will be embedded using Octa-PVD where the optimal range for all pairs in this block is $R_i = [0, 7]$ as shown in Fig. 5(d). Consequently, the number of secret data bits can be obtained as follows: $t_i = \log_2 |7 - 0 + 1| = 3 \text{ bits}$, which will be embedded in each pixel pair.

Figure 4

The flowchart of the embedding procedure in MDPVD-MLSB algorithm

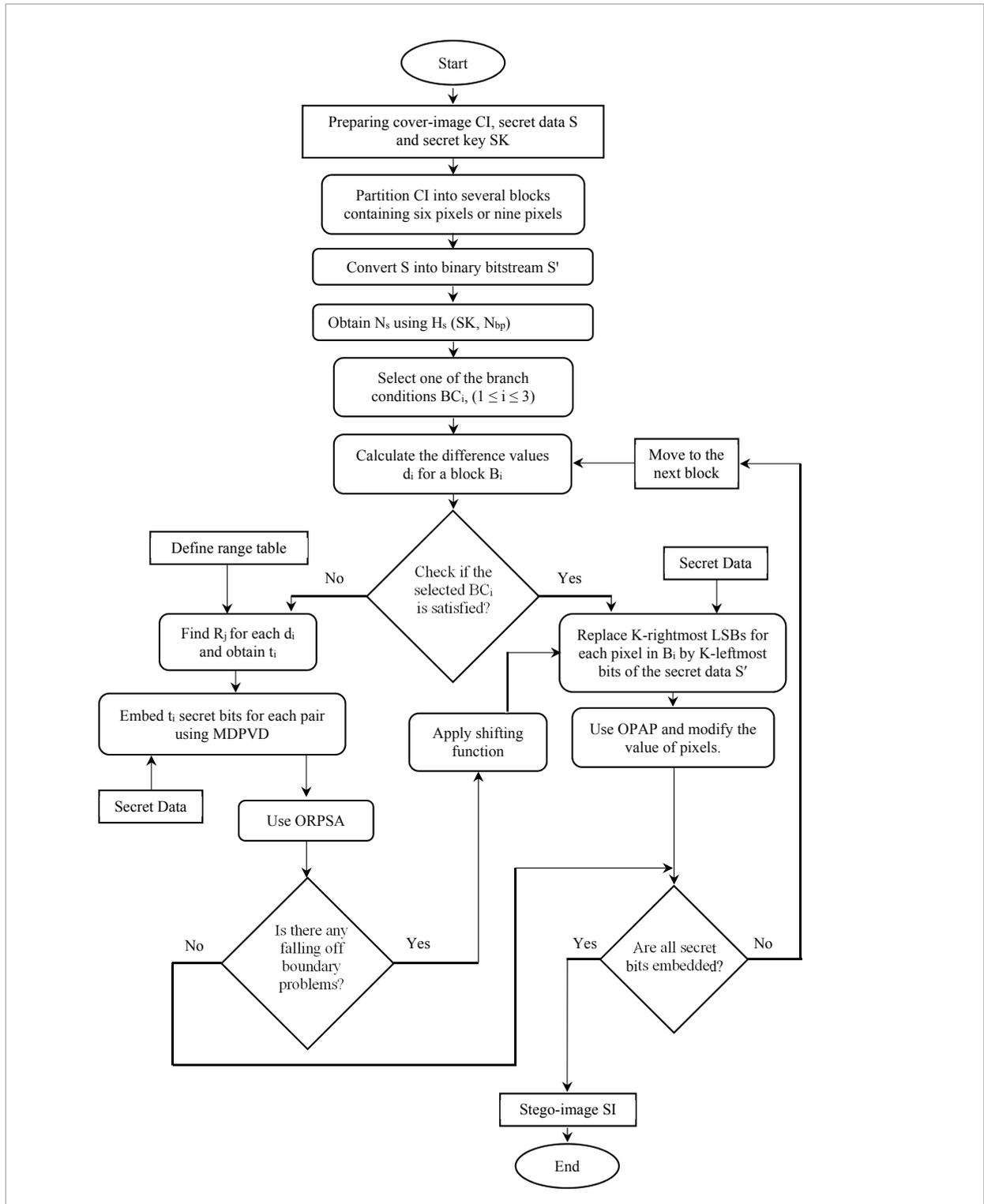
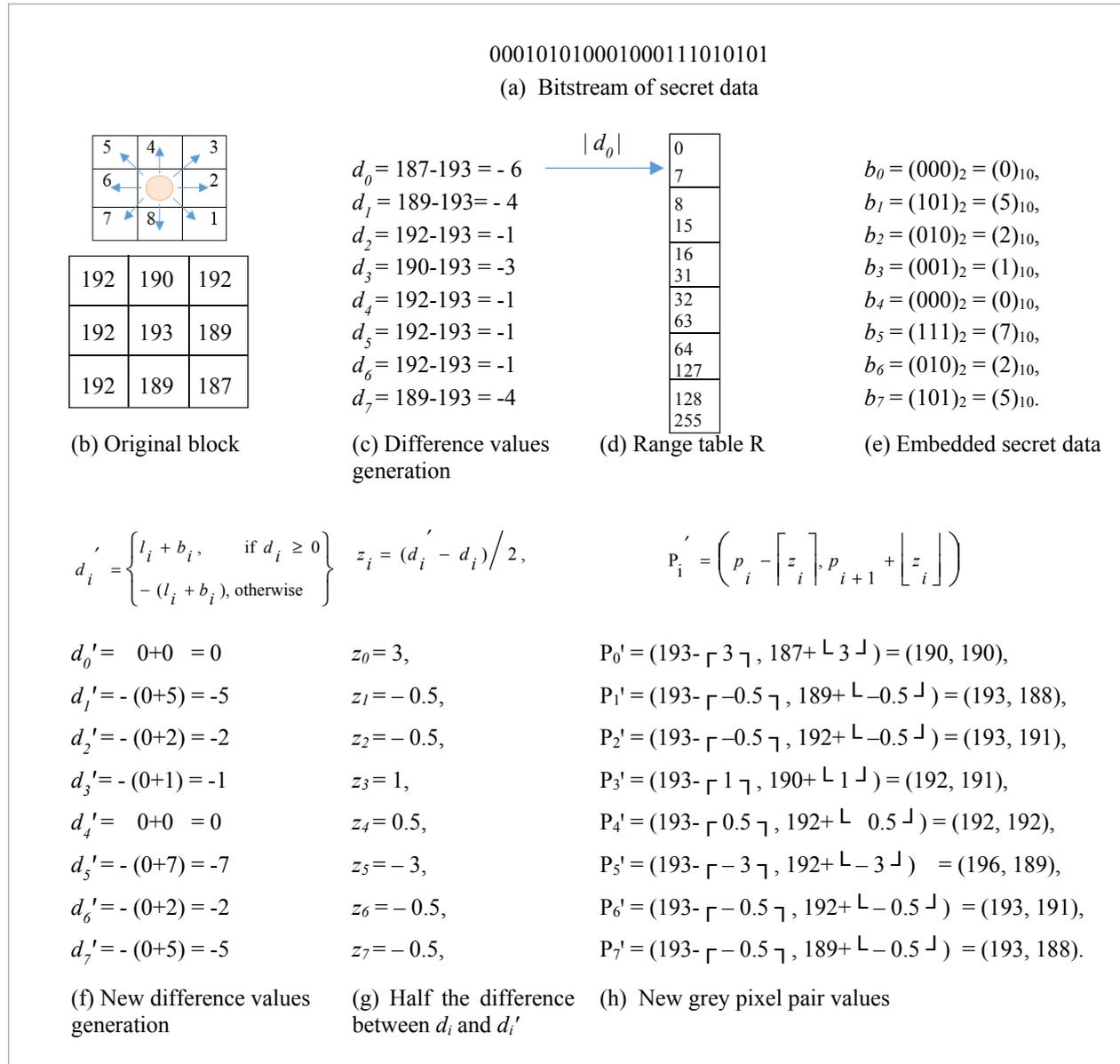


