



Extended color visual cryptography for black and white secret image



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ABSTRACT

Recently, De Prisco and De Santis introduced a (k, n) colored-black-and-white visual cryptography scheme (CBW-VCS), which adopts colored pixels in shadow images to share a black and white secret image. Only k or more participants can reconstruct the secret image by combining their shadow images. It seems that no one study (k, n) -CBW-VCS with the extended ability revealing meaningful shadow images (denoted as CBW-extended VCS (CBW-EVCS)). In this paper, we extend the conventional EVCS (i.e., BW-EVCS) to the CBW-EVCS. This paper has two main contributions: (i) we construct (k, n) -CWB-EVCSs, (ii) all constructions are proven to satisfy security, contrast, and cover image conditions.

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1. Introduction

The cryptographic technique for the visual sharing of secret images, known as a visual cryptographic scheme (VCS) or a Visual Secret Sharing scheme (VSSS) was first proposed by Naor and Shamir [16]. VCS is a method to decompose a secret image into shadow images (called shadows) and distribute them to a number of participants. Only legitimate subsets of shadows can be used to reconstruct the original image. VCS is usually implemented as a threshold k -out-of- n scheme. A (k, n) -VCS, where $k \leq n$, divides a secret image into n shadows by expanding a pixel in the secret image into m (the pixel expansion) pixels in shadows. In a (k, n) -VCS, we can reconstruct the secret image by simply stacking together the shadows they own; but $(k - 1)$ or fewer shadows cannot recover the secret image. VCS has a novel “stacking-to-see” property by which participants may photocopy their shadows onto transparencies and stack them to visually decode the secret. Its decoding requires neither knowledge of cryptography nor a computer.

Naor and Shamir's (k, n) -VCS shares a black-and-white secret image into black-and-white shadows. Because sizes of the pixel in secret image and the pixel in shadow are equal, the shadow size is expanded m times. Therefore, most studies attempted to reduce pixel expansion and enhance contrast [11,13,24,3,22], and some of them even had no pixel expansion. Various VCSs with specific functions, such as sharing multiple secrets, cheating prevention, achieving the ideal contrast, sharing gray image, retaining invariant aspect ratio, providing region incrementing property, providing progressive recovery, and adopting XOR operation were subsequently proposed [18,15,4,20,25,28,29,10,30].

Ito et al. [11] first proposed a VCS with no pixel expansion (i.e., $m = 1$) via randomly choosing a column from the black and white matrices of VCS. Yang [25] formally defined the so-called probabilistic VCS (PVCS) and provided various construction methods. The conventional VCS with the fixed $m (>1)$ is referred to as the deterministic VCS (DVCS). The

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major advantage of PVCS over DVCS is its non-expandable shadow size, while DVCS has the clear reconstructed image. Afterwards, a generalized PVCS with the pixel expansion $1 \leq m_p \leq m$ was further extended by Cimato et al. [3]. For $m_p = 1$, there is no pixel expansion and Cimato et al.'s VCS is the PVCS. On the other hand, Cimato et al.'s VCS becomes DVCS for $m_p = m$.

A very similar secret imaging technology like PVCS, the random grid (RG), was introduced by Kafri and Keren [12]. Kafri and Keren's RG shares a binary secret image into two noise-like random grids. When superimposing two random grids, only two stacked white pixels will let the light through it, while other stacked results yielding a black pixel stops the light. Finally, RG also has both the stacking-to-see property and the un-expandable shadows. In fact, RG and VCS have no difference other than the terminology. They are only different ways of looking at the same problem. Recently, the authors in [8,31] have shown that these models (PVCS, DVCS, and RG) are related to each other.

VCS can also be applied on colored secret images. A trivial solution is to convert the colored image into a binary image by the halftoning technique and then processes it by black-and-white VCS [9]. Various colored VCS (CVCS) were chronologically proposed [19,23,2,17,5,27,22,6]. Recently, De Prisco and De Santis [7] introduced a new notion of colored-black-and-white VCS (known as CBW-VCS). Same as the CVCS, the CBW-VCS has colored pixels in shadows but it shares the black-and-white secret image not the colored image. When compared with the black-and-white VCS, the CBW-VCS can reduce the pixel expansion and enhance the contrast due to using colored pixels in shadows.

In implementing VCS it would be useful to have a meaningful image on shadow. Any user will recognize the cover image on his shadow, so that the involved participants can easily identify and manage their shadows. Also, the meaningful shadows can conceal the existence of the secret image [1]. However, De Prisco and De Santis's CWB-VCS has noise-like shadows. In this paper, we study the CWB-VCS with the extended capability revealing a meaningful image on shadow (CWB-EVCS for short). There were already some (k, n) -EVCSs [1,26,21,14,32]. However, it seems that no one study CBW-EVCS. In this paper, via the idea of Ateniese et al.'s EVCS in [1], we give the conditions (*security, contrast, and cover image conditions*) of (k, n) -CWB-EVCS and extend the EVCS from BW model to CBW model.

The rest of this paper is organized as follows. In Section 2, (k, n) -VCS and (k, n) -CBW-VCS are briefly reviewed. In Section 3, we give the conditions of (k, n) -CBW-EVCS and show different constructions. Additionally, we prove that our (k, n) -CBW-EVCSs satisfy these conditions. Experiments and comparisons are included in Section 4. Examples and experimental results demonstrate that our (k, n) -CBW-EVCSs at least have the same pixel expansion and contrast when compared with the (k, n) -VCS. Meanwhile, our scheme provides the extended capability that shadows reveal the cover image. Finally, the conclusion is provided in Section 5.

2. Preliminary

2.1. (k, n) -VCS

A conventional (k, n) -VCS (a black-and-white VCS) is defined as the black and white $(n \times m)$ Boolean base matrices B_{\bullet} and B_{\circ} , on which we can determine the color. The collection C_{\bullet} (respectively, C_{\circ}) is obtained by permuting the columns of the corresponding matrix B_{\bullet} (respectively, B_{\circ}) in all possible ways. When sharing a black (respectively, white) secret pixel, the dealer randomly chooses one row of the matrix in C_{\bullet} (respectively, C_{\circ}) to a relative shadow. In (k, n) -VCS, every pixel is subdivided into m black-and-white pixels in each of n shadows. A (k, n) -VCS uses h black pixels and $(m - h)$ white pixels (denoted as $(h \bullet (m - h)\circ)$), and $(l \bullet (m - l)\circ)$, where $0 \leq l < h \leq m$, to represent black and white secret pixels, respectively. The values of h and l are the blackness of black color and white color. A VCS is the perfect black VCS (PB-VCS) for $h = m$. Let X be a set of involved participants. We denote $w_{\bullet}(v)$ as the number of elements of v equal to \bullet . Suppose that M is an $n \times m$ matrix. Let $(M|X)$ be a $|X| \times m$ matrix selecting the rows of the corresponding participants in X from M . Then, $\text{add}(M|X)$ denotes the OR-ed vector of all rows in $(M|X)$ and $D(M|X)$ denotes a set including all distribution matrices obtained by permuting all the columns in $(M|X)$. The formal definition of a VCS is given as follows [16,19].

Definition 1. A (k, n) -VCS is given by $(n \times m)$ black and white base matrices B_{\bullet} and B_{\circ} satisfying the following two conditions.

- (V-1) (Contrast condition): Given any qualified set X , where $|X| = k$, we have that $w_{\bullet}(\text{add}(B_{\bullet} | X)) \geq h$ (respectively, $w_{\bullet}(\text{add}(B_{\circ} | X)) \leq l$), where $0 \leq l < h \leq m$.
- (V-2) (Security condition): Given any forbidden set X , where $|X| < k$, we have $D_{\bullet} = D(B_{\bullet} | X)$ and $D_{\circ} = D(B_{\circ} | X)$. Then, the collections D_{\bullet} and D_{\circ} are indistinguishable in the sense that they contain the same matrices with the same frequencies. For simplicity, we say that both collections D_{\bullet} and D_{\circ} are equivalent.

The first condition (V-1) is the contrast condition and the second condition (V-2) is the security condition. In [16], the contrast is defined as the difference in blackness between a black pixel and a white pixel in the reconstructed image, i.e., $\alpha = \frac{(h-l)}{m}$. Obviously, our aim in designing a VCS is to achieve the contrast that is as large as possible (or to minimize the pixel expansion as small as possible) for reconstructing a clearer image.

Table 1.1
Full intensity colors.

Color	(R, G, B)
R	(○●●)
G	(●○●)
B	(●●○)
C	(●○○)
M	(○●○)
Y	(○○●)
●	(●●●)
○	(○○○)

Table 1.2
Stacked results of full intensity colors.

+	R	G	B	C	M	Y	●	○
R	●	●	●	●	R	R	●	R
G	●	G	●	G	●	G	●	G
B	●	●	B	B	B	●	●	B
C	●	G	B	C	B	G	●	C
M	R	●	B	B	M	R	●	M
Y	R	G	●	G	R	Y	●	Y
●	●	●	●	●	●	●	●	●
○	R	G	B	C	M	Y	●	○

2.2. De Prisco and De Santis's (k, n) -CBW-VCS

In a color mixing formula, there are three primary color lights with equal intensities – red (R), green (G) and blue (B). Two or more colors come together can form a different color. For example, R mixed with G (respectively, B) yields yellow (Y) (respectively, magenta (M)), and B mixed with G produces cyan (C). Additionally, a combination of all R, G, and B creates white (○), whereas black (●) color is shown when there is an absence of all three primary colors. These eight colors are full intensity colors, as shown in Table 1.1, and their stacked results are given in Table 1.2.

In [7], De Prisco and De Santis showed that using only full intensity colors can construct the CBW-VCS with the optimal contrast. Without loss of generality, they only use full intensity colors to construct $(2, 2)$ -CBW-VCS, $(2, 3)$ -CBW-VCS, $(2, n)$ -CBW-VCS, $(2, n)$ -PB-CBW-VCS, and (k, n) -PB-CBW-VCS. In these CBW-VCSs, the black pixel in the reconstructed image is used as the (●) color and any non-black pixel ($R, G, B, C, M, Y, ○$) are regarded as white pixel like the (○) in the conventional VCS. Here, we denote these non-black pixels as (*) in CBW-VCS. A (k, n) -CBW-VCS uses $(h \bullet (m - h)*))$ and $(l \bullet (m - l)*))$, as black and white secret pixels, respectively. Conditions (V-1) and (V-2) in VCS are also satisfied for the CBW-VCS. We use the following simple example, a $(2, 2)$ -CWB-VCS, to show the design concept of De Prisco and De Santis's (k, n) -CBW-VCS.

Example 1. Construct De Prisco and De Santis's $(2, 2)$ -CBW-EVCS with background colors $\{RGB\}$ and $\{RGBYMC\}$, respectively.

The collections C_{\bullet} and C_{\circ} of $(2, 2)$ -CBW-EVCS with $\{RGB\}$ and $(2, 2)$ -CBW-EVCS with $\{RGBYMC\}$ are shown in Eq. (1.1) and Eq. (1.2), respectively.

$$C_{\bullet} = \left\{ \begin{bmatrix} R \\ G \end{bmatrix} \begin{bmatrix} G \\ R \end{bmatrix} \begin{bmatrix} R \\ B \end{bmatrix} \begin{bmatrix} B \\ R \end{bmatrix} \begin{bmatrix} G \\ B \end{bmatrix} \begin{bmatrix} B \\ G \end{bmatrix} \right\}, \quad C_{\circ} = \left\{ \begin{bmatrix} R \\ R \end{bmatrix} \begin{bmatrix} G \\ G \end{bmatrix} \begin{bmatrix} B \\ B \end{bmatrix} \right\}. \quad (1.1)$$

$$C_{\bullet} = \left\{ \begin{bmatrix} R \\ C \end{bmatrix} \begin{bmatrix} C \\ R \end{bmatrix} \begin{bmatrix} G \\ M \end{bmatrix} \begin{bmatrix} M \\ G \end{bmatrix} \begin{bmatrix} B \\ Y \end{bmatrix} \begin{bmatrix} Y \\ B \end{bmatrix} \right\}, \quad C_{\circ} = \left\{ \begin{bmatrix} R \\ R \end{bmatrix} \begin{bmatrix} G \\ G \end{bmatrix} \begin{bmatrix} B \\ B \end{bmatrix} \begin{bmatrix} C \\ C \end{bmatrix} \begin{bmatrix} M \\ M \end{bmatrix} \begin{bmatrix} Y \\ Y \end{bmatrix} \right\}. \quad (1.2)$$

In Eq. (1.1), it is observed that $w_{\bullet}(M | X_2) = 1$ for $M \in C_{\bullet}$, $w_{\bullet}(M | X_2) = 0$ for $M \in C_{\circ}$, and $D_{\bullet} = D_{\circ} = \{R, G, B\}$ satisfy

contrast condition (V-1) and security condition (V-2). Here we use $(x \bullet y^*)$ to represent $(\overbrace{\bullet \cdots \bullet}^x * \overbrace{\cdots \bullet}^y)$ and its permutations, where * denotes a color in $\{R, G, B\}$. In a reconstructed image, the black color is $(1\bullet)$ and the white color is $(1*)$. Because every pixel in two shadows S_1 and S_2 is $(1*)$, shadows are noise-like (see Fig. 1(a)). As shown in Fig. 1(b), we can visually decode the secret image, a printed-text **VCS**. It can be verified that Eq. (1.2) also satisfies conditions (V-1) and (V-2). The $(2, 2)$ -CBW-EVCS with $\{RGBYMC\}$ also has $(1\bullet)$ and $(1*)$ as black color and white color in reconstructed image, where $* \in \{R, G, B, Y, M, C\}$. Fig. 2 shows two shadows and the reconstructed image. For this case, the background color is a mixture of the colored pixels in $\{R, G, B, C, M, Y\}$. The average light transmission of background λ can be enhanced from

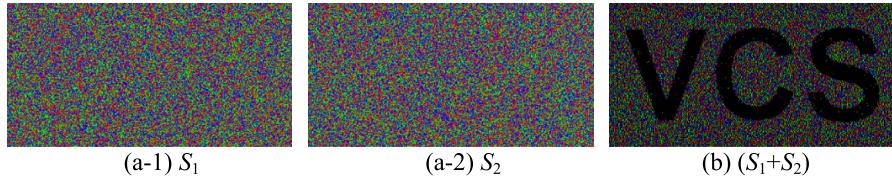


Fig. 1. (2, 2)-CBW-VCS with {RGB}: (a) shadows, (b) the reconstructed image.

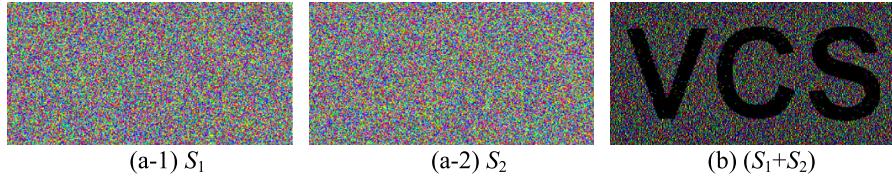


Fig. 2. (2, 2)-CBW-VCS with {RGBCMY}: (a) shadows, (b) the reconstructed image.

Table 2

Description and example of notations.

Notation	Description/Example
$S^{(1)}$	A set of all the strings of length m' over the alphabet Σ . For example, in Construction 1 , $m' = 1$ and $\Sigma = \{R, G, B\}$ and thus $S^{(1)} = \{R, G, B\}$ has 3 elements. On the other hand, in Construction 4 , we have $S^{(1)} = \{RR, RG, RB, GR, GG, GB, BR, BG, BB\}$ with 9 elements for $m' = 2$ and $\Sigma = \{R, G, B\}$.
$S^{(i)}$	$S^{(i)}$, $2 \leq i \leq n$, is a set by removing the chosen element from the set $S^{(i-1)}$, above which is the set $S^{(i)}$ located in the same column of a matrix. For example, in Construction 1 , if R is chosen from $S^{(1)} = \{R, G, B\}$, then $S^{(2)}$ is $\{G, B\}$.
\bar{s}	\bar{s} is complementary of s . For example, $\bar{s} = (\bullet \circ \circ) = C$ for $s = R = (\circ \bullet \bullet)$ and $\bar{s} = (\bullet \bullet \circ) = B$ for $s = Y = (\circ \circ \bullet)$.