Figure 5 (a-h)

An example of the embedding process in our Octa-PVD-MLSb scheme when using the branch condition 1. Since all the difference values are less than 8, the Octa-PVD is used to embed secret data bits



- Using the bitstream of secret data bits in Fig. 5(a), the embedded bits b_i for each pixel pair will be as shown in Fig. 5(e).
- The new eight difference values d'_i can be generated as shown in Fig. 5(f).
- z_i values are computed as shown in Fig. 5(g).

- The new pixel pairs are calculated as shown in Fig.5 (h). Therefore, the new pixel pairs will be as follows: $\{(190, 190), (193, 188), (193, 191), (192, 191), (192, 192), (196, 189), (193, 191), (193, 188)\}$.
- Finally, the optimal pixel pair can be found by using the ORPSA. This procedure is demonstrated as follows:

- 1 When having the pixel pairs that are shown in Fig.5 (h), which are $\{P'_0 = (190,190), P'_1 = (193,188), P'_2 = (193,191), P'_3 = (192,191), P'_4 = (192,192), P'_5 = (196,189), P'_6 = (193,191), P'_7 = (193,188)\}$, then the first pixel pair denoted by P'_0 is the optimal reference point. Consequently, the offsetting process for the other seven pixel pairs will be as follows: $P'_1 = (193+190-193, 188+190-193) = (190,185)$, $P'_2 = (193+190-193, 191+190-193) = (190,188)$, $P'_3 = (192+190-192, 191+190-192) = (190,189)$, $P'_4 = (192+190-192, 192+190-192) = (190,190)$, $P'_5 = (196+190-196, 189+190-196) = (190,183)$, $P'_6 = (193+190-193, 191+190-193) = (190,188)$, $P'_7 = (193+190-193, 188+190-193) = (190,185)$. The resulting pixel pairs are shown in the modified sub-block 'Msb1' that is shown in Fig. 6 (a). After that, the PSNR of Msb1 is found.
- 2 Moving to the second pixel pair ' P'_1 ' and assuming it is the optimal reference point, the same procedure done in Step 1 is considered to offset the other remaining pixel pairs in order to obtain 'Msb₂' and calculate its PSNR value.
- 3 Repeating the same work performed in Steps 1 and 2 for the rest of pixel pairs.
- 4 Obtaining a total of eight modified sub-blocks (msb_1, \dots, msb_8) and eight values of the PSNR, as shown in Fig. 6 (a-h). As depicted in Fig. 6 (a-h), the optimal reference pair may be one of the following pixel pairs: $\{2, 3, \text{ or } 7\}$ since all of them achieve the highest PSNR (i.e., PSNR = 38.5884). For simplicity,

ty, the first one is proposed to be dynamically chosen. Thus, the final embedded block is the Msb_2 , as illustrated in Fig. 6 (b).

Extracting phase

To extract secret data bits correctly from each block B_i in a stego-image SI , the following steps are considered:

Input: A $W \times H$ stego-image SI , range table R and secret key SK .

Output: The secret data S .

Step 1. Partition SI into 3×3 non-overlapping sub-blocks.

Step 2. As in the embedding phase, obtain N_s according to $H_s(SK, N_{bp})$.

Step 3. Choose the same branch condition that is selected by the embedding algorithm $BC_i (1 \leq i \leq 3)$.

Step 4. Repeat Step 2 in the embedding phase to calculate the difference values $d'_i (0 \leq i \leq 7)$ for eight pixel pairs.

Step 5. If the branch condition is satisfied, then extract this block by using MLSB. Otherwise, extract this block by using Octa-PVD and apply Steps from 8 to 10.

Step 6. Extract the secret bits s_i from k-rightmost LSBs of each pixel p'_i in this sub-block. As in the embedding process, the secret bits are extracted according to the pixel indices in N_s set.

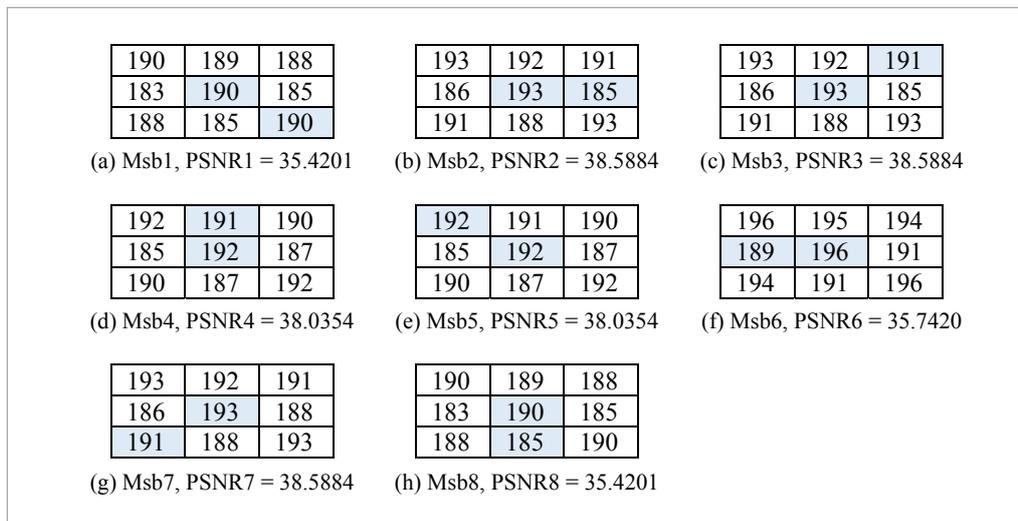


Figure 6 (a-h)
Obtaining the optimal reference point for example 1

Step 7. Move to the next block and then return back to Step 4.

Step 8. Find the value of lower bound l_i by obtaining the optimal range R_i .

Step 9. Extract the embedding secret data bits from each pixel pair in a block using:

$$s_i = |d'_i| - l_i. \quad (10)$$

Move to the next block and then go to Step 4 until the recovery process of all the secret bits is finished. Then, concatenate all the secret bits s_i in the same order as in the embedding process to obtain S' . Finally, convert S' to form the original secret data S .

The extraction process is illustrated in the flowchart shown in Fig. 7.

It is noteworthy to mention that all of the side information used in this algorithm such as the secret key and the branch condition are offline, except the amount of the payload which occupies only 32 bits of the embedding capacity.

Experimental results and discussion

In this section, the performance of the proposed algorithm is evaluated, discussed and justified.

Firstly, the simulation setup and its parameters are introduced. Secondly, a detailed description of the used metrics is presented. Lastly, the simulation results and comparisons are presented.