$\lambda = 33\% (= (\frac{1}{3} + \frac{1}{3} + \frac{1}{3})/3)$ in [Fig. 1\(b\)](#) to $\lambda = 50\% (= (\frac{1}{3} + \frac{1}{3} + \frac{1}{3} + \frac{2}{3} + \frac{2}{3} + \frac{2}{3})/6)$ in [Fig. 2\(b\)](#), so that we can more clearly decode the secret. We use the definition of $\alpha = \frac{(h-l)}{m}$ to calculate the contrasts. All the contrasts of these two (2, 2)-CBW-VCS are $\alpha = \frac{(1-0)}{1} = 1$. Notice that a contrast of 1 cannot be achieved in the conventional VCS. Because any non-black pixel is regarded as a “white” pixel, it is possible for the CBW-VCS.

3. The proposed CBW-EVCS

Let the black and white matrices be $A_s^{c_1 c_2 \cdots c_n}$, where $c_i \in \{\bullet, \circ\}$ and $1 \leq i \leq n$, denotes the color on the i -th cover image C_i , and the secret pixel is $s \in \{\bullet, \circ\}$. For example, the $A_{\circ}^{\bullet\bullet}$ of a (2, 2)-CBW-EVCS is a base matrix that the pixels of both cover images c_1 and c_2 are all black colors, and their stacked result is a white color. We use the notations $(h_s \bullet (m - h_s)*))$ and $(l_s \bullet (m - l_s)*))$ to represent black and white pixels on shadows, where h_s and l_s are the blackness of black and white colors in cover images. Analogous to Ateniese et al.’s EVCS [1], the formal definition of the proposed (k, n) -CBW-EVCS is given as follows.

Definition 2. A (k, n) -CBW-EVCS is given by $(n \times m)$ black and white base matrices $A_{\bullet}^{c_1 c_2 \cdots c_n}$ and $A_{\circ}^{c_1 c_2 \cdots c_n}$ satisfying the following three conditions.

- (C-1) (Contrast condition): Given any c_1, c_2, \dots, c_n , and any qualified set X with $|X| \geq k$, we have that $w_{\bullet}(\text{add}(A_{\bullet}^{c_1 c_2 \cdots c_n} | X)) \geq h$ and $w_{\bullet}(\text{add}(A_{\circ}^{c_1 c_2 \cdots c_n} | X)) \leq l$. Note: h and l are the blackness of black and white colors in the reconstructed image and $0 \leq l < h \leq m$.
- (C-2) (Security condition): Given any c_1, c_2, \dots, c_n , and any forbidden set X with $|X| < k$, we have $D_{\bullet} = D(A_{\bullet}^{c_1 c_2 \cdots c_n} | X)$ and $D_{\circ} = D(A_{\circ}^{c_1 c_2 \cdots c_n} | X)$. Then, the collections D_{\bullet} and D_{\circ} are equivalent.
- (C-3) (Cover image condition): Given any $c_1, \dots, c_{i-1}, c_{i+1}, \dots, c_n$, $i \in \{1, n\}$ and the secret pixel $s \in \{\bullet, \circ\}$, we have that $w_{\bullet}(\text{add}(A_s^{c_1, \dots, c_i=\bullet, \dots, c_n} | i)) \geq h_s$ and $w_{\bullet}(\text{add}(A_s^{c_1, \dots, c_i=\circ, \dots, c_n} | i)) \leq l_s$, where $0 \leq l_s < h_s \leq m$. Note: the notation $(A_s^{c_1, \dots, c_n} | i)$ denotes the i -th row in $A_s^{c_1, \dots, c_n}$.

In subsequent sub sections, we describe (2, n)-CBW-EVCS and (k, n) -PB-CBW-EVCS. Some notations in our constructions are first described in [Table 2](#).

The 2-out-f-2 CBW-EVCS is a special case for $k = n = 2$. We first introduce (2, 2)-CBW-EVCS, and then describe the general $(2, n)$ -CBW-EVCS and (k, n) -CBW-EVCS. In all constructions, we show the generation of $A_s^{c_1 c_2 \dots c_n}$. The collection $C_s^{c_1 c_2 \dots c_n}$ is composed of all possible distribution matrices $A_s^{c_1 c_2 \dots c_n}$.

3.1. $(2, n)$ -CBW-EVCS

3.1.1. $(2, 2)$ -CBW-EVCS

There are three different ways to implement $(2, 2)$ -CBW-EVCS. In [Construction 1](#), [Construction 2](#), and [Construction 3](#), we show how to construct $(2, 2)$ -CBW-EVCS with background colors $\{RGB\}$, background colors $\{RGBCMY\}$, and two background colors $\{b_1 b_2\}$, respectively.

Construction 1. Let $A_{\bullet}^{c_1 c_2}$ and $A_{\circ}^{c_1 c_2}$ be $2 \times m$ black and white base matrices. Then, $(2, 2)$ -CBW-EVCS with background colors $\{RGB\}$, where $c_i \in \{\bullet, \circ\}$, $i = 1, 2$, has $A_{\bullet}^{c_1 c_2} = \begin{bmatrix} a_{\bullet}^{11} & a_{\bullet}^{12} \\ a_{\bullet}^{21} & a_{\bullet}^{22} \end{bmatrix}$ and $A_{\circ}^{c_1 c_2} = \begin{bmatrix} a_{\circ}^{11} & a_{\circ}^{12} \\ a_{\circ}^{21} & a_{\circ}^{22} \end{bmatrix}$. Let $S^{(1)}$ have $m' = 1$ and $\Sigma = \{R, G, B\}$, i.e., $S^{(1)} = \{R, G, B\}$. Elements of $A_{\bullet}^{c_1 c_2}$ and $A_{\circ}^{c_1 c_2}$ are obtained from Eq. [\(2\)](#).

$$\left\{ \begin{array}{l} \text{Case 1: } c_1 = \circ \text{ and } c_2 = \circ: \\ a_{\bullet}^{11} \leftarrow S^{(1)}, a_{\bullet}^{12} \leftarrow S^{(1)}, a_{\bullet}^{21} \leftarrow S^{(2)}, a_{\bullet}^{22} \leftarrow S^{(2)}; a_{\circ}^{11} \leftarrow S^{(1)}, a_{\circ}^{12} \leftarrow S^{(1)}, a_{\circ}^{21} = a_{\circ}^{11}, a_{\circ}^{22} \leftarrow S^{(2)} \\ \text{Case 2: } c_1 = \circ \text{ and } c_2 = \bullet: \\ a_{\bullet}^{11} \leftarrow S^{(1)}, a_{\bullet}^{12} \leftarrow S^{(1)}, a_{\bullet}^{21} \leftarrow S^{(2)}, a_{\bullet}^{22} = \bullet; a_{\circ}^{11} \leftarrow S^{(1)}, a_{\circ}^{12} \leftarrow S^{(1)}, a_{\circ}^{21} = a_{\circ}^{11}, a_{\circ}^{22} = \bullet \\ \text{Case 3: } c_1 = \bullet \text{ and } c_2 = \circ: \\ a_{\bullet}^{11} \leftarrow S^{(1)}, a_{\bullet}^{12} = \bullet, a_{\bullet}^{21} \leftarrow S^{(2)}, a_{\bullet}^{22} \leftarrow S^{(2)}; a_{\circ}^{11} \leftarrow S^{(1)}, a_{\circ}^{12} = \bullet, a_{\circ}^{21} = a_{\circ}^{11}, a_{\circ}^{22} \leftarrow S^{(2)} \\ \text{Case 4: } c_1 = \bullet \text{ and } c_2 = \bullet: \\ a_{\bullet}^{11} \leftarrow S^{(1)}, a_{\bullet}^{12} = \bullet, a_{\bullet}^{21} \leftarrow S^{(2)}, a_{\bullet}^{22} = \bullet; a_{\circ}^{11} \leftarrow S^{(1)}, a_{\circ}^{12} = \bullet, a_{\circ}^{21} = a_{\circ}^{11}, a_{\circ}^{22} = \bullet \end{array} \right. \quad (2)$$

According to the definition of $S^{(i)}$, we can obtain the sets $S^{(2)}$ in Eq. [\(2\)](#). For example, in Case 1 of Eq. [\(2\)](#), the sets $S^{(2)}$ in $(a_{\bullet}^{21} \leftarrow S^{(2)})$ and $(a_{\bullet}^{22} \leftarrow S^{(2)})$ are $\{S^{(1)} - \{a_{\bullet}^{11}\}\}$ and $\{S^{(1)} - \{a_{\bullet}^{12}\}\}$, respectively. In Case 3 of Eq. [\(2\)](#), the set $S^{(2)}$ in $(a_{\bullet}^{22} \leftarrow S^{(2)})$ is $\{S^{(1)} - \{a_{\bullet}^{12}\}\} = S^{(1)} - \{\bullet\} = S^{(1)}$. By the same argument, other sets $S^{(2)}$ can be obtained.

Considering the example of generating all possible $A_{\bullet}^{c_1 c_2}$, we always have $a_{\bullet}^{12} = a_{\bullet}^{22} = \bullet$. Because $a_{\bullet}^{11} \leftarrow S^{(1)}$ and $a_{\bullet}^{21} \leftarrow S^{(2)}$, we have 6 matrices of $A_{\bullet}^{c_1 c_2} = \begin{bmatrix} a_{\bullet}^{11} & a_{\bullet}^{12} \\ a_{\bullet}^{21} & a_{\bullet}^{22} \end{bmatrix} : \begin{bmatrix} R\bullet \\ G\bullet \end{bmatrix}, \begin{bmatrix} R\bullet \\ B\bullet \end{bmatrix}, \begin{bmatrix} G\bullet \\ R\bullet \end{bmatrix}, \begin{bmatrix} G\bullet \\ B\bullet \end{bmatrix}, \begin{bmatrix} B\bullet \\ R\bullet \end{bmatrix}, \text{ and } \begin{bmatrix} B\bullet \\ G\bullet \end{bmatrix}$. After column permutation, we have the collection $C_{\bullet}^{c_1 c_2} = \left\{ \begin{bmatrix} R\bullet \\ G\bullet \end{bmatrix} \begin{bmatrix} R\bullet \\ B\bullet \end{bmatrix} \begin{bmatrix} \bullet R \\ \bullet G \end{bmatrix} \begin{bmatrix} \bullet R \\ \bullet B \end{bmatrix} \begin{bmatrix} G\bullet \\ R\bullet \end{bmatrix} \begin{bmatrix} G\bullet \\ B\bullet \end{bmatrix} \begin{bmatrix} \bullet G \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet G \\ \bullet B \end{bmatrix} \begin{bmatrix} B\bullet \\ R\bullet \end{bmatrix} \begin{bmatrix} B\bullet \\ G\bullet \end{bmatrix} \begin{bmatrix} \bullet B \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet B \\ \bullet G \end{bmatrix} \right\}$. The collections $C_{\bullet}^{c_1 c_2}$ and $C_{\circ}^{c_1 c_2}$ of all distribution matrices are shown in [Table A1](#).

Theorem 1. The scheme from [Construction 1](#) is a $(2, 2)$ -CBW-EVCS with the pixel expansion $m = 2$. The blackness is $h = 2$, $l = 1$, $h_s = 1$, and $l_s = 0$.

Proof. There are 2 columns in $A_{\bullet}^{c_1 c_2}$ and $A_{\circ}^{c_1 c_2}$, and thus the pixel expansion is $m = 2$. We first prove that this $(2, 2)$ -CBW-EVCS satisfies contrast condition (C-1). Consider the case $c_1 = \circ, c_2 = \circ$. Because $a_{\bullet}^{11} \in S^{(1)}, a_{\bullet}^{21} \in \{S^{(1)} - \{a_{\bullet}^{11}\}\}, a_{\bullet}^{12} \in S^{(1)}$ and $a_{\bullet}^{22} \in \{S^{(1)} - \{a_{\bullet}^{11}\}\}$ in $A_{\bullet}^{c_1 c_2}$, we have $a_{\bullet}^{11} + a_{\bullet}^{21} = \bullet$ and $a_{\bullet}^{12} + a_{\bullet}^{22} = \bullet$, respectively. In fact, it can be verified that the stacked results of two rows in $A_{\bullet}^{c_1 c_2}$ and $A_{\circ}^{c_1 c_2}$ are $(\bullet\bullet)$ and $(\bullet\ast)$, respectively, where $\ast \in \{R, G, B\}$. Then, we have $w_{\bullet}(\text{add}(A_{\bullet}^{c_1 c_2} | X_2)) = 2$ and $w_{\bullet}(\text{add}(A_{\circ}^{c_1 c_2} | X_2)) = 1$, and thus $h = 2$ and $l = 1$, where the notation X_k denotes a set with any k participants.

For a proof of Condition (C-2), given any c_1 and c_2 , and a set X_1 (note: the threshold of $(2, 2)$ -CBW-EVCS is 2), we should prove that the collections D_{\bullet} and D_{\circ} are equivalent. Suppose $c_1 = \bullet$ and $c_2 = \bullet$. For the first row, because $a_{\bullet}^{11} \leftarrow S^{(1)}$ and $a_{\bullet}^{12} = \bullet$, we have $D_{\bullet} = \{[R\bullet][G\bullet][B\bullet][\bullet R][\bullet G][\bullet B]\}$, and because $a_{\circ}^{11} \leftarrow S^{(1)}$ and $a_{\circ}^{12} = \bullet$, we have $D_{\circ} = \{[R\bullet][G\bullet][B\bullet][\bullet R][\bullet G][\bullet B]\}$. For the second row, since $a_{\bullet}^{21} \leftarrow S^{(2)} = \{S^{(1)} - \{a_{\bullet}^{11}\}\}$ and after randomly choosing a_{\bullet}^{11} from $S^{(1)} = \{R, G, B\}$, the value of a_{\bullet}^{21} could be R, G , and B , and meantime a_{\bullet}^{22} is always \bullet . Thus, $D_{\bullet} = \{[R\bullet][G\bullet][B\bullet][\bullet R][\bullet G][\bullet B]\}$. On the other hand, because $a_{\circ}^{21} = a_{\circ}^{11}$ and a_{\circ}^{11} is chosen from $S^{(1)} = \{R, G, B\}$, the value of a_{\circ}^{21} could be R, G , and B , and meantime a_{\circ}^{22} is always \bullet . Thus, we have $D_{\circ} = \{[R\bullet][G\bullet][B\bullet][\bullet R][\bullet G][\bullet B]\}$. The above implies that D_{\bullet} and D_{\circ} are equivalent (i.e., contain the same matrices with the same frequencies) for $c_1 = \bullet$ and $c_2 = \bullet$. By the same argument, it can be verified that D_{\bullet} and D_{\circ} are also equivalent for other c_1 and c_2 . Next, we prove the condition (C-3). From Eq. [\(2\)](#), we have $w_{\bullet}(\text{add}(A_s^{c_1=\bullet, c_2} | 1)) = 1, w_{\bullet}(\text{add}(C_s^{c_1=\circ, c_2} | 1)) = 0, w_{\bullet}(\text{add}(A_s^{c_1=c_2=\bullet} | 2)) = 1$, and $w_{\bullet}(\text{add}(C_s^{c_1=c_2=\circ} | 2)) = 0$. Thus, we have $h_s = 1$ and $l_s = 0$. \square

In [Construction 1](#), the background colors of shadows are colored pixels in $\{R, G, B\}$. We can construct another $(2, 2)$ -CBW-EVCS with $\{R, G, B, C, M, Y\}$, as shown in [Construction 2](#). Same as the $(2, 2)$ -CBW-VCS (see [Example 1](#)), λ in

(1) choose any n ($\leq 3m'$) elements from $S^{(1)}$ to form a new set $S^{(1)}$ (2) for $i = 1$ to n do $\{a_i^i \leftarrow S^{(i)}\}$; (3) for $i = 1$ to n do if $c_i = \bullet$ then $a_i^i = (\overbrace{\bullet \cdots \bullet}^{m'})$;
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Fig. 3. Generation of $A^{c_1 c_2 \cdots c_n}$.

Construction 1 can be enhanced from 33% to 50% in **Construction 2**, so that we can decode the clearer image. Meanwhile, the clearness of shadow image is enhanced.

Construction 2. Let $A_{\bullet}^{c_1 c_2}$ and $A_{\circ}^{c_1 c_2}$ be $2 \times m$ black and white matrices. Then, (2, 2)-CBW-EVCS with background colors $\{RGBCMY\}$, where $c_i \in \{\bullet, \circ\}$, $i = 1, 2$, has $A_{\bullet}^{c_1 c_2} = \begin{bmatrix} a_{\bullet}^{11} & a_{\bullet}^{12} \\ a_{\bullet}^{21} & a_{\bullet}^{22} \end{bmatrix}$ and $A_{\circ}^{c_1 c_2} = \begin{bmatrix} a_{\circ}^{11} & a_{\circ}^{12} \\ a_{\circ}^{21} & a_{\circ}^{22} \end{bmatrix}$. Let $S^{(1)}$ have $m' = 1$ and $\Sigma = \{R, G, B, C, M, Y\}$, i.e., $S^{(1)} = \{R, G, B, C, M, Y\}$. Elements of $A_{\bullet}^{c_1 c_2}$ and $A_{\circ}^{c_1 c_2}$ are obtained from Eq. (3).

$$\left\{ \begin{array}{l} \text{Case 1: } c_1 = \circ \text{ and } c_2 = \circ : \\ a_{\bullet}^{11} \leftarrow S^{(1)}, a_{\bullet}^{12} \leftarrow S^{(1)}, a_{\bullet}^{21} = \overline{a_{\bullet}^{11}}, a_{\bullet}^{22} = \overline{a_{\bullet}^{12}}; a_{\circ}^{11} \leftarrow S^{(1)}, a_{\circ}^{12} \leftarrow S^{(1)}; a_{\circ}^{21} = a_{\circ}^{11}, a_{\circ}^{22} = \overline{a_{\circ}^{12}} \\ \text{Case 2: } c_1 = \circ \text{ and } c_2 = \bullet : \\ a_{\bullet}^{11} \leftarrow S^{(1)}, a_{\bullet}^{12} \leftarrow S^{(1)}, a_{\bullet}^{21} = \overline{a_{\bullet}^{11}}, a_{\bullet}^{22} = \bullet; a_{\circ}^{11} \leftarrow S^{(1)}, a_{\circ}^{12} \leftarrow S^{(1)}; a_{\circ}^{21} = a_{\circ}^{11}, a_{\circ}^{22} = \bullet \\ \text{Case 3: } c_1 = \bullet \text{ and } c_2 = \circ : \\ a_{\bullet}^{11} \leftarrow S^{(1)}, a_{\bullet}^{12} = \bullet, a_{\bullet}^{21} = \overline{a_{\bullet}^{11}}, a_{\bullet}^{22} \leftarrow S^{(2)}; a_{\circ}^{11} \leftarrow S^{(1)}, a_{\circ}^{12} = \bullet; a_{\circ}^{21} = a_{\circ}^{11}, a_{\circ}^{22} \leftarrow S^{(2)} \\ \text{Case 4: } c_1 = \bullet \text{ and } c_2 = \bullet : \\ a_{\bullet}^{11} \leftarrow S^{(1)}, a_{\bullet}^{12} = \bullet, a_{\bullet}^{21} = \overline{a_{\bullet}^{11}}, a_{\bullet}^{22} = \bullet; a_{\circ}^{11} \leftarrow S^{(1)}, a_{\circ}^{12} = \bullet; a_{\circ}^{21} = a_{\circ}^{11}, a_{\circ}^{22} = \bullet \end{array} \right. \quad (3)$$

According to the definition of $S^{(i)}$, in Case 3 of Eq. (3), the sets $S^{(2)}$ in $(a_{\bullet}^{22} \leftarrow S^{(2)})$ and $(a_{\circ}^{22} \leftarrow S^{(2)})$ are $\{S^{(1)} - \{a_{\bullet}^{12}\}\} = S^{(1)} - \{\bullet\} = S^{(1)}$ and $\{S^{(1)} - \{a_{\circ}^{12}\}\} = S^{(1)} - \{\bullet\} = S^{(1)}$.

Obviously, it is possible to construct (2, 2)-CBW-EVCS with only two colors, as shown in **Construction 3**.

Construction 3. Let $A_{\bullet}^{c_1 c_2}$ and $A_{\circ}^{c_1 c_2}$ be $2 \times m$ black and white matrices. Then, (2, 2)-CBW-EVCS with two background colors $\{b_1 b_2\} = \{RG\}, \{RB\}, \{GB\}, \{CR\}, \{MG\}$, or $\{YB\}$, where $c_i \in \{\bullet, \circ\}$, $i = 1, 2$, has $A_{\bullet}^{c_1 c_2} = \begin{bmatrix} a_{\bullet}^{11} & a_{\bullet}^{12} \\ a_{\bullet}^{21} & a_{\bullet}^{22} \end{bmatrix}$ and $A_{\circ}^{c_1 c_2} = \begin{bmatrix} a_{\circ}^{11} & a_{\circ}^{12} \\ a_{\circ}^{21} & a_{\circ}^{22} \end{bmatrix}$. Let $S^{(1)}$ have $m' = 1$ and $\Sigma = \{b_1, b_2\}$, i.e., $S^{(1)} = \{b_1, b_2\}$. Elements of $A_{\bullet}^{c_1 c_2}$ and $A_{\circ}^{c_1 c_2}$ can be obtained by the same approach in Eq. (2).

The proofs of **Construction 2** and **Construction 3** being (2, 2)-CBW-EVCS are straight forward, and exactly the same as the proof of **Theorem 1**.

3.1.2. (2, n)-CBW-EVCS

Construction 4. Let $A_{\bullet}^{c_1 c_2 \cdots c_n}$ and $A_{\circ}^{c_1 c_2 \cdots c_n}$ be $n \times m$ black and white matrices. Then, (2, n)-CBW-EVCS with background colors $\{RGB\}$, where $c_i \in \{\bullet, \circ\}$, $i = 1, 2, \dots, n$, has $A_{\bullet}^{c_1 c_2 \cdots c_n} = (A_{\bullet} \parallel A^{c_1 c_2 \cdots c_n})$ and $A_{\circ}^{c_1 c_2 \cdots c_n} = (A_{\circ} \parallel A^{c_1 c_2 \cdots c_n})$, where A_{\bullet} , A_{\circ} , and $A^{c_1 c_2 \cdots c_n}$ are obtained from the following. Let $S^{(1)}$ have $m' = \lceil \log_3 n \rceil$ characters over the alphabet $\Sigma = \{R, G, B\}$, i.e.,

$S^{(1)} = \{\overbrace{R \cdots R}^{m'}, \overbrace{R \cdots G}^{m'}, \overbrace{R \cdots B}^{m'}, \dots, \overbrace{B \cdots B}^{m'}\}$ with $3^{m'}$ elements. We first choose any n ($\leq 3^{m'}$) elements from $S^{(1)}$ to

form a new set $S^{(1)}$. Black and white matrices $A_{\bullet} = \begin{bmatrix} a_{\bullet}^1 \\ a_{\bullet}^2 \\ \vdots \\ a_{\bullet}^n \end{bmatrix}$ and $A_{\circ} = \begin{bmatrix} a_{\circ}^1 \\ a_{\circ}^2 \\ \vdots \\ a_{\circ}^n \end{bmatrix}$ are obtained from De Prisco and De Santis's

(2, n)-CBW-VCS (Construction 5.5 in [7]) with the pixel expansion m' , which are constructed from Eq. (4).

$$\left\{ \begin{array}{l} (1) a_{\bullet}^1 \leftarrow S^{(1)}, a_{\bullet}^2 \leftarrow S^{(2)}, \dots, a_{\bullet}^n \leftarrow S^{(n)} \\ (2) a_{\circ}^1 \leftarrow S^{(1)}, a_{\circ}^2 = a_{\circ}^1, \dots, a_{\circ}^n = a_{\circ}^1. \end{array} \right. \quad (4)$$

For the matrices of cover images $A^{c_1 c_2 \cdots c_n}$, we use another $S^{(1)}$ of $m' = \lceil n/3 \rceil$ characters with $(m' - 1)\bullet$ and 1 alphabet $\{m'-1, m'-1, m'-1, m'-1, m'-1, m'-1\}$ in $\Sigma = \{R, G, B\}$, i.e., $S^{(1)} = \{\overbrace{\bullet \cdots \bullet}^{m'-1} R, \overbrace{\bullet \cdots \bullet}^{m'-1} G, \overbrace{\bullet \cdots \bullet}^{m'-1} B, \dots, \overbrace{R \bullet \cdots \bullet}^{m'-1}, \overbrace{G \bullet \cdots \bullet}^{m'-1}, \overbrace{B \bullet \cdots \bullet}^{m'-1}\}$ with $3^{m'}$ elements. Then, $A^{c_1 c_2 \cdots c_n} = \begin{bmatrix} a^1 \\ a^2 \\ \vdots \\ a^n \end{bmatrix}$, where $n \leq 3^{m'}$, with the pixel expansion m' can be constructed from the algorithm in Fig. 3.

Lemma 1. The matrix of cover image $A^{c_1 c_2 \dots c_n}$ has $w_{\bullet}(\text{add}(A^{c_1, \dots, c_i = \bullet, \dots, c_n} | i)) = \lceil n/3 \rceil$ and $w_{\bullet}(\text{add}(A^{c_1, \dots, c_i = o, \dots, c_n} | i)) = (\lceil n/3 \rceil - 1)$ where $i = 1, 2, \dots, n$. Also, $w_{\bullet}(\text{add}(A^{c_1 c_2 \dots c_n} | X_{k \geq 2})) = \lceil n/3 \rceil$.

Proof. According to the generation of $A^{c_1 c_2 \dots c_n}$ in Fig. 3, we have the element a^i chosen from $S^{(i)}$ for $c_i = o$. From the definition of $S^{(1)} = \{\overbrace{\bullet \dots \bullet}^{\lceil n/3 \rceil - 1}, \overbrace{\bullet \dots \bullet}^{\lceil n/3 \rceil - 1}\}$, we have $w_{\bullet}(\text{add}(A^{c_1, \dots, c_i = o, \dots, c_n} | i)) = w_{\bullet}(a^i) = \lceil n/3 \rceil - 1$. Because the element a^i is $(\overbrace{\bullet \dots \bullet}^{\lceil n/3 \rceil})$ for $c_i = \bullet$, we have $w_{\bullet}(\text{add}(A^{c_1, \dots, c_i = \bullet, \dots, c_n} | i)) = w_{\bullet}(a^i) = \lceil n/3 \rceil$. Obviously, it can be easily verified that $\text{add}(A^{c_1 c_2 \dots c_n} | X_{k \geq 2}) = (\overbrace{\bullet \dots \bullet}^{\lceil n/3 \rceil})$, and thus we have $w_{\bullet}(\text{add}(A^{c_1 c_2 \dots c_n} | X_{k \geq 2})) = \lceil n/3 \rceil$. \square

Theorem 2. The scheme from Construction 4 is a $(2, n)$ -CBW-EVCS with the pixel expansion $m = \lceil \log_3 n \rceil + \lceil n/3 \rceil$. The blackness is $h \geq (\lceil n/3 \rceil + 1)$, $l = \lceil n/3 \rceil$, $h_s = \lceil n/3 \rceil$, and $l_s = (\lceil n/3 \rceil - 1)$.

Proof. Because $A_{\bullet}^{c_1 c_2 \dots c_n} = (A_{\bullet} \parallel A^{c_1 c_2 \dots c_n})$ and $A_{\circ}^{c_1 c_2 \dots c_n} = (A_{\circ} \parallel A^{c_1 c_2 \dots c_n})$, we have $|A_{\bullet}^{c_1 c_2 \dots c_n}| = |A_{\bullet}| + |A^{c_1 c_2 \dots c_n}| = \lceil \log_3 n \rceil + \lceil n/3 \rceil$ and $|A_{\circ}^{c_1 c_2 \dots c_n}| = |A_{\circ}| + |A^{c_1 c_2 \dots c_n}| = \lceil \log_3 n \rceil + \lceil n/3 \rceil$. The pixel expansion is $m = \lceil \log_3 n \rceil + \lceil n/3 \rceil$. We first prove that $(2, n)$ -CBW-EVCS satisfies contrast condition (C-1). Because A_{\bullet} and A_{\circ} are base matrices of De Prisco and De Santis's $(2, n)$ -CBW-VCS, we have $w_{\bullet}(\text{add}(A_{\bullet} | X_2)) \geq 1$ and $w_{\bullet}(\text{add}(A_{\circ} | X_2)) = 0$. From Lemma 1, we have $w_{\bullet}(\text{add}(A^{c_1 c_2 \dots c_n} | X_2)) = \lceil n/3 \rceil$. The values of $w_{\bullet}(\text{add}(A_{\bullet}^{c_1 c_2 \dots c_n} | X_2))$ and $w_{\bullet}(\text{add}(A_{\circ}^{c_1 c_2 \dots c_n} | X_2))$ are determined from Eq. (5.1) and Eq. (5.2), respectively. Finally, we have the blackness $h = w_{\bullet}(\text{add}(A_{\bullet}^{c_1 c_2 \dots c_n} | X_2)) \geq (\lceil n/3 \rceil + 1)$ and $l = w_{\bullet}(\text{add}(A_{\circ}^{c_1 c_2 \dots c_n} | X_2)) = \lceil n/3 \rceil$.

$$\left\{ \begin{array}{l} w_{\bullet}(\text{add}(A_{\bullet}^{c_1 c_2 \dots c_n} | X_2)) = w_{\bullet}(\text{add}((A_{\bullet} \parallel A^{c_1 c_2 \dots c_n}) | X_2)) = w_{\bullet}(\text{add}(A_{\bullet} | X_2)) \\ \quad + w_{\bullet}(\text{add}(A^{c_1 c_2 \dots c_n} | X_2)) \geq (\lceil n/3 \rceil + 1)(5 - 1) \end{array} \right. \quad (5.1)$$

$$\left\{ \begin{array}{l} w_{\bullet}(\text{add}(A_{\circ}^{c_1 c_2 \dots c_n} | X_2)) = w_{\bullet}(\text{add}((A_{\circ} \parallel A^{c_1 c_2 \dots c_n}) | X_2)) \\ \quad = w_{\bullet}(\text{add}(A_{\circ} | X_2)) + w_{\bullet}(\text{add}(A^{c_1 c_2 \dots c_n} | X_2)) = \lceil n/3 \rceil \end{array} \right. \quad (5.2)$$

For a proof of Condition (C-2), by definition, we can derive $D_{\bullet} = D(A_{\bullet}^{c_1 c_2 \dots c_n} | X)$ and $D_{\circ} = D(A_{\circ}^{c_1 c_2 \dots c_n} | X)$, where $|X| < 2$, as follows.

$$\left\{ \begin{array}{l} D(A_{\bullet}^{c_1 c_2 \dots c_n} | X) = D((A_{\bullet} \parallel A^{c_1 c_2 \dots c_n}) | X) = D(A_{\bullet} | X) \cup D(A^{c_1 c_2 \dots c_n} | X), \\ D(A_{\circ}^{c_1 c_2 \dots c_n} | X) = D((A_{\circ} \parallel A^{c_1 c_2 \dots c_n}) | X) = D(A_{\circ} | X) \cup D(A^{c_1 c_2 \dots c_n} | X). \end{array} \right. \quad (6)$$

Collections $D(A_{\bullet} | X)$ and $D(A_{\circ} | X)$ are equivalent since they comes from De Prisco and De Santis's $(2, n)$ -CBW-VCS. Therefore, $D_{\bullet} = D(A_{\bullet}^{c_1 c_2 c_3} | X)$ and $D_{\circ} = D(A_{\circ}^{c_1 c_2 c_3} | X)$ are equivalent. Next, we prove the condition (C-3). Because A_{\bullet} and A_{\circ} are the base matrices of De Prisco and De Santis's $(2, n)$ -CBW-VCS, we have $w_{\bullet}(\text{add}(A_{\bullet} | i)) = w_{\bullet}(\text{add}(A_{\circ} | i)) = 0$. We can derive $w_{\bullet}(\text{add}(A_s^{c_1 c_2 \dots c_n} | i)) = w_{\bullet}(\text{add}(A_s \parallel A^{c_1 c_2 \dots c_n} | i)) = w_{\bullet}(\text{add}(A_s | i)) + w_{\bullet}(\text{add}(A^{c_1 c_2 \dots c_n} | i)) = w_{\bullet}(\text{add}(A^{c_1 c_2 \dots c_n} | i))$. By Lemma 1, we have $w_{\bullet}(\text{add}(A^{c_1, \dots, c_i = \bullet, \dots, c_n} | i)) = \lceil n/3 \rceil$ and $w_{\bullet}(\text{add}(A^{c_1, \dots, c_i = o, \dots, c_n} | i)) = (\lceil n/3 \rceil - 1)$, i.e., $h_s = \lceil n/3 \rceil$, and $l_s = (\lceil n/3 \rceil - 1)$. \square

Construction 4 shows a $(2, n)$ -CBW-EVCS. However, a PB-VCS has the clearer reconstructed image via perfect reconstruction of black pixels. Therefore, we propose a $(2, n)$ -PB-CBW-EVCS.