Simulation setup

Simulation parameters

The proposed algorithm has been implemented and simulated using MATLAB 8.2.0.701 (R2013b) on Windows 7 platform with an Intel Core i7-4600U CPU working at 2.1 GHz with a 4 MB cache and 4 GB RAM. The results for different grayscale images are obtained for various random secret data. The various simulation parameters are as given in Table 2.

Performance metrics

The performance of the proposed algorithms was evaluated in terms of imperceptibility and embedding capacity which are defined as follows:

Table 2

Simulation parameters

Cover image pixel size (N×N)	N = 512
Image type	Tiff, jpg, bmp, gif
Simulation tool	MATLAB 8.2.0.701
Secret data	Random strings using randseq ()

Imperceptibility: To measure the imperceptibility or the perceptual quality of a stego-image, the PSNR is used as well as the structural similarity index measure (SSIM) metrics. PSNR is the simplest and most widely used metric [48-51]. It can be found as follows [40, 47]:

$$PSNR = 10 \times \log_{10} \left(\frac{(255)^2}{MSE} \right), \quad (11)$$

where

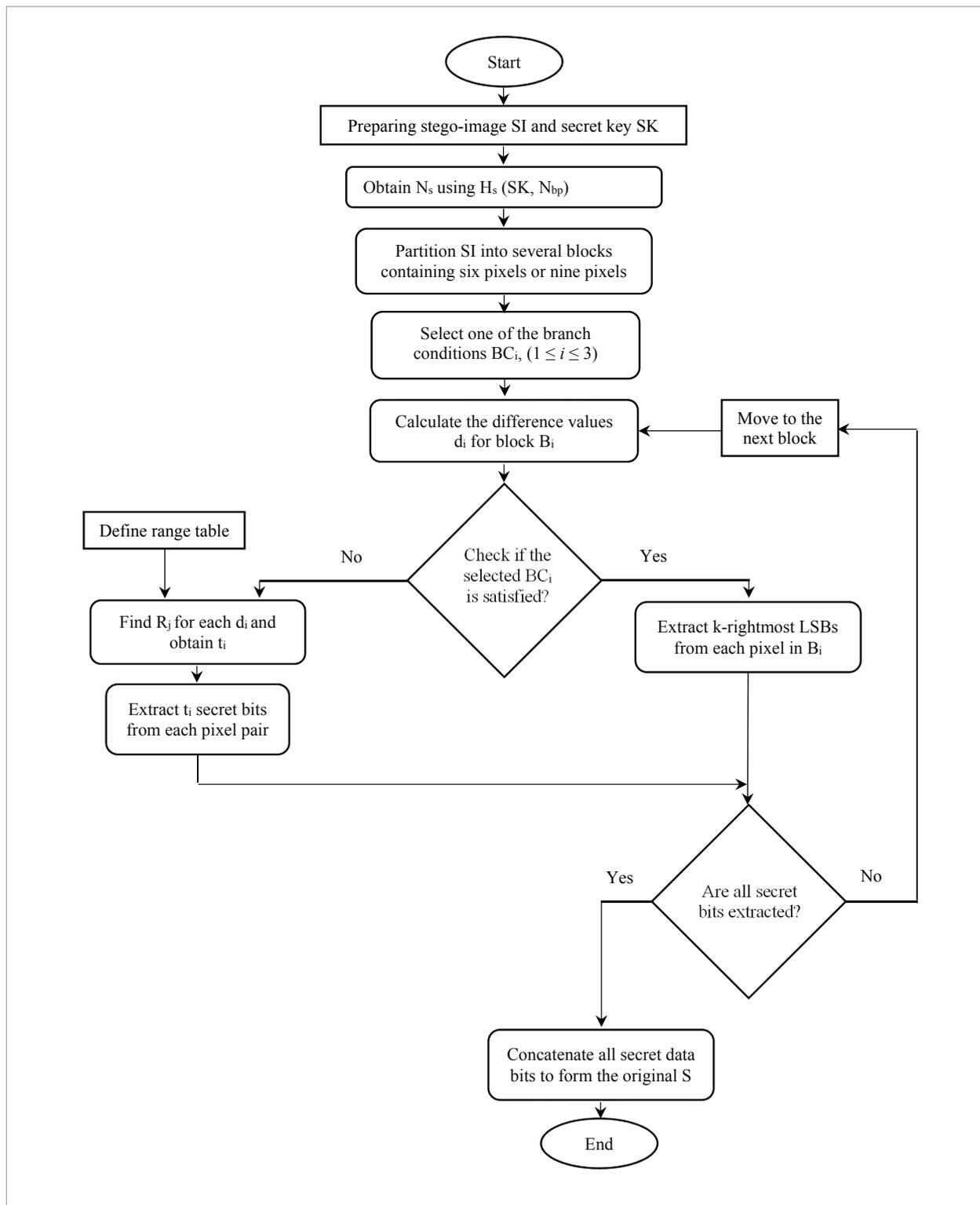
the value 255 refers to the maximum value of the pixel intensity for 8-bit grayscale images and MSE is the mean square error, which is used to compute the average square of the difference between the grey-scale cover image and its stego-image and is calculated using [52-56]:

$$MSE = \frac{1}{M \times N} \times \sum_{i=0}^{M-1} \sum_{j=0}^{N-1} (x_{ij} - x'_{ij})^2. \quad (12)$$