Construction 5. Let $A_{\bullet}^{c_1 c_2 \dots c_n}$ and $A_{\circ}^{c_1 c_2 \dots c_n}$ be $n \times m$ black and white matrices. Then, $(2, n)$ -PB-CBW-EVCS with background colors $\{RGB\}$, where $c_i \in \{\bullet, o\}$, $i = 1, 2, \dots, n$, has $A_{\bullet}^{c_1 c_2 \dots c_n} = (A_{\bullet} \parallel A^{c_1 c_2 \dots c_n})$ and $A_{\circ}^{c_1 c_2 \dots c_n} = (A_{\circ} \parallel A^{c_1 c_2 \dots c_n})$, where A_{\bullet} and A_{\circ} are obtained from the following. The matrix $A^{c_1 c_2 \dots c_n}$ is the same as that in Construction 4.

Let $S^{(1)}$ be the set $S^{(1)}$ used to generate $A^{c_1 c_2 \dots c_n}$ in Fig. 3. We first choose any n ($\leq 3^{m'}$) elements from $S^{(1)}$ to form

a new set $S^{(1)}$. Black and white matrices $A_{\bullet} = \begin{bmatrix} a_{\bullet}^1 \\ a_{\bullet}^2 \\ \vdots \\ a_{\bullet}^n \end{bmatrix}$ and $A_{\circ} = \begin{bmatrix} a_{\circ}^1 \\ a_{\circ}^2 \\ \vdots \\ a_{\circ}^n \end{bmatrix}$ are base matrices of De Prisco and De Santis's $(2, n)$ -PB-CBW-VCS (Construction 7.1 in [7]), which can also be constructed by the same approach in Eq. (4).

The proof of Construction 5 being a $(2, n)$ -PB-CBW-VCS is similar to the proof of Theorem 2. This $(2, n)$ -PB-CBW-VCS has $m = 2 \times \lceil n/3 \rceil$, $h = 2 \times \lceil n/3 \rceil$, $l = (2 \times \lceil n/3 \rceil - 1)$, $h_s = (2 \times \lceil n/3 \rceil - 1)$ and $l_s = (2 \times \lceil n/3 \rceil - 2)$.

3.2. (k, n) -PB-CBW-EVCS

This section shows a general (k, n) -PB-CBW-EVCS for any k and n , where $k \leq n$. This scheme is constructed from the (k, n) -PB-VCS.

Construction 6. Let $A_{\bullet}^{c_1 c_2 \dots c_n}$ and $A_{\circ}^{c_1 c_2 \dots c_n}$ be $n \times m$ black and white matrices. Then, (k, n) -PB-CBW-EVCS with $2 < k \leq n$ and background colors $\{RGBCMY \bullet \circ\}$, where $c_i \in \{\bullet, \circ\}$, $i = 1, 2, \dots, n$, has $A_{\bullet}^{c_1 c_2 \dots c_n} = (A_{\bullet} \parallel \overbrace{A^{c_1 c_2 \dots c_n} \parallel \dots \parallel A^{c_1 c_2 \dots c_n}}^d)$ and $A_{\circ}^{c_1 c_2 \dots c_n} = (A_{\circ} \parallel \overbrace{A^{c_1 c_2 \dots c_n} \parallel \dots \parallel A^{c_1 c_2 \dots c_n}}^d)$, where A_{\bullet} and A_{\circ} are obtained from the following. The matrix $A^{c_1 c_2 \dots c_n}$ is the same as that in Construction 4.

Matrices A_{\bullet} and A_{\circ} are the black and white matrices of De Prisco and De Santis's (k, n) -PB-CBW-VCS (Construction 7.2 in [7]), which are transformed from the (k, n) -PB-VCS. This transformation is briefly described as follows. By adding $(\lceil m_B/3 \rceil \times 3) - m_B$ all-• columns into base matrices of (k, n) -PB-VCS where m_B is the pixel expansion of this (k, n) -PB-VCS, we can keep the pixel expansion a multiple of 3. When generating shadows, we randomly permute the columns in matrices and group every triple to form a color into 8 full intensity colors. Finally De Prisco and De Santis's (k, n) -PB-CBW-VCS has the pixel expansion $\lceil m_B/3 \rceil$. Let the maximum number and minimum number of • of a row in this (k, n) -PB-CBW-VCS be n_{\max} and n_{\min} . Then, we concatenate $A^{c_1 c_2 \dots c_n} d$ times to A_{\bullet} and A_{\circ} to form $A_{\bullet}^{c_1 c_2 \dots c_n}$ and $A_{\circ}^{c_1 c_2 \dots c_n}$, respectively, where $d = (n_{\max} - n_{\min} + 1)$.

Lemma 2. The matrix $\overbrace{(A^{c_1 c_2 \dots c_n} \parallel \dots \parallel A^{c_1 c_2 \dots c_n}}^d)$ has $w_{\bullet}(\text{add}((\overbrace{A^{c_1, \dots, c_i=\bullet, \dots, c_n} \parallel \dots \parallel A^{c_1, \dots, c_i=\bullet, \dots, c_n}}^d) \mid i)) = (\lceil n/3 \rceil \times d)$ and $w_{\bullet}(\text{add}((\overbrace{A^{c_1, \dots, c_i=\circ, \dots, c_n} \parallel \dots \parallel A^{c_1, \dots, c_i=\circ, \dots, c_n}}^d) \mid i)) = ((\lceil n/3 \rceil - 1) \times d)$ for $1 \leq i \leq n$. Also, $w_{\bullet}(\text{add}(\overbrace{A^{c_1 c_2 \dots c_n} \parallel \dots \parallel A^{c_1 c_2 \dots c_n}}^d \mid X_{k \geq 2})) = (\lceil n/3 \rceil \times d)$.

Proof. Obviously, we have $w_{\bullet}(\text{add}(\overbrace{A^{c_1 c_2 \dots c_n} \parallel \dots \parallel A^{c_1 c_2 \dots c_n}}^d) \mid i) = w_{\bullet}(\underbrace{a^i \dots a^i}_d) = \underbrace{w_{\bullet}(a^i) + \dots + w_{\bullet}(a^i)}_d = w_{\bullet}(a^i) \times d$. Because $w_{\bullet}(a^i) = \lceil n/3 \rceil$ and $w_{\bullet}(a^i) = (\lceil n/3 \rceil - 1)$ for $c_i = \bullet$ and $c_i = \circ$ from Lemma 1, we have $w_{\bullet}(\text{add}((\overbrace{A^{c_1, \dots, c_i=\bullet, \dots, c_n} \parallel \dots \parallel A^{c_1, \dots, c_i=\bullet, \dots, c_n}}^d) \mid i)) = (\lceil n/3 \rceil \times d)$ and $w_{\bullet}(\text{add}((\overbrace{A^{c_1, \dots, c_i=\circ, \dots, c_n} \parallel \dots \parallel A^{c_1, \dots, c_i=\circ, \dots, c_n}}^d) \mid i)) = ((\lceil n/3 \rceil - 1) \times d)$. It can be verified that $w_{\bullet}(\text{add}(M \mid X_{k \geq 2})) = w_{\bullet}(\underbrace{(\bullet \dots \bullet)}_{\lceil n/3 \rceil} \underbrace{(\bullet \dots \bullet)}_{\lceil n/3 \rceil} \dots \underbrace{(\bullet \dots \bullet)}_{\lceil n/3 \rceil}) = (\lceil n/3 \rceil \times d)$. \square

Theorem 3. The scheme from Construction 6 is a (k, n) -PB-CBW-EVCS with $m = (\lceil m_B/3 \rceil + (\lceil n/3 \rceil \times d))$, $h = (\lceil m_B/3 \rceil + (\lceil n/3 \rceil \times d))$, $l \leq (\lceil m_B/3 \rceil + (\lceil n/3 \rceil \times d)) - 1$, $h_s \geq ((\lceil n/3 \rceil \times d) + n_{\min})$, and $l_s \leq ((\lceil n/3 \rceil \times d) + n_{\min} - 1)$.

Proof. Because $A_{\bullet}^{c_1 c_2 \dots c_n} = (A_{\bullet} \parallel \overbrace{A^{c_1 c_2 \dots c_n} \parallel \dots \parallel A^{c_1 c_2 \dots c_n}}^d)$ and $A_{\circ}^{c_1 c_2 \dots c_n} = (A_{\circ} \parallel \overbrace{A^{c_1 c_2 \dots c_n} \parallel \dots \parallel A^{c_1 c_2 \dots c_n}}^d)$, we have $|A_{\bullet}^{c_1 c_2 \dots c_n}| = |A_{\bullet}| + |\overbrace{A^{c_1 c_2 \dots c_n} \parallel \dots \parallel A^{c_1 c_2 \dots c_n}}^d| = (\lceil m_c/3 \rceil + (\lceil n/3 \rceil \times d))$, and $|A_{\circ}^{c_1 c_2 \dots c_n}| = |A_{\circ}| + |\overbrace{A^{c_1 c_2 \dots c_n} \parallel \dots \parallel A^{c_1 c_2 \dots c_n}}^d| = (\lceil m_B/3 \rceil + (\lceil n/3 \rceil \times d))$. The pixel expansion is $m = (\lceil m_B/3 \rceil + (\lceil n/3 \rceil \times d))$. We first prove that (k, n) -PB-CBW-EVCS satisfies contrast condition (C-1). Because A_{\bullet} and A_{\circ} are base matrices of De Prisco and De Santis's (k, n) -PB-CBW-VCS, we have $w_{\bullet}(\text{add}(A_{\bullet} \mid X_k)) = \lceil m_B/3 \rceil$ and $w_{\bullet}(\text{add}(A_{\circ} \mid X_k)) \leq (\lceil m_B/3 \rceil - 1)$. Additionally, from Lemma 2, we have $w_{\bullet}(\text{add}(A_{\bullet}^{c_1 c_2 \dots c_n} \mid X_{k \geq 2})) = (\lceil n/3 \rceil \times d)$. The values of $h = w_{\bullet}(\text{add}(A_{\bullet}^{c_1 c_2 \dots c_n} \mid X_k))$ and $l = w_{\bullet}(\text{add}(A_{\circ}^{c_1 c_2 \dots c_n} \mid X_k))$ are derived from Eq. (7.1) and Eq. (7.2), respectively.

$$\left\{ \begin{array}{l} h = w_{\bullet}(\text{add}(A_{\bullet}^{c_1 c_2 \dots c_n} \mid X_k)) = w_{\bullet}(\text{add}((A_{\bullet} \parallel \overbrace{A^{c_1 c_2 \dots c_n} \parallel \dots \parallel A^{c_1 c_2 \dots c_n}}^d) \mid X_k)) \\ \quad = w_{\bullet}(\text{add}(A_{\bullet} \mid X_k)) + w_{\bullet}(\text{add}((\overbrace{A^{c_1 c_2 \dots c_n} \parallel \dots \parallel A^{c_1 c_2 \dots c_n}}^d) \mid X_k)) = (\lceil m_c/3 \rceil + (\lceil n/3 \rceil \times d)) \end{array} \right. \quad (7.1)$$

$$\left\{ \begin{array}{l} l = w_{\bullet}(\text{add}(A_{\circ}^{c_1 c_2 \dots c_n} \mid X_k)) = w_{\bullet}(\text{add}((A_{\circ} \parallel \overbrace{A^{c_1 c_2 \dots c_n} \parallel \dots \parallel A^{c_1 c_2 \dots c_n}}^d) \mid X_k)) \\ \quad = w_{\bullet}(\text{add}(A_{\circ} \mid X_k)) + w_{\bullet}(\text{add}((\overbrace{A^{c_1 c_2 \dots c_n} \parallel \dots \parallel A^{c_1 c_2 \dots c_n}}^d) \mid X_k)) \leq (\lceil m_c/3 \rceil + (\lceil n/3 \rceil \times d)) - 1 \end{array} \right. \quad (7.2)$$

By the same argument in the proof of Theorem 2, $D_{\bullet} = D(A_{\bullet}^{c_1 c_2 \dots c_n} \mid X)$ and $D_{\circ} = D(A_{\circ}^{c_1 c_2 \dots c_n} \mid X)$, where $|X| < k$, are equivalent. Thus condition (C-2) is satisfied. Next, we prove the condition (C-3). Since A_{\bullet} and A_{\circ} are the base matrices of De Prisco and De Santis's (k, n) -PB-CBW-VCS, we have $n_{\min} \leq w_{\bullet}(A_{\bullet} \mid i) \leq n_{\max}$. From Lemma 2,

$w_{\bullet}(\text{add}((\overbrace{A^{c_1, \dots, c_i=\bullet, \dots, c_n} \| \dots \| A^{c_1, \dots, c_i=\bullet, \dots, c_n}}^d) | i)) = ([n/3] \times d)$ and $w_{\bullet}(\text{add}((\overbrace{A^{c_1, \dots, c_i=o, \dots, c_n} \| \dots \| A^{c_1, \dots, c_i=o, \dots, c_n}}^d) | i)) = (([n/3] - 1) \times d)$, we can determine the blackness of shadows $h_s = w_{\bullet}(\text{add}(A_s^{c_1, \dots, c_i=\bullet, \dots, c_n} | i))$ and $l_s = w_{\bullet}(\text{add}(A_s^{c_1, \dots, c_i=o, \dots, c_n} | i))$ from the following equation.

$$\begin{cases} h_s = w_{\bullet}(\text{add}(A_s^{c_1, \dots, c_i=\bullet, \dots, c_n} | i)) = w_{\bullet}(\text{add}(A_s \| A^{c_1, \dots, c_i=\bullet, \dots, c_n}) | i)) \\ = w_{\bullet}(\text{add}(A_s | i)) + w_{\bullet}(\text{add}(A^{c_1, \dots, c_i=\bullet, \dots, c_n} | i)) \\ \geq ([n/3] \times d) + n_{\min} \end{cases} \quad (8.1)$$

$$\begin{cases} l_s = w_{\bullet}(\text{add}(A_s^{c_1, \dots, c_i=o, \dots, c_n} | i)) = w_{\bullet}(\text{add}((A_s \| A^{c_1, \dots, c_i=o, \dots, c_n}) | i)) \\ = w_{\bullet}(\text{add}(A_s | i)) + w_{\bullet}(\text{add}(A^{c_1, \dots, c_i=o, \dots, c_n} | i)) \\ \leq n_{\max} + ([n/3] - 1) \times d = ([n/3] \times d) + n_{\max} - d = ([n/3] \times d) + n_{\min} - 1 \end{cases} \quad \square \quad (8.2)$$

3.3. Examples

Five examples are given to easily understand the proposed constructions. For $(2, n)$ -CWB-EVCS, we shows the $(2, 2)$ -CWB-EVCSs from [Constructions 1, 2](#), and [3](#), the $(2, 3)$ -CWB-EVCS and the $(2, 4)$ -CWB-EVCS from [Construction 4](#), and the $(2, 4)$ -PB-CWB-EVCS from [Construction 5](#). For (k, n) -PB-CWB-EVCS, we show a $(3, 3)$ -PB-CWB-EVCS from [Construction 6](#).

Example 2. Construct $(2, 2)$ -CBW-EVCSs by [Construction 1](#), [Construction 2](#), and [Construction 3](#).