In this equation, " $M \times N$ " represents the size of the cover-image and stego-image. In addition, x_{ij} and x'_{ij} represent the pixel values at certain index (i, j) of the cover-image and stego-image, respectively. In general, to keep the grayscale cover-image and stego-image indiscernible to the human eye, a high value for the PSNR should be kept. Thus, a high PSNR means that the cover-image and stego-image are very similar to each other whereas a low PSNR means the opposite [57-58]. Besides PSNR, which is error-based metric, SSIM is very popular due to the fact that the human visual system is highly sensitive to structural information. In other words, any loss of structural data can give a good approximation for distortion of the per-

Figure 7

The flowchart of the extracting procedure in MDPVD-MLSB algorithm



ceived image. However, if the cover and stego images are defined as x and y respectively, then SSIM is computed using [59]:

$$SSIM(x, y) = \frac{(2\mu_x\mu_y + C_1)(2\sigma_{xy} + C_2)}{(\mu_x^2 + \mu_y^2 + C_1)(\sigma_x^2 + \sigma_y^2 + C_2)}, \quad (13)$$

where

μ_x and σ_x^2 refer to the mean and variance of x , respectively, μ_y and σ_y^2 refer to the mean and variance of y , respectively, σ_{xy} denotes for the covariance of x and y . $C_1 = (k_1 L)^2$ and $C_2 = (k_2 L)^2$ are two constants required to stabilize the division when the mean and variance get close to zero, where $k_1 = 0.01$, $k_2 = 0.03$, and $L = 2^{N_{bpp}} - 1$ (where N_{bpp} is the number of bits per pixel), representing the maximum possible value of the image pixel. It is good to mention that the results of SSIM fall within the range of $\{0, 1\}$ in which '1' indicates that the two images are completely identical, while '0' indicates that the two images are entirely different. However, for each image, there are several SSIM indices where each one is calculated within 11×11 local window using a certain circular-symmetric Gaussian weighting value (between 0 and 1) and the final SSIM image index is the average of these indexes.

– **Embedding Capacity (EC):** The embedding capacity is defined as the number of secret data bits that can be hidden in a cover-image. It is also known as embedding rate (ER), which represents the number of secret data bits that can be hidden per pixel, that is basically measured as the ratio of the total number of embedded secret bits to the total number of pixels in a cover-image. Actually, whenever this ratio exceeds one, it indicates that the steganographic embedding scheme has high embedding capacity [3, 19]. It can be calculated as follows:

$$EC = \text{Total number of embedded secret bits}. \quad (14)$$

$$ER = \frac{\text{Total number of embedded secret bits}}{\text{Total number of pixels in CI}} \quad (\text{bits/pixel}). \quad (15)$$

$$ER = \frac{EC}{M \times N} \quad (bpp). \quad (16)$$

Experimental results

In our experiments, eight 8-bit 512x512 benchmark images, which are obtained from *USC-SIPI Image database* and *UWATERLOO-LINKS Image Repository* [60, 61], are used.

These grayscale images namely Lena, Peppers, Boat, Jet, Splash, Airplane, Baboon and Tiffany are shown in Fig. 8 (a-h). Data to be embedded is randomly generated. The range table used in the experiments is shown in Fig. 3. The maximum embedding capacity in this method can be calculated as follows:

1 Find the embedding capacity for MDPVD blocks:

$$EC_1 = ASDB \times Nb_1, \quad (17)$$

where

$ASDB$ is the average of the secret data bits per block and Nb_1 is the number of MDPVD blocks (either using Quinary-PVD or Octa-PVD). Moreover, $ASDB$ can be calculated as follows:

$$ASDB = \frac{\sum(\sum SDB/Np)/(L/2)}{Np}, \quad (18)$$

where

SDB is the sum of all secret data bits in a block, Np is the number of pixels for each block and L is the length of all pixels that is used in embedding.

2 Find the capacity for MLSB blocks as:

$$EC_2 = (k \times Np) \times Nb_2, \quad (19)$$

where

k may be 3 bits or 4 bits and Nb_2 is the number of MLSB blocks.

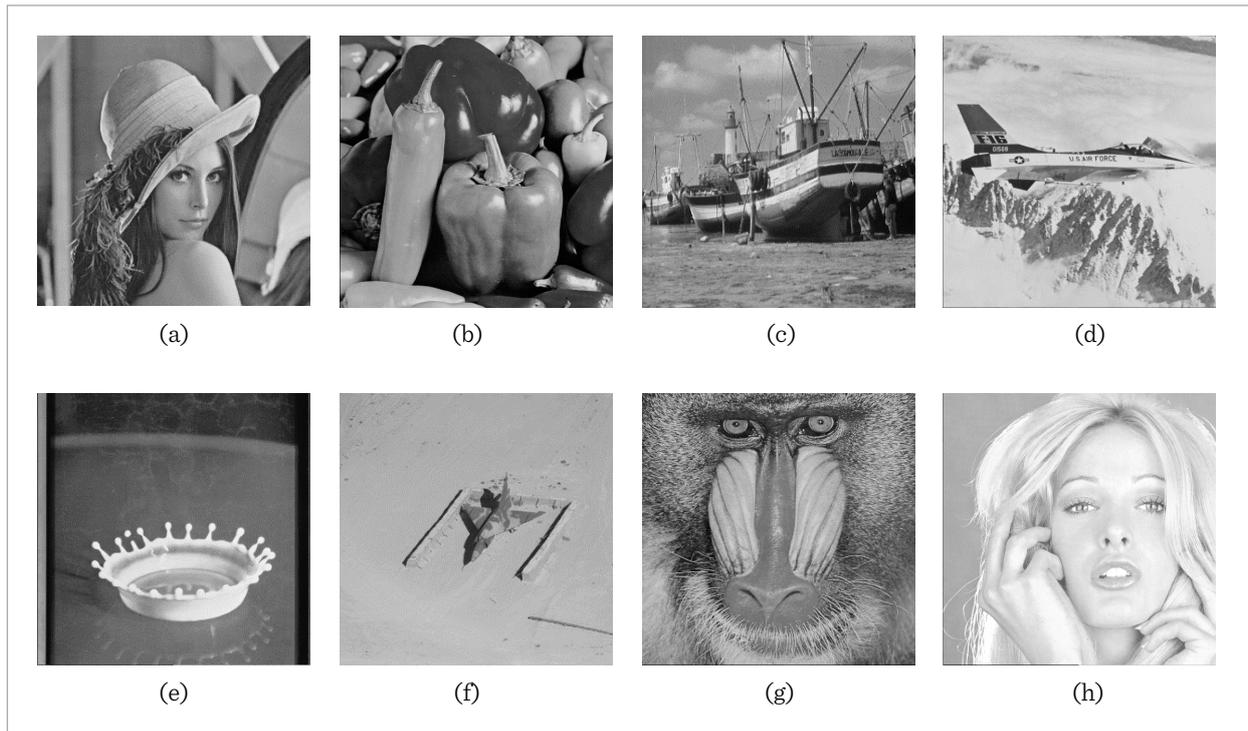
From (16) and (18), we can obtain the maximum embedding capacity (in bits) as follows:

$$EC = \lfloor EC_1 + EC_2 \rfloor. \quad (20)$$

Figs. 9 and 10 show the stego images produced by the Quinary-PVD-MLSB scheme for Lena and Peppers cover images when using $k = 3$ bits, to be embedded utilizing the MLSB, and using all branch conditions introduced.

Figure 8 (a-h)

Test images used in the experiments (a) Lena, (b) Peppers, (c) Boat, (d) Jet, (e) Splash, (f) Airplane, (g) Baboon, (h) Tiffany

**Figure 9**

Results of our proposed Quinary-PVD-MLSB scheme on Stego-Lena with $k = 3$ and all three branch conditions

Lena Cover Image	Stego-Lena with Branch Condition 1	Stego-Lena with Branch Condition 2	Stego-Lena with Branch Condition 3
			
EC (Bits)	715,976	679,848	663,024
SSIM	0.9950	0.9957	0.9955
PSNR	40.1495	40.2623	39.7181

Figure 10

Results of our proposed Quinary-PVD-MLSB scheme on Stego-Peppers with $k = 3$ and all three branch conditions

Peppers Cover Image	Stego-Peppers with Branch Condition 1	Stego-Peppers with Branch Condition 2	Stego-Peppers with Branch Condition 3
			
EC (Bits)	727,208	679,976	783,328
SSIM	0.9946	0.9955	0.9951
PSNR	39.8849	39.8061	38.6543

As the figures show, there are no visual artifacts present between the cover images and their corresponding stego images. Similarly, in case of using Octa-PVD-MLSB scheme, the results are shown in Figs. 11 and 12.

When comparing the resulting stego images along with their corresponding cover images, it is obvious that all the stego images have an amazing visual quality for any human eye.