From [Table A1](#), it is observed that this $(2, 2)$ -CBW-EVCS has $(1 \bullet 1*)$ and $(2*)$, where $* \in \{R, G, B\}$, in shadows to represent the black and white color, i.e., $h_s = 1$ and $l_s = 0$, s. The black and white colors in reconstructed image are $(2\bullet)$ and $(1 \bullet 1*)$, i.e., $h = 2$ and $l = 1$. By the definition of Naor and Sham'r's contrast, all the contrasts of these three constructions are $\alpha = \frac{(h-l)}{m} = \frac{(2-1)}{2} = \frac{1}{2}$ and $\alpha_s = \frac{(h_s-l_s)}{m} = \frac{(1-0)}{2} = \frac{1}{2}$ for the reconstructed image and the shadows. The collections $C_{\bullet}^{c_1 c_2}$ and $C_o^{c_1 c_2}$ of all distribution matrices using [Construction 2](#) and [Construction 3](#) with $\{RG\}$ are shown in [Table A2](#) and [Table A3](#), respectively. [Construction 2](#) has $\lambda = 50\%$, while [Construction 3](#) may have $\lambda = 50\%$ or $\lambda = 33\%$. [Construction 3](#) with $\{RG\}$, $\{RB\}$, and $\{GB\}$ has $\lambda = 33\%$, while [Construction 3](#) with $\{CR\}$, $\{MG\}$, and $\{YB\}$ has the high $\lambda = (\frac{1}{3} + \frac{2}{3})/2 = 50\%$.

Example 3. Construct a $(2, 3)$ -CBW-EVCS by [Construction 4](#).

For $n = 3$, [Construction 4](#) has the pixel expansion $m = ([\log_3 3] + [3/3]) = 2$. The matrices A_{\bullet} and A_o of $(2, 3)$ -CBW-VCS are $A_{\bullet} = \left\{ \begin{bmatrix} R \\ G \\ B \end{bmatrix} \begin{bmatrix} B \\ R \\ G \end{bmatrix} \begin{bmatrix} G \\ B \\ R \end{bmatrix} \right\}$ and $A_o = \left\{ \begin{bmatrix} R \\ R \\ G \\ G \\ B \\ B \end{bmatrix} \begin{bmatrix} G \\ G \\ B \\ B \\ R \\ R \end{bmatrix} \begin{bmatrix} B \\ B \\ R \\ R \\ G \\ G \end{bmatrix} \right\}$. The matrices of cover mage $A^{c_1 c_2 c_3}$ are given in [Table A4](#). Consider the example $c_1 = o$, $c_2 = o$, $c_3 = \bullet$. Matrices $A_o^{\circ\circ\bullet} = (A_{\bullet} \| A^{\circ\circ\bullet})$ and $A_{\bullet}^{\circ\circ\bullet} = (A_o \| A^{\circ\circ\bullet})$ are shown in Eq. (9).

$$\begin{cases} A_o^{\circ\circ\bullet} = \left\{ \begin{bmatrix} RR \\ RG \\ R\bullet \\ GR \\ G\bullet \\ BB \\ GG \\ GB \\ G\bullet \\ BR \\ B\bullet \\ BB \\ BG \\ BG \\ B\bullet \\ BR \\ B\bullet \\ BB \\ BG \end{bmatrix} \right\}, \\ A_{\bullet}^{\circ\circ\bullet} = \left\{ \begin{bmatrix} RR \\ GG \\ B\bullet \\ RB \\ GR \\ B\bullet \\ RG \\ B\bullet \\ BR \\ G\bullet \\ BB \\ RR \\ RB \\ BG \\ G\bullet \\ BR \\ R\bullet \\ BB \\ R\bullet \\ GG \end{bmatrix} \right\}. \end{cases} \quad (9)$$

This $(2, 3)$ -CBW-EVCS has $(1 \bullet 1*)$ and $(2*)$ in shadows to represent the black and white colors, i.e., $h_s = 1$ and $l_s = 0$. When stacking any two shadows, $A_{\bullet}^{\circ\circ\bullet}$ has $(2\bullet)$, while $A_o^{\circ\circ\bullet}$ only has $(1 \bullet 1*)$, i.e., $h = 2$ and $l = 1$. Finally, we have the contrasts $\alpha = \frac{1}{2}$ and $\alpha_s = \frac{1}{2}$.

Example 4. Construct a $(2, 4)$ -CBW-EVCS by [Construction 4](#).

For $n = 4$, the matrices A_{\bullet} and A_o of $(2, 4)$ -CBW-VCS with the pixel expansion $[\log_3 4] = 2$ are shown in Eq. (10).

$$A_{\bullet} = \left\{ \begin{bmatrix} RR \\ GG \\ BB \\ RG \end{bmatrix} \begin{bmatrix} RG \\ RR \\ GG \\ BB \end{bmatrix} \begin{bmatrix} BB \\ RG \\ GG \\ GG \end{bmatrix} \begin{bmatrix} GG \\ BB \\ BB \\ RR \end{bmatrix} \right\}, \quad A_o = \left\{ \begin{bmatrix} RR \\ RR \\ RR \\ RR \end{bmatrix} \begin{bmatrix} GG \\ GG \\ GG \\ GG \end{bmatrix} \begin{bmatrix} BB \\ BB \\ BB \\ BB \end{bmatrix} \begin{bmatrix} RG \\ RG \\ RG \\ RG \end{bmatrix} \right\}. \quad (10)$$

The matrices of cover image $A^{c_1 c_2 c_3 c_4}$ of [4/3] = 2 characters with 4 elements $\{(R\bullet), (G\bullet), (B\bullet), (\bullet R)\}$ are given in [Table A5](#). Consider the example $c_1 = \circ, c_2 = \circ, c_3 = \circ, c_4 = \bullet$. Matrices $A_{\bullet}^{\circ\circ\circ\bullet} = (A_{\bullet} \parallel A^{\circ\circ\circ\bullet})$ and $A_{\circ}^{\circ\circ\circ\bullet} = (A_{\circ} \parallel A^{\circ\circ\circ\bullet})$ are shown in Eq. (11).

$$\left\{ \begin{array}{l} A_{\bullet}^{\circ\circ\circ\bullet} = \left\{ \begin{array}{cccccccc} \begin{bmatrix} RRR\bullet \\ GGG\bullet \\ BBB\bullet \\ RG\bullet\bullet \end{bmatrix} & \begin{bmatrix} RR\bullet R \\ GGR\bullet \\ BBG\bullet \\ RG\bullet\bullet \end{bmatrix} & \begin{bmatrix} RRB\bullet \\ GG\bullet R \\ BBR\bullet \\ RG\bullet\bullet \end{bmatrix} & \begin{bmatrix} RRG\bullet \\ GGB\bullet \\ BB\bullet R \\ RG\bullet\bullet \end{bmatrix} & \begin{bmatrix} RGR\bullet \\ RRG\bullet \\ BB\bullet\bullet \\ RG\bullet\bullet \end{bmatrix} & \begin{bmatrix} RG\bullet R \\ RRR\bullet \\ GGG\bullet \\ BB\bullet\bullet \end{bmatrix} & \begin{bmatrix} RGB\bullet \\ RGR\bullet \\ GGR\bullet \\ BB\bullet\bullet \end{bmatrix} & \begin{bmatrix} RGG\bullet \\ RRB\bullet \\ GG\bullet R \\ BB\bullet\bullet \end{bmatrix} \\ \begin{bmatrix} BBR\bullet \\ RGG\bullet \\ RRB\bullet \\ GG\bullet\bullet \end{bmatrix} & \begin{bmatrix} BB\bullet R \\ RGR\bullet \\ RRG\bullet \\ GG\bullet\bullet \end{bmatrix} & \begin{bmatrix} BBB\bullet \\ RGG\bullet \\ RRR\bullet \\ GG\bullet\bullet \end{bmatrix} & \begin{bmatrix} BBG\bullet \\ RGB\bullet \\ RRR\bullet \\ GG\bullet\bullet \end{bmatrix} & \begin{bmatrix} GGR\bullet \\ BBG\bullet \\ RR\bullet R \\ GG\bullet\bullet \end{bmatrix} & \begin{bmatrix} GG\bullet R \\ BBR\bullet \\ RGB\bullet \\ RR\bullet\bullet \end{bmatrix} & \begin{bmatrix} GGB\bullet \\ BGR\bullet \\ RGG\bullet \\ RR\bullet\bullet \end{bmatrix} & \begin{bmatrix} GGG\bullet \\ BBB\bullet \\ RG\bullet R \\ RR\bullet\bullet \end{bmatrix} \end{array} \right\}, \\ A_{\circ}^{\circ\circ\circ\bullet} = \left\{ \begin{array}{cccccccc} \begin{bmatrix} RRR\bullet \\ RRG\bullet \\ RRB\bullet \\ RR\bullet\bullet \end{bmatrix} & \begin{bmatrix} RR\bullet R \\ RRR\bullet \\ RR\bullet R \\ RR\bullet\bullet \end{bmatrix} & \begin{bmatrix} RRB\bullet \\ RRR\bullet \\ RRG\bullet \\ RR\bullet\bullet \end{bmatrix} & \begin{bmatrix} RRG\bullet \\ RRB\bullet \\ RRR\bullet \\ RR\bullet\bullet \end{bmatrix} & \begin{bmatrix} GGR\bullet \\ GGG\bullet \\ GGB\bullet \\ GG\bullet\bullet \end{bmatrix} & \begin{bmatrix} GG\bullet R \\ GGR\bullet \\ GGG\bullet \\ GG\bullet\bullet \end{bmatrix} & \begin{bmatrix} GGB\bullet \\ GGR\bullet \\ GGG\bullet \\ GG\bullet\bullet \end{bmatrix} & \begin{bmatrix} GGG\bullet \\ GGB\bullet \\ GG\bullet R \\ GG\bullet\bullet \end{bmatrix} \\ \begin{bmatrix} BBR\bullet \\ BBG\bullet \\ BBB\bullet \\ BB\bullet\bullet \end{bmatrix} & \begin{bmatrix} BB\bullet R \\ BBR\bullet \\ BBB\bullet \\ BB\bullet\bullet \end{bmatrix} & \begin{bmatrix} BBB\bullet \\ BBR\bullet \\ BBG\bullet \\ BB\bullet\bullet \end{bmatrix} & \begin{bmatrix} BBG\bullet \\ BBR\bullet \\ BB\bullet R \\ BB\bullet\bullet \end{bmatrix} & \begin{bmatrix} RGR\bullet \\ RGG\bullet \\ RGB\bullet \\ RG\bullet\bullet \end{bmatrix} & \begin{bmatrix} RG\bullet R \\ RGR\bullet \\ RGG\bullet \\ RG\bullet\bullet \end{bmatrix} & \begin{bmatrix} RGB\bullet \\ RGR\bullet \\ RGG\bullet \\ RG\bullet\bullet \end{bmatrix} & \begin{bmatrix} RGG\bullet \\ RGB\bullet \\ RG\bullet R \\ RG\bullet\bullet \end{bmatrix} \end{array} \right\}. \end{array} \right\} \quad (11)$$

As shown in Eq. (11), in shadows, we have $(2\bullet 2\circ)$ for black color (see S_4) and $(1\bullet 3\ast)$ for white color (see S_1, S_2 , and S_3), i.e., $h_s = 2$ and $l_s = 1$. Thus, we have we have $\alpha_s = \frac{(h_s - l_s)}{4} = \frac{(2-1)}{4} = \frac{1}{4}$. When stacking any two shadows, $A_{\bullet}^{\circ\circ\circ\bullet}$ has $(4\bullet)$ or $(3\bullet 1\ast)$. However, $A_{\circ}^{\circ\circ\circ\bullet}$ only has $(2\bullet 2\circ)$, i.e., $h \geq 3$ and $l = 2$. Any two stacked shadows has $(4\bullet)$ with $\frac{5}{6}$ probability and $(3\bullet 1\ast)$ with $\frac{1}{6}$ probability. We have the average contrast $\alpha = (\frac{(4-2)}{4} \times \frac{5}{6}) + (\frac{(3-2)}{4} \times \frac{1}{6}) = \frac{11}{24}$.

Example 5. Construct a (2, 4)-PB-CBW-EVCS by [Construction 5](#).

For $n = 4$, the matrices of cover images $A^{c_1 c_2 c_3 c_4}$ are the same as those in [Table A5](#). Consider the example $c_1 = \circ, c_2 = \circ, c_3 = \circ, c_4 = \bullet$. Matrices $A_{\bullet}^{\circ\circ\circ\bullet} = (A_{\bullet} \parallel A^{\circ\circ\circ\bullet})$ and $A_{\circ}^{\circ\circ\circ\bullet} = (A_{\circ} \parallel A^{\circ\circ\circ\bullet})$ are similar to those in Eq. (11), but using $A_{\bullet} = \left\{ \begin{bmatrix} \bullet R \\ \bullet G \\ \bullet B \\ \bullet R \\ R\bullet \\ \bullet B \\ \bullet G \\ \bullet R \end{bmatrix} \right\}$ and $A_{\circ} = \left\{ \begin{bmatrix} \bullet R \\ \bullet G \\ \bullet B \\ \bullet R \\ R\bullet \\ \bullet B \\ \bullet G \\ \bullet R \end{bmatrix} \right\}$ instead. This (2, 4)-PB-CBW-EVCS has $(3\bullet 1\ast)$ and $(2\bullet 2\circ)$ in shadows to represent the black and white colors, i.e., $h_s = 3$ and $l_s = 2$. The black and white colors in reconstructed image are $(4\bullet)$ and $(3\bullet 1\ast)$, i.e., $h = 4$ (note: $h = m = 4$, this scheme is perfect black). We have the contrasts $\alpha = \frac{(4-3)}{4} = \frac{1}{4}$ and $\alpha_s = \frac{(3-2)}{4} = \frac{1}{4}$.

Example 6. Construct a (3, 3)-PB-CBW-EVCS by [Construction 6](#).

By transforming Naor and Shamir's base matrices $B_{\bullet} = \begin{bmatrix} \bullet\bullet\bullet \\ \bullet\bullet\bullet \\ \bullet\bullet\bullet \end{bmatrix}$ and $B_{\circ} = \begin{bmatrix} \bullet\bullet\bullet \\ \bullet\bullet\bullet \\ \bullet\bullet\bullet \end{bmatrix}$, we obtain the collections C_{\circ} and C_{\bullet} of (3, 3)-PB-CBW-VCS (see Table 7 and Table 8 in [\[7\]](#)). After adding 2 all-• columns into B_{\bullet} and B_{\circ} , we have $(4\bullet 2\circ)$ in each shadow. We have $n_{\max} = 1$ and $n_{\min} = 0$ for this case, and thus $d = (n_{\max} - n_{\min} + 1) = 2$. The matrix of cover image $A^{c_1 c_2 c_3}$ is the concatenation of two $A^{c_1 c_2 c_3}$ in [Table A4](#). Consider the example $c_1 = \circ, c_2 = \circ, c_3 = \bullet$, and then we have the matrix $(A^{\circ\circ\bullet} \parallel A^{\circ\circ\bullet}) = \left\{ \begin{bmatrix} R \\ G \\ R \\ \bullet \end{bmatrix} \begin{bmatrix} B \\ R \\ B \\ \bullet \end{bmatrix} \begin{bmatrix} G \\ R \\ B \\ \bullet \end{bmatrix} \right\} \parallel \left\{ \begin{bmatrix} R \\ G \\ R \\ \bullet \end{bmatrix} \begin{bmatrix} B \\ R \\ B \\ \bullet \end{bmatrix} \begin{bmatrix} G \\ R \\ B \\ \bullet \end{bmatrix} \right\} = \left\{ \begin{bmatrix} RR \\ GG \\ GR \\ \bullet \bullet \end{bmatrix} \begin{bmatrix} RB \\ GR \\ GB \\ \bullet \bullet \end{bmatrix} \begin{bmatrix} RG \\ BR \\ RG \\ \bullet \bullet \end{bmatrix} \begin{bmatrix} BB \\ RR \\ BG \\ \bullet \bullet \end{bmatrix} \begin{bmatrix} BG \\ RB \\ BG \\ \bullet \bullet \end{bmatrix} \begin{bmatrix} GR \\ BG \\ BR \\ \bullet \bullet \end{bmatrix} \begin{bmatrix} GB \\ BR \\ BB \\ \bullet \bullet \end{bmatrix} \begin{bmatrix} GG \\ BB \\ BB \\ \bullet \bullet \end{bmatrix} \right\}$.