Figure 11

Results of our proposed Octa-PVD-MLSB scheme on Stego-Lena with $k = 3$ and all three branch conditions

Lena Cover Image	Stego-Lena with Branch Condition 1	Stego-Lena with Branch Condition 2	Stego-Lena with Branch Condition 3
			
EC (Bits)	741,464	715,056	702,304
SSIM	0.9942	0.9945	0.9943
PSNR	40.0316	39.9513	39.3894

Figure 12

Results of our proposed Octa-PVD-MLSb scheme on Stego-Peppers with $k = 3$ and all three branch conditions

Peppers Cover Image	Stego-Peppers with Branch Condition 1	Stego-Peppers with Branch Condition 2	Stego-Peppers with Branch Condition 3
			
EC (Bits)	750,952	716,264	780,264
SSIM	0.9942	0.9945	0.9940
PSNR	39.8919	39.5898	38.5076

Discussion

To improve the stego-image visual quality in the Octa-PVD method proposed in [46], the proposed approach is used to select the optimal reference point. In addition, the pair directions are changed by taking the directions for each pixel block as proposed in MD-PV-MLSb algorithm, see Fig. 1 (b). However, Table 3 shows the results of our improved Octa-PVD and

original Octa-PVD algorithms. The average improvement ratio (AIR) in regards to the PSNR is about 36% bearing in mind that the improvement ratio can rise up to 52%.

To have a comprehensive evaluation, our MDPVD-MLSb algorithm is further compared with the original PVD algorithm [40] and five different PVD-based steganographic algorithms proposed in [30,

Table 3

The embedding capacity and PSNR for our improved Octa-PVD method and the original Octa-PVD method proposed in [46] for eight cover images

Cover-image	Octa-PVD [46]		Our improved Octa-PVD	
	EC (Bits)	PSNR (dB)	EC (Bits)	PSNR (dB)
Lena	706,872	26.4350	708,496	34.6617
Peppers	705,480	24.3073	709,272	33.1697
Boat	721,008	23.6600	721,688	32.5751
Jet	704,360	24.6409	713,336	34.1273
Splash	700,688	25.5480	701,408	36.0862
Airplane	698,056	28.9938	699,760	37.4130
Baboon	777,944	19.1521	787,664	29.1745
Tiffany	703,800	27.3457	705,264	34.0199
Average	714,776	25.0104	718,360	33.9034

42-43, 45-46]. Table 4 (a-b) shows the results of our algorithm and these six algorithms with respect to the maximum EC and PSNR. As shown in Table 4 (a), our proposed algorithm, the Octa-PVD-MLSB, provides higher values of EC than those obtained by PVD, TPVD, and Octa-PVD algorithms. The EC AIRs of our algorithm and these three methods are about 80%, 21% and 4%, respectively. Moreover, our algorithm provides higher PSNR than that achieved by TPVD and Octa-PVD. The PSNR AIRs are about 5% and 62%, respectively (keeping in mind that the improvement ratio can reach up to 15% and 108%, respectively).

When comparing our algorithm with PVD in regards to PSNR, the average degradation ratio (ADR) of our algorithm is only about 4%. Table 4 (b) illustrates that our proposed algorithm outperforms those proposed in [42, 43, and 45]. Specifically, the EC AIRs are about 16%, 19% and 84%, respectively (the improvement ratio can reach up to 22%, 23% and 87%, respectively). In addition, our proposed algorithm achieves higher values of PSNR than those in methods [42] and [43]. In other words, the PSNR AIRs are about 9% and 36%, respectively (the improvement ratio can rise up to 16% and 54%, respectively).

Table 4 (a-b)

The performance efficiency of our MDPVD-MLSB method against PVD [40], TPVD [30], Octa-PVD [46], and methods in [42, 43, and 45]

Cover image name	PVD [40]		TPVD [30]		Octa-PVD [46]		Quinary-PVD-MLSB 3 LSBs, Branch condition 1		Octa-PVD-MLSB 3 LSBs, Branch condition 1	
	EC (Bytes)	PSNR (dB)	EC (Bytes)	PSNR (dB)	EC (Bytes)	PSNR (dB)	EC (Bytes)	PSNR (dB)	EC (Bytes)	PSNR (dB)
<i>Panel a</i>										
Lena	50,960	42.12	75,836	38.84	88,359	26.44	89,497	40.15	92,683	40.03
Peppers	50,685	41.89	75,579	38.42	88,185	24.31	90,901	39.88	93,869	39.89
Boat	52,635	40.01	77,982	37.12	90,126	23.66	93,383	39.89	95,356	39.89
Jet	51,025	41.79	76,123	38.43	88,045	24.64	87,208	40.49	91,037	40.26
Splash	49,932	42.18	74,700	39.29	87,586	25.55	86,057	40.38	90,584	40.15
Airplane	49,738	43.48	74,372	40.38	87,257	28.99	84,017	40.74	88,782	40.46
Baboon	56,291	38.10	82,407	34.62	97,243	19.15	96,645	39.89	97,987	39.94
Tiffany	50,919	42.16	75,650	38.92	87,975	27.35	88,511	39.96	92,033	39.93
Average	51,523	41.47	76,581	38.25	89,347	25.01	89,527	40.17	92,791	40.07
Cover image name	Method in [42]		Method in [43]		Method in [45]		Quinary-PVD-MLSB 3 LSBs, Branch condition 1		Octa-PVD-MLSB 3 LSBs, Branch condition 1	
	EC (Bytes)	PSNR (dB)	EC (Bytes)	PSNR (dB)	EC (Bytes)	PSNR (dB)	EC (Bytes)	PSNR (dB)	EC (Bytes)	PSNR (dB)
<i>Panel b</i>										
Lena	77,265	41.25	76,849	31.94	50,311	42.46	89,497	40.15	92,683	40.03
Peppers	76,938	37.62	76,424	30.42	50,136	42.68	90,901	39.88	93,869	39.89
Boat	80,839	36.33	--	--	51,097	41.61	93,383	39.89	95,356	39.89
Airplane	78,039	38.89	76,853	30.66	--	--	84,017	40.74	88,782	40.46
Baboon	92,382	34.33	85,777	25.96	55,434	38.88	96,645	39.89	97,987	39.94
Average	81,093	37.68	78,976	29.75	51,745	41.41	90,889	40.11	93,735	40.04

Conclusions and future work

Although the TPVD and Octa PVD methods, which are based on multi-directional PVD, provide high capability of embedding data, these methods, especially Octa PVD method, may suffer from unacceptable distortion in image quality due to the offsetting process of the pixel pairs employed. Therefore, this article proposes an efficient and robust dynamic algorithm through taking the advantage of both multi-directional PVD and MLSB, namely, MDPVD-MLSB algorithm, which mainly intends to improve the capacity, maintain good visual quality, and provide high protection for embedded data. The performance of

MDPVD-MLSB algorithm is comprehensively evaluated and compared with other six related algorithms where the experimental results demonstrate the strength and effectiveness of our proposed algorithm.

Many directions can be given for further enhancement to the proposed algorithm. The algorithm's framework can be extended to the RGB color images for enhancing the capability of embedding. Moreover, it can be a good addition to develop an approach that takes into account the hybrid domain. Additionally, there are future plans to develop PVD-based schemes for another media such as audios and videos.

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Summary / Santrauka

Steganographic techniques can be utilized to conceal data within digital images with small or invisible changes in the perceived appearance of the image. Generally, five main objectives are used to assess the performance of steganographic algorithms which include embedding capacity, imperceptibility, security, robustness and complexity. However, steganographic algorithms hardly take all of these factors into account. In this paper, a novel steganographic algorithm for digital images is proposed based on the pixel-value differencing (PVD) and modified least-significant bit (LSB) substitution (MDPVD-MLSB) techniques to address most of aforementioned objectives. Although there are many techniques for concealing data within pixels, the restricting factor is always the amount of bits adjusted in every pixel. Therefore, the main contributions of this paper aim to achieve a balance between the amount of embedded data, the level of acceptable distortion, as well as providing high level of security. The performance of this algorithm has been extensively evaluated in terms of embedding capacity, peak signal-to-noise ratio (PSNR) and structural similarity index measure (SSIM). Simulation results and comparisons with six relevant algorithms are presented to demonstrate the effectiveness of this proposed algorithm.

Steganografinės technikos gali būti panaudotos paslėpti duomenims skaitmeniniuose atvaizduose, padarant mažų arba nepastebimų pokyčių atvaizdo vaizdinėje kokybėje. Įprastai steganografiniai algoritmai yra vertinami pagal penkis siekinius, kurie apima įterptinę talpą, nepastebimumą, saugumą, patikimumą ir sudėtingumą. Deja, steganografiniai algoritmai retai atsižvelgia į šiuos faktorius. Atsižvelgiant į minėtus siekinius, šis straipsnis siūlo naujovišką steganografinį algoritmą, paremtą pikselių vertės diferencijavimu (angl. Pixel-value differencing (PVD)) ir modifikuotomis mažiausiai reikšmingo bito (angl. Least significant bit (LSB)) pakeitimo (MDPVD-MLSB) technikomis. Nors yra daug technikų duomenų paslėpimui pikseliuose, ribojantysis faktorius visuomet yra kiekviename pikselyje pakeistų pikselių kiekis. Šiuo straipsniu siekiama užtikrinti balansą tarp įtvirtintųjų duomenų ir priimtino atvaizdo iškraipymo, kartu garantuojant aukštą saugumo lygį. Algoritmo efektyvumas išsamiai įvertintas atsižvelgiant į įtvirtintuosius gebėjimus, signalo maksimumo taškų santykį su triukšmu (angl. Peak Signal-to-Noise Ratio (PSNR)) ir struktūrinį panašumo indekso parametą (angl. Structural Similarity Index Measure (SSIM)). Kad būtų parodytas siūlomo algoritmo efektyvumas, pateikiami simuliacijos rezultatai ir palyginimai su šešiais aktualiais algoritmais.