Suppose that the secret s is \bullet , and that two matrices $A_{\bullet} = \begin{bmatrix} Y\bullet \\ Y\bullet \\ GR \\ \bullet \bullet \end{bmatrix}$ and $A_{\circ} = \begin{bmatrix} C\bullet \\ C\bullet \\ M\bullet \\ \bullet \bullet \end{bmatrix}$ are chosen (note: because $(B + Y + G) = \bullet$ and $(C + Y + M) = \bullet$ we have $(\bullet\bullet)$ in these two matrices for staking 3 shadows). The matrices $A_{\bullet}^{\circ\circ\bullet} = (A_{\bullet} \parallel A^{\circ\circ\bullet} \parallel A^{\circ\circ\bullet})$ using $A_{\bullet} = \begin{bmatrix} BR \\ Y\bullet \\ GR \\ \bullet \bullet \end{bmatrix}$ and $A_{\circ} = \begin{bmatrix} C\bullet \\ Y\bullet \\ M\bullet \\ \bullet \bullet \end{bmatrix}$ are given in Eq. (12).

$$\left\{ \begin{array}{l} A_{\bullet}^{\circ\circ\bullet} = \left\{ \begin{array}{cccccccc} \begin{bmatrix} BRRR \\ Y\bullet GG \\ GR\bullet\bullet \end{bmatrix} & \begin{bmatrix} BRRB \\ Y\bullet GR \\ GR\bullet\bullet \end{bmatrix} & \begin{bmatrix} BRRG \\ Y\bullet GB \\ GR\bullet\bullet \end{bmatrix} & \begin{bmatrix} BRBR \\ Y\bullet RG \\ GR\bullet\bullet \end{bmatrix} & \begin{bmatrix} BRBB \\ Y\bullet RR \\ GR\bullet\bullet \end{bmatrix} & \begin{bmatrix} BRBG \\ Y\bullet RB \\ GR\bullet\bullet \end{bmatrix} & \begin{bmatrix} BRGR \\ Y\bullet BG \\ GR\bullet\bullet \end{bmatrix} & \begin{bmatrix} BRGB \\ Y\bullet BR \\ GR\bullet\bullet \end{bmatrix} \\ \begin{bmatrix} C\bullet RR \\ Y\bullet GG \\ M\bullet\bullet \end{bmatrix} & \begin{bmatrix} C\bullet RB \\ Y\bullet GR \\ M\bullet\bullet \end{bmatrix} & \begin{bmatrix} C\bullet RG \\ Y\bullet GB \\ M\bullet\bullet \end{bmatrix} & \begin{bmatrix} C\bullet BR \\ Y\bullet RG \\ M\bullet\bullet \end{bmatrix} & \begin{bmatrix} C\bullet BB \\ Y\bullet RR \\ M\bullet\bullet \end{bmatrix} & \begin{bmatrix} C\bullet BG \\ Y\bullet RB \\ M\bullet\bullet \end{bmatrix} & \begin{bmatrix} C\bullet GR \\ Y\bullet BG \\ M\bullet\bullet \end{bmatrix} & \begin{bmatrix} C\bullet GB \\ Y\bullet BR \\ M\bullet\bullet \end{bmatrix} \end{array} \right\}, \\ A_{\circ}^{\circ\circ\bullet} = \left\{ \begin{array}{cccccccc} \begin{bmatrix} C\bullet RR \\ Y\bullet GG \\ M\bullet\bullet \end{bmatrix} & \begin{bmatrix} C\bullet RB \\ Y\bullet GR \\ M\bullet\bullet \end{bmatrix} & \begin{bmatrix} C\bullet RG \\ Y\bullet GB \\ M\bullet\bullet \end{bmatrix} & \begin{bmatrix} C\bullet BR \\ Y\bullet RG \\ M\bullet\bullet \end{bmatrix} & \begin{bmatrix} C\bullet BB \\ Y\bullet RR \\ M\bullet\bullet \end{bmatrix} & \begin{bmatrix} C\bullet BG \\ Y\bullet RB \\ M\bullet\bullet \end{bmatrix} & \begin{bmatrix} C\bullet GR \\ Y\bullet BG \\ M\bullet\bullet \end{bmatrix} & \begin{bmatrix} C\bullet GB \\ Y\bullet BR \\ M\bullet\bullet \end{bmatrix} \end{array} \right\}. \end{array} \right\} \quad (12)$$

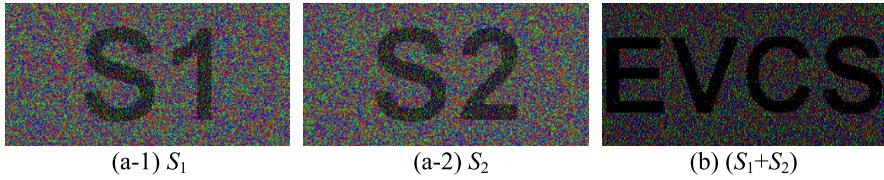


Fig. 4. (2, 2)-CBW-EVCS with {RGB}: (a) shadows, (b) the reconstructed image.

Suppose that the secret s is \circ , and that two matrices $A_\circ = \begin{bmatrix} BR \\ RR \\ GR \end{bmatrix}$ and $A_\circ = \begin{bmatrix} C\bullet \\ Y\bullet \\ GR \end{bmatrix}$ are chosen (note: because $(R + R + R) = R$ and $(C + Y + G) = G$ we have $(*)$ in these two matrices for staking 3 shadows). Then the matrices $A_\circ^{\circ\circ\bullet} = (A_\circ \parallel A^{\circ\circ\bullet} \parallel A^{\circ\circ\bullet})$ using $A_\circ = \begin{bmatrix} BR \\ RR \\ GR \end{bmatrix}$ and $\begin{bmatrix} C\bullet \\ Y\bullet \\ GR \end{bmatrix}$ are given in Eq. (13).

$$\left\{ \begin{array}{l} A_\circ^{\circ\circ\bullet} = \left\{ \begin{bmatrix} BRRR \\ RRGG \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} BRRR \\ RRGG \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} BRRG \\ RRRG \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} BRBR \\ RRRR \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} BRBB \\ RRRB \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} BRBG \\ RRBG \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} BRGR \\ RRBR \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} BRGB \\ RRBR \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} BRGG \\ RRBB \\ GR\bullet\bullet \end{bmatrix} \right\}, \\ A_\circ^{\circ\circ\bullet} = \left\{ \begin{bmatrix} C\bullet RR \\ Y\bullet GG \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} C\bullet RB \\ Y\bullet GR \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} C\bullet RG \\ Y\bullet GB \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} C\bullet BR \\ Y\bullet RG \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} C\bullet BB \\ Y\bullet RR \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} C\bullet BG \\ Y\bullet RB \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} C\bullet GR \\ Y\bullet BG \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} C\bullet GB \\ Y\bullet BR \\ GR\bullet\bullet \end{bmatrix} \begin{bmatrix} C\bullet GG \\ Y\bullet BB \\ GR\bullet\bullet \end{bmatrix} \right\}. \end{array} \right. \quad (13)$$

In the reconstructed image, we have $(4\bullet)$ and $(3\bullet 1\ast)$ for black and white colors. Thus, the contrast is $\alpha = \frac{(4-3)}{4} = \frac{1}{4}$. The contrast of shadow is calculated as follows. From Eqs. (12) and (13), for S_3 , we have $(3\bullet 1\ast)$ or $(2\bullet 2\ast)$ for black color in shadows. Additionally, we have $(1\bullet 3\ast)$ or $(4\ast)$ for white color in shadows (see S_1 and S_2). From all the distribution matrices in (3, 3)-CBW-VCS (see Table 7 and Table 8 in [7]), we have $h_s = 2$ and 3 with probability 1/2 for each, an $l_s = 0$ and 1 with probability 1/2 for each. Thus, we have the average $h_s = \frac{(3+2)}{2} = 2.5$ and the average $l_s = \frac{(1+0)}{2} = \frac{1}{2}$. Then, the average contrast is $\alpha_s = \frac{(2.5-0.5)}{4} = \frac{1}{2}$. This average contrast can be represented as a general term derived as follows. Because the maximum number and minimum number of \bullet for each row in this (k, n) -PB-CBW-VCS is n_{\max} and n_{\min} , we have the

following average blackness of h_s and l_s when concatenating the matrix $(\overbrace{A^{c_1 c_2 \dots c_n} \parallel \dots \parallel A^{c_1 c_2 \dots c_n}}^d)$.

$$\left\{ \begin{array}{l} h_s = \frac{(\lceil n/3 \rceil + n_{\min}) + (\lceil n/3 \rceil + n_{\min} + 1) + \dots + (\lceil n/3 \rceil + n_{\max})}{(n_{\max} - n_{\min} + 1)} = \lceil n/3 \rceil + \frac{(n_{\min} + n_{\max})}{2} \\ l_s = \frac{(\lceil n/3 \rceil - d + n_{\min}) + (\lceil n/3 \rceil - d + n_{\min} + 1) + \dots + (\lceil n/3 \rceil - d + n_{\max})}{(n_{\max} - n_{\min} + 1)} = (\lceil n/3 \rceil - d) + \frac{(n_{\min} + n_{\max})}{2} \end{array} \right. \quad (14)$$

We then have the average contrast $\alpha_s = \frac{(h_s - l_s)}{m} = ((\lceil n/3 \rceil + \frac{(n_{\min} + n_{\max})}{2}) - (\lceil n/3 \rceil - d + \frac{(n_{\min} + n_{\max})}{2})) / m = \frac{d}{m}$. By this general term, we have the average contrast of (3, 3)-PB-CBW-VCS $\alpha_s = \frac{d}{m} = \frac{2}{4} = \frac{1}{2}$.

4. Experiments and comparisons

4.1. Experimental results

Four experiments are conducted to evaluate the performance of the proposed CBW-EVCSs. There are 7 schemes (denoted as Schemes A–G) in these four experiments. **Experiment 1** tests the clearness of shadow and reconstructed image among three (2, 2)-CWB-EVCSs: (2, 2)-CWB-EVCS with {RGB} from **Construction 1** (Scheme A), (2, 2)-CWB-EVCS with {RGBCMY} from **Construction 2** (Scheme B), and (2, 2)-CWB-EVCS with $\{b_1 b_2\}$ background colors from **Construction 3** (Scheme C). **Experiment 1** only demonstrates the simple 2-out-of-2 schemes. In **Experiment 2**, we show the (2, 3)-CWB-EVCS with {RGB} from **Construction 4** (Scheme D). **Experiment 3** shows the (2, 4)-CWB-EVCS with {RGB} from **Construction 4** (Scheme E) and the (2, 4)-PB-CWB-EVCS with {RGB} from **Construction 5** (Scheme F). Both Scheme E and Scheme F are (2, 4) schemes, but Scheme F has the perfect reconstruction of black pixels. In **Experiment 4**, we give the proposed (3, 3)-PB-CBW-EVCS from **Construction 6** (Scheme G), which is an example with $k > 2$.

Experiment 1. Consider Scheme A, Scheme B and Scheme C in **Example 2**. All tested images are printed-text images. The secret is **EVCS**, and the cover images on shadows are **S1** and **S2**.

All the three schemes have $m = 1$. **Fig. 4** and **Fig. 5** illustrate 2 shadows and the reconstructed images of Scheme A and Scheme B, respectively. It is observed that Scheme B has $\lambda = 50\%$, while Scheme A only has $\lambda = 33\%$. **Fig. 6** shows the stacked results of Scheme C with {RG}, {RB}, {GB}, {CR}, {MG}, and {YB}. The average light transmission of **Figs. 6(a–c)** are $\lambda = 33\%$, which are the same as that of **Fig. 4**, while its background color only has a mixture of 2 colors. **Figs. 6(d–f)**

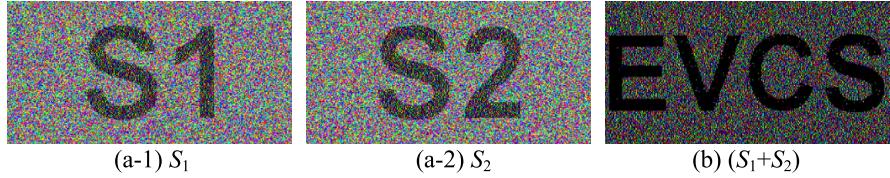


Fig. 5. (2, 2)-CBW-EVCS with {RGBCMY}: (a) shadows (b) the reconstructed image.

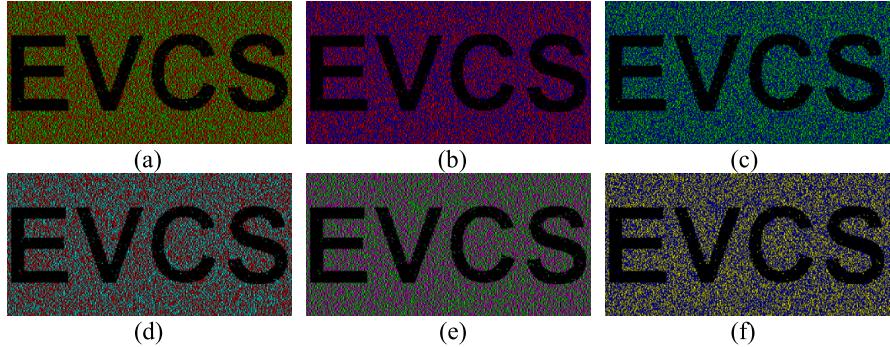


Fig. 6. The stacked results of (2, 2)-CBW-EVCS with $\{b_1b_2\}$: (a) {RG}, (b) {RB}, (c) {GB}, (d) {CR}, (e) {MG}, (f) {YB}.

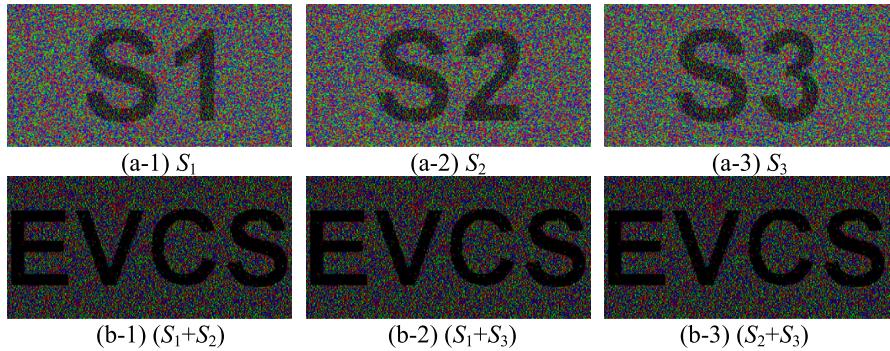


Fig. 7. (2, 3)-CBW-EVCS with {RGB}: (a) shadows, (b) the reconstructed images.

have $\lambda = (\frac{1}{3} + \frac{2}{3})/2 = 50\%$. All the contrasts of Schemes A, B, and C are $\alpha_s = \frac{1}{2}$ and $\alpha = \frac{1}{2}$. So, we can see **S1** and **S2** on shadows, and visually decode the secret image **EVCS**.

Experiment 2. Consider Scheme D in [Example 3](#). The secret is **EVCS**, and the cover images on shadows are **S1**, **S2**, and **S3**.

From [Example 3](#), Scheme D has $m = 2$, $\alpha_s = \frac{1}{2}$ and $\alpha = \frac{1}{2}$. Three shadows (S_1 , S_1 and S_3) with cover images **S1**, **S2**, and **S3** are shown in [Fig. 7\(a\)](#). The reconstructed images **EVCS** for $(S_1 + S_2)$, $(S_1 + S_3)$, and $(S_2 + S_3)$ are illustrated in [Fig. 7\(b\)](#). Since the background color is a mixture of {RGB}, thus $\lambda = 33\%$.

Experiment 3. Consider Scheme E in [Example 4](#) and Scheme F in [Example 5](#). The secret is **EVCS**, and the cover images on shadows are **S1**, **S2**, **S3**, and **S4**.

Both schemes are (2, 4) schemes and have the pixel expansion $m = 4$. Actually, by using $m = 4$, the values of n for (2, n) schemes from [Construction 4](#) and [Construction 5](#) can be, respectively, at most up to 9 and 6. Scheme E has $\alpha_s = \frac{1}{4}$ and the average contrast $\alpha = \frac{11}{24}$, while Scheme F has $\alpha_s = \frac{1}{4}$ and $\alpha = \frac{1}{4}$. The blackness h of Scheme E is 3 or 4, but the blackness of Scheme F is $h = m = 4$. Thus, Scheme F is a perfect black scheme. [Figs. 8](#) and [9](#) show 4 shadows and two reconstructed images $(S_1 + S_2)$ and $(S_3 + S_4)$ for Scheme E and Scheme F, respectively. It is observed [Fig. 9\(b\)](#) has the perfect blackness.

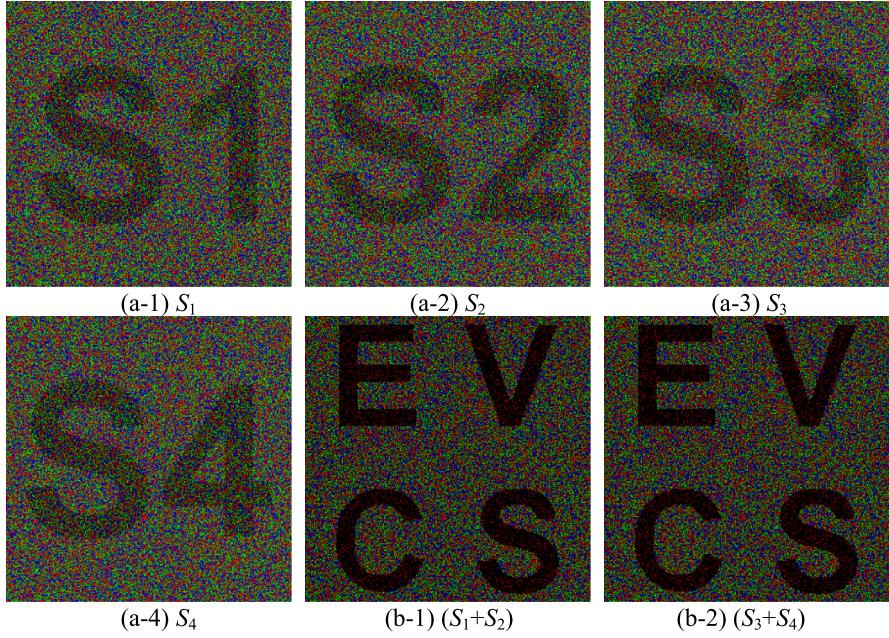


Fig. 8. (2, 4)-CBW-EVCS with {RGB}: (a) shadows, (b) the reconstructed images.

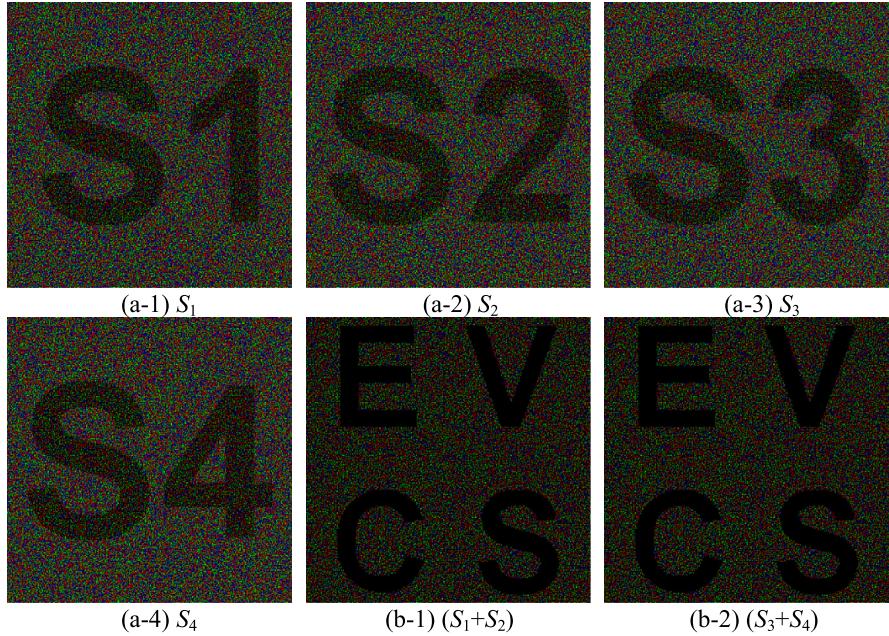


Fig. 9. (2, 4)-PB-CBW-EVCS with {RGB}: (a) shadows, (b) the reconstructed images.

Experiment 4. Consider Scheme G in [Example 6](#). The secret is $\begin{bmatrix} EV \\ CS \end{bmatrix}$, and the cover images on shadows are $\boxed{S1}$, $\boxed{S2}$, and $\boxed{S3}$.

Scheme G is a scheme with $k > 2$. From [Example 6](#), we have $m = 4$, $\alpha_s = \frac{1}{2}$, and $\alpha = \frac{1}{4}$. Additionally, the blackness is $h = m = 4$, and thus Scheme G has the perfect blackness. [Fig. 10](#) shows 3 shadows, the stacked results of any 2 stacked shadows, and the reconstructed image of stacking all 3 shadows. Since the threshold is 3, so we cannot see anything from any 2 stacked shadows. In [Fig. 10\(c\)](#), we can visually decode the secret when stacking 3 shadows.

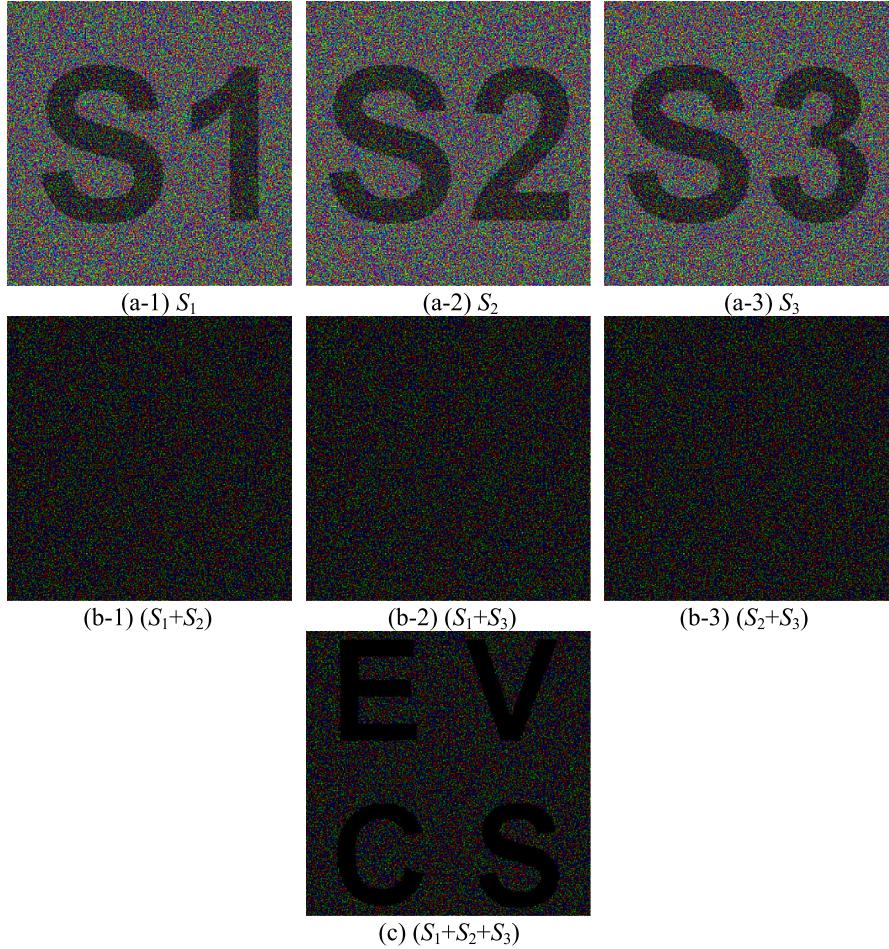


Fig. 10. (3, 3)-PB-CBW-EVCS with {RGB}: (a) shadows, (b) the results of stacking any 2 shadows, (c) the reconstructed image of stacking all 3 shadows.

4.2. Comparison

Generally, the CBW-VCS has the less pixel expansion when compared with the conventional VCS. Even our pixel expansion is not larger than the conventional scheme. The reason is that CBW-VCS and our CBW-EVCS use the color pixels, while the conventional VCS uses the black and white pixels.

In comparing with the black and white (k, n) -EVCS, our (k, n) -CBW-EVCS has the less pixel expansion. For example, Naor and Shamir's $(2, 2)$ -EVCS has $m = 4$, where using $(\bullet\bullet\bullet\circ)/(\bullet\bullet\circ\circ)$ to represent black/white colors in shadows and $(\bullet\bullet\bullet\bullet)/(\bullet\bullet\bullet\circ)$ to represent black/white colors in the reconstructed image. The proposed $(2, 2)$ -CBW-EVCS only has $m = 2$. Although the proposed (k, n) -CBW-EVCS has the larger pixel expansion than De Prisco and De Santis's (k, n) -CBW-VCS, our scheme has the extended capability revealing cover images on shadows. The pixel expansions of (k, n) -VCS, (k, n) -CBW-VCS, and our (k, n) -CBW-EVCS, where $2 \leq k \leq n \leq 9$, are illustrated in Table 3. Consider the example $(3, 5)$ scheme. For $(3, 5)$ -CBW-VCS, we should add 2 all-• columns into a $(3, 5)$ -PB-VCS with $m_B = 16$ to get 18 pixels (a multiple of 3) and we can implement a $(3, 5)$ -CBW-VCS with $m = 6$. There are $(14 \bullet 4\circ)$ in these 18 pixels, and thus $d = (n_{\max} - n_{\min}) + 1 = (4 - 2) + 1 = 3$ (note: Table 4 shows the values of d for the schemes in Table 3). So, we append $\lceil n/3 \rceil \times d = \lceil 5/3 \rceil \times 3 = 6$ colored pixels to construct our $(3, 5)$ -CBW-EVCS with $m = 6 + 6 = 12$.

Contrast comparison for some small values of k and n is shown in Table 5. All these schemes are perfect black schemes. It is observed that our scheme at least has the same contrast of the reconstructed image when compared with VCS. However, shadows of VCS and CBW-VCS are noise-like, but our CBW-EVCS has the extended capability. In Table 5, our $(2, n)$ -PB-CBW-EVCS has $\alpha = \frac{1}{m}$ (see Construction 5). About the contrasts of $(3, n)$ -PB-CBW-EVCS, and (n, n) -PB-CBW-EVCS, since they are transformed from the conventional (k, n) -PB-VCS with the pixel expansion m_B and the blackness $h_B = m_B$ and $l_B = (m_B - 1)$. After adding $(\lceil m_B/3 \rceil \times 3) - m_B$ all-• columns, this (k, n) -PB-VCS has new $h_B = (\lceil m_B/3 \rceil \times 3)$ and $l_B = (\lceil m_B/3 \rceil \times 3) - 1$. By grouping every 3 black/white pixels to $\lceil m_B/3 \rceil$ color pixels, we have $w_\bullet(\text{add}(A_\bullet | X_k)) = \lceil m_B/3 \rceil$ (since $h_B = (\lceil m_B/3 \rceil \times 3)$) and $w_\bullet(\text{add}(A_\circ | X_k)) = (\lceil m_c/3 \rceil - 1)$ (since $l_B = (\lceil m_B/3 \rceil \times 3) - 1$). Therefore, the contrasts of both $(3, n)$ -PB-CBW-EVCS

Table 3

Comparison of pixel expansion.

(2, n)				(3, n)†				(n – 1, n)†				(n, n)†			
n	#1	#2	#3	n	#1	#2	#3	n	#1	#2	#3	n	#1	#2	#3
2	2	1 (1*)	2 (2*)	2	–	–	–	2	–	–	–	2	2	1	2
3	3	1 (1*)	2 (2*)	3	4	2	4	3	3	1	2	3	4	2	4
4	4	2 (2*)	4 (4*)	4	9	3	9	4	9	3	9	4	8	3	7
5	5	2 (2*)	4 (4*)	5	16	6	12	5	25	9	23	5	16	6	14
6	6	2 (2*)	4 (4*)	6	25	9	17	6	65	22	50	6	32	12	24
7	7	2 (3*)	5 (6*)	7	36	12	27	7	161	54	153	7	64	22	58
8	8	2 (3*)	5 (6*)	8	49	17	32	8	385	129	357	8	128	43	109
9	9	2 (3*)	5 (6*)	9	64	22	40	9	897	299	812	9	256	86	218

#1: (k, n)-VCS; #2: (k, n)-CBW-VCS; #3: (k, n)-CBW-EVCS; *: Perfect Black.

Table 4

The values of d for constructing (k, n)-PB-CBW-EVCS.

(3, n)			(n – 1, n)			(n, n)		
n	(x • y○)	d	n	(x • y○)	d	n	(x • y○)	d
2	–	–	2	–	–	2	(2 • 1○)	1
3	(4 • 2○)	2	3	(2 • 1○)	1	3	(4 • 2○)	2
4	(6 • 3○)	3	4	(6 • 3○)	3	4	(5 • 4○)	2
5	(14 • 4○)	3	5	(18 • 9○)	7	5	(10 • 8○)	4
6	(22 • 5○)	4	6	(41 • 25○)	14	6	(17 • 16○)	6
7	(33 • 6○)	5	7	(97 • 65○)	33	7	(34 • 32○)	12
8	(44 • 7○)	5	8	(226 • 161○)	76	8	(65 • 64○)	22
9	(58 • 8○)	6	9	(512 • 385○)	171	9	(130 • 128○)	44

Table 5

Comparison of contrast for (k, n) schemes for some small values of k and n.

(k, n)		The reconstructed image (α)			Shadow (α_s)		
		#1	#2	#3	#1, #2	#3	
k = 2	n = 2	1/2	1	1/2	0	1/2	
	n = 3	1/3	1	1/2	0	1/2	
	n = 4	1/4	1/2	1/4	0	1/4	
	n = 5	1/5	1/2	1/4	0	1/4	
	n = 6	1/6	1/2	1/4	0	1/4	
k = 3	n = 3	1/4	1/2	1/4	0	1/2	
	n = 4	1/9	1/3	1/9	0	1/3	
	n = 5	1/16	1/6	1/12	0	1/4	
k = 4	n = 4	1/8	1/3	1/7	0	2/7	

#1: (k, n)-PB-VCS (note: for k = 2 we use Naor and Shamir's reversed (2, n)-PB-VCS).

#2: De Prisco and De Santis's (2, n)-PB-CBW-VCS.

#3: the proposed (k, n)-PB-CBW-EVCS.

and (n, n)-PB-CBW-EVCS are all $\alpha = \frac{1}{m}$. For example, in Table 5, our (3, n)-PB-CBW-EVCS has $\alpha = \frac{1}{4}, \frac{1}{9}$ and $\frac{1}{12}$ for $n = 3, 4$ and 5, and the (4, 4)-PB-CBW-EVCS has $\alpha = \frac{1}{7}$.

5. Conclusion

In this paper, we take the lead to study CBW-EVCS. Our paper presented (2, 2)-CBW-EVCS, (2, n)-CBW-EVCS, (2, n)-PB-CBW-EVCS, and (k, n)-PB-CBW-EVCS. Via the idea of Ateniese et al.'s EVCS, we give the conditions (security, contrast, and cover image conditions) of (k, n)-CWB-EVCS, and prove that the proposed CBW-EVCSs satisfy these conditions. Our scheme at least has the same pixel expansion and contrast when compared with (k, n)-VCS. Meanwhile, our scheme provides the extended capability. For arbitrary k and n, De Prisco and De Santis's (k, n)-PB-CBW-VCS and our (k, n)-PB-CBW-EVCS are all perfect black. How to construct the non-perfect black (k, n)-CBW-VCS and (k, n)-CBW-EVCS is interesting and deserve further studying.

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Appendix A

Table A1

The collections of $C_o^{t_1 t_2}$ and $C_{\bullet}^{t_1 t_2}$ in Construction 1.

$$C_{\circ}^{\circ\circ} = \left\{ \begin{array}{cccccccccccccccccccccc} RR & RR & RG & RG & RB & RB & RR & RR & GR & GR & BR & BR & BR & GR & GR & GG & GG & GB & GB \\ RG & RB & RR & RB & RR & RG & GR & BR & RR & BR & RR & GR & GG & GB & GR & GB & GR & GR & GG \\ RG & RG & GG & GG & BG & BG & BR & BR & BG & BG & BB & BB & BB & RB & RB & GB & GB & BB & BB \\ GG & BG & RG & BG & RG & GG & BG & BB & BR & BB & BR & BG & GB & BB & RB & BB & RB & BB & GB \end{array} \right\}$$

$$C_{\circ}^{\bullet\bullet} = \left\{ \begin{bmatrix} RR \\ R\bullet \end{bmatrix}, \begin{bmatrix} RG \\ R\bullet \end{bmatrix}, \begin{bmatrix} RB \\ R\bullet \end{bmatrix}, \begin{bmatrix} RR \\ \bullet R \end{bmatrix}, \begin{bmatrix} GR \\ \bullet R \end{bmatrix}, \begin{bmatrix} BR \\ \bullet R \end{bmatrix}, \begin{bmatrix} GR \\ G\bullet \end{bmatrix}, \begin{bmatrix} GG \\ G\bullet \end{bmatrix}, \begin{bmatrix} GB \\ G\bullet \end{bmatrix}, \begin{bmatrix} RG \\ \bullet G \end{bmatrix}, \begin{bmatrix} GG \\ \bullet G \end{bmatrix}, \begin{bmatrix} BG \\ \bullet G \end{bmatrix}, \begin{bmatrix} BR \\ B\bullet \end{bmatrix}, \begin{bmatrix} BG \\ B\bullet \end{bmatrix}, \begin{bmatrix} BB \\ B\bullet \end{bmatrix}, \begin{bmatrix} RB \\ B\bullet \end{bmatrix}, \begin{bmatrix} GB \\ \bullet B \end{bmatrix}, \begin{bmatrix} BB \\ \bullet B \end{bmatrix} \right\}$$

$$C_{\circ}^{\circ} = \left\{ \begin{bmatrix} R\bullet \\ RR \end{bmatrix}, \begin{bmatrix} R\bullet \\ RG \end{bmatrix}, \begin{bmatrix} R\bullet \\ RB \end{bmatrix}, \begin{bmatrix} \bullet R \\ RR \end{bmatrix}, \begin{bmatrix} \bullet R \\ GR \end{bmatrix}, \begin{bmatrix} \bullet R \\ BR \end{bmatrix}, \begin{bmatrix} G\bullet \\ GR \end{bmatrix}, \begin{bmatrix} G\bullet \\ GG \end{bmatrix}, \begin{bmatrix} G\bullet \\ GB \end{bmatrix}, \begin{bmatrix} \bullet G \\ RG \end{bmatrix}, \begin{bmatrix} \bullet G \\ GG \end{bmatrix}, \begin{bmatrix} \bullet G \\ BG \end{bmatrix}, \begin{bmatrix} B\bullet \\ BR \end{bmatrix}, \begin{bmatrix} B\bullet \\ BG \end{bmatrix}, \begin{bmatrix} B\bullet \\ BB \end{bmatrix}, \begin{bmatrix} \bullet B \\ BB \end{bmatrix}, \begin{bmatrix} \bullet B \\ RB \end{bmatrix}, \begin{bmatrix} \bullet B \\ GB \end{bmatrix}, \begin{bmatrix} \bullet B \\ BB \end{bmatrix} \right\}$$

$$C_{\circ}^{\bullet\bullet} = \left\{ \begin{bmatrix} R\bullet \\ R\bullet \end{bmatrix} \begin{bmatrix} G\bullet \\ G\bullet \end{bmatrix} \begin{bmatrix} B\bullet \\ B\bullet \end{bmatrix} \begin{bmatrix} \bullet R \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet G \\ \bullet G \end{bmatrix} \begin{bmatrix} \bullet B \\ \bullet B \end{bmatrix} \right\}$$

$$C_{\bullet}^{\circ\circ} = \left\{ \begin{bmatrix} RR \\ GG \\ GR \\ BG \end{bmatrix}, \begin{bmatrix} RR \\ GB \\ GR \\ BB \end{bmatrix}, \begin{bmatrix} RG \\ GR \\ GG \\ BR \end{bmatrix}, \begin{bmatrix} RG \\ GB \\ GG \\ BR \end{bmatrix}, \begin{bmatrix} RB \\ GR \\ BB \\ BG \end{bmatrix}, \begin{bmatrix} RB \\ GG \\ BB \\ BG \end{bmatrix}, \begin{bmatrix} RB \\ BG \\ BB \\ BG \end{bmatrix}, \begin{bmatrix} RR \\ BB \\ BR \\ BG \end{bmatrix}, \begin{bmatrix} RR \\ BR \\ BB \\ BG \end{bmatrix}, \begin{bmatrix} RG \\ BB \\ BB \\ BG \end{bmatrix}, \begin{bmatrix} RG \\ BG \\ BB \\ BB \end{bmatrix}, \begin{bmatrix} RB \\ BG \\ BB \\ BR \end{bmatrix}, \begin{bmatrix} RB \\ RG \\ BB \\ BR \end{bmatrix}, \begin{bmatrix} GR \\ RG \\ BR \\ BR \end{bmatrix}, \begin{bmatrix} GR \\ RB \\ RR \\ BG \end{bmatrix}, \begin{bmatrix} GG \\ RB \\ RR \\ BG \end{bmatrix}, \begin{bmatrix} GG \\ RB \\ BG \\ BB \end{bmatrix}, \begin{bmatrix} GB \\ RR \\ BB \\ BB \end{bmatrix}, \begin{bmatrix} GB \\ RB \\ BB \\ BB \end{bmatrix}, \begin{bmatrix} GB \\ RG \\ BB \\ GG \end{bmatrix} \right\}$$

$$C_{\bullet}^{\infty} = \left\{ \begin{bmatrix} RR \\ G\bullet \\ RG \\ GR \\ RG \\ R\bullet \end{bmatrix}, \begin{bmatrix} RG \\ G\bullet \\ G\bullet \\ BG \\ GG \\ \bullet R \end{bmatrix}, \begin{bmatrix} RB \\ B\bullet \\ B\bullet \\ B\bullet \\ BR \\ \bullet R \end{bmatrix}, \begin{bmatrix} RR \\ B\bullet \\ BG \\ RG \\ \bullet B \\ \bullet B \end{bmatrix}, \begin{bmatrix} RG \\ BG \\ BB \\ BR \\ R\bullet \\ R\bullet \end{bmatrix}, \begin{bmatrix} BR \\ \bullet G \\ BB \\ BG \\ R\bullet \\ G\bullet \end{bmatrix}, \begin{bmatrix} RR \\ \bullet G \\ BR \\ \bullet B \\ BG \\ G\bullet \end{bmatrix}, \begin{bmatrix} GR \\ \bullet B \\ \bullet B \\ BB \\ G\bullet \\ G\bullet \end{bmatrix}, \begin{bmatrix} BR \\ R\bullet \\ \bullet B \\ GB \\ \bullet R \\ \bullet R \end{bmatrix}, \begin{bmatrix} GR \\ R\bullet \\ R\bullet \\ GB \\ R\bullet \\ \bullet R \end{bmatrix}, \begin{bmatrix} GG \\ R\bullet \\ GB \\ BB \\ R\bullet \\ \bullet G \end{bmatrix}, \begin{bmatrix} GB \\ R\bullet \\ GB \\ BB \\ RB \\ \bullet G \end{bmatrix}, \begin{bmatrix} GR \\ B\bullet \\ B\bullet \\ BB \\ GB \\ \bullet G \end{bmatrix}, \begin{bmatrix} GG \\ B\bullet \\ BB \\ GB \\ GB \\ \bullet G \end{bmatrix}, \begin{bmatrix} GB \\ B\bullet \\ BB \\ BB \\ RB \\ \bullet G \end{bmatrix} \right\}$$

$$C_{\bullet}^{\circ\circ} = \left\{ \begin{bmatrix} R\bullet \\ GR \\ \bullet G \\ RR \end{bmatrix}, \begin{bmatrix} R\bullet \\ GG \\ \bullet G \\ GR \end{bmatrix}, \begin{bmatrix} R\bullet \\ GB \\ \bullet G \\ BR \end{bmatrix}, \begin{bmatrix} R\bullet \\ BR \\ \bullet G \\ RB \end{bmatrix}, \begin{bmatrix} R\bullet \\ BG \\ \bullet G \\ GB \end{bmatrix}, \begin{bmatrix} R\bullet \\ BB \\ \bullet G \\ BB \end{bmatrix}, \begin{bmatrix} \bullet R \\ RG \\ B\bullet \\ RR \end{bmatrix}, \begin{bmatrix} \bullet R \\ GG \\ B\bullet \\ RG \end{bmatrix}, \begin{bmatrix} \bullet R \\ BG \\ B\bullet \\ RB \end{bmatrix}, \begin{bmatrix} \bullet R \\ RB \\ B\bullet \\ GR \end{bmatrix}, \begin{bmatrix} \bullet R \\ GB \\ B\bullet \\ GG \end{bmatrix}, \begin{bmatrix} \bullet R \\ BB \\ B\bullet \\ GB \end{bmatrix}, \begin{bmatrix} G\bullet \\ RR \\ \bullet B \\ RR \end{bmatrix}, \begin{bmatrix} G\bullet \\ RG \\ \bullet B \\ GR \end{bmatrix}, \begin{bmatrix} G\bullet \\ RB \\ \bullet B \\ BR \end{bmatrix}, \begin{bmatrix} G\bullet \\ BR \\ \bullet B \\ RG \end{bmatrix}, \begin{bmatrix} G\bullet \\ BG \\ \bullet B \\ GG \end{bmatrix}, \begin{bmatrix} G\bullet \\ BB \\ \bullet B \\ BG \end{bmatrix} \right\}$$

$$C_{\bullet}^{\bullet} = \left\{ \begin{bmatrix} R_0 \\ G_0 \end{bmatrix}, \begin{bmatrix} R_0 \\ B_0 \end{bmatrix}, \begin{bmatrix} \bullet R \\ \bullet G \end{bmatrix}, \begin{bmatrix} \bullet R \\ \bullet B \end{bmatrix}, \begin{bmatrix} G_0 \\ R_0 \end{bmatrix}, \begin{bmatrix} G_0 \\ B_0 \end{bmatrix}, \begin{bmatrix} \bullet G \\ \bullet R \end{bmatrix}, \begin{bmatrix} \bullet G \\ \bullet B \end{bmatrix}, \begin{bmatrix} B_0 \\ R_0 \end{bmatrix}, \begin{bmatrix} B_0 \\ G_0 \end{bmatrix}, \begin{bmatrix} \bullet B \\ \bullet R \end{bmatrix}, \begin{bmatrix} \bullet B \\ \bullet G \end{bmatrix} \right\}$$

Table A2

The collections of $C_{\circ}^{c_1 c_2}$ and $C_{\bullet}^{c_1 c_2}$ in Construction 2.

$$C_{\circ}^{\circ} = \left\{ \begin{bmatrix} RR \\ RC \end{bmatrix}, \begin{bmatrix} RG \\ RM \end{bmatrix}, \begin{bmatrix} RB \\ RY \end{bmatrix}, \begin{bmatrix} RC \\ RR \end{bmatrix}, \begin{bmatrix} RM \\ RG \end{bmatrix}, \begin{bmatrix} RY \\ RB \end{bmatrix}, \begin{bmatrix} RR \\ CR \end{bmatrix}, \begin{bmatrix} GR \\ MR \end{bmatrix}, \begin{bmatrix} BR \\ YR \end{bmatrix}, \begin{bmatrix} CR \\ RR \end{bmatrix}, \begin{bmatrix} MR \\ GR \end{bmatrix}, \begin{bmatrix} YR \\ BR \end{bmatrix}, \begin{bmatrix} GR \\ GC \end{bmatrix}, \begin{bmatrix} GG \\ GM \end{bmatrix}, \begin{bmatrix} GB \\ GY \end{bmatrix}, \begin{bmatrix} GC \\ GR \end{bmatrix}, \begin{bmatrix} GM \\ GG \end{bmatrix}, \begin{bmatrix} GY \\ GB \end{bmatrix}, \right. \\ \left. \begin{bmatrix} RG \\ GG \end{bmatrix}, \begin{bmatrix} CG \\ MG \end{bmatrix}, \begin{bmatrix} BG \\ YG \end{bmatrix}, \begin{bmatrix} CG \\ RG \end{bmatrix}, \begin{bmatrix} MG \\ GG \end{bmatrix}, \begin{bmatrix} YG \\ BG \end{bmatrix}, \begin{bmatrix} BR \\ BC \end{bmatrix}, \begin{bmatrix} BG \\ BM \end{bmatrix}, \begin{bmatrix} BB \\ BY \end{bmatrix}, \begin{bmatrix} BC \\ BR \end{bmatrix}, \begin{bmatrix} BM \\ BG \end{bmatrix}, \begin{bmatrix} BY \\ BB \end{bmatrix}, \begin{bmatrix} RB \\ CB \end{bmatrix}, \begin{bmatrix} GB \\ CB \end{bmatrix}, \begin{bmatrix} BB \\ MB \end{bmatrix}, \begin{bmatrix} CB \\ YB \end{bmatrix}, \begin{bmatrix} MB \\ RB \end{bmatrix}, \begin{bmatrix} YB \\ BB \end{bmatrix}, \right. \\ \left. \begin{bmatrix} CR \\ CC \end{bmatrix}, \begin{bmatrix} CG \\ CM \end{bmatrix}, \begin{bmatrix} CB \\ CY \end{bmatrix}, \begin{bmatrix} CC \\ CR \end{bmatrix}, \begin{bmatrix} CM \\ CG \end{bmatrix}, \begin{bmatrix} CY \\ CB \end{bmatrix}, \begin{bmatrix} RC \\ CC \end{bmatrix}, \begin{bmatrix} GC \\ MC \end{bmatrix}, \begin{bmatrix} BC \\ YC \end{bmatrix}, \begin{bmatrix} CC \\ RC \end{bmatrix}, \begin{bmatrix} MC \\ GC \end{bmatrix}, \begin{bmatrix} YC \\ BC \end{bmatrix}, \begin{bmatrix} MR \\ MC \end{bmatrix}, \begin{bmatrix} MG \\ MM \end{bmatrix}, \begin{bmatrix} MB \\ MY \end{bmatrix}, \begin{bmatrix} MC \\ MR \end{bmatrix}, \begin{bmatrix} MM \\ MG \end{bmatrix}, \begin{bmatrix} MM \\ MG \end{bmatrix}, \begin{bmatrix} MY \\ MB \end{bmatrix}, \right. \\ \left. \begin{bmatrix} RM \\ CM \end{bmatrix}, \begin{bmatrix} GM \\ MM \end{bmatrix}, \begin{bmatrix} BM \\ YM \end{bmatrix}, \begin{bmatrix} CM \\ RM \end{bmatrix}, \begin{bmatrix} MM \\ GM \end{bmatrix}, \begin{bmatrix} YM \\ BM \end{bmatrix}, \begin{bmatrix} YR \\ YC \end{bmatrix}, \begin{bmatrix} YG \\ YM \end{bmatrix}, \begin{bmatrix} YB \\ YY \end{bmatrix}, \begin{bmatrix} YC \\ YR \end{bmatrix}, \begin{bmatrix} YM \\ YG \end{bmatrix}, \begin{bmatrix} YY \\ YB \end{bmatrix}, \begin{bmatrix} RY \\ CY \end{bmatrix}, \begin{bmatrix} GY \\ MY \end{bmatrix}, \begin{bmatrix} BY \\ YY \end{bmatrix}, \begin{bmatrix} CY \\ RY \end{bmatrix}, \begin{bmatrix} MY \\ GY \end{bmatrix}, \begin{bmatrix} YY \\ BY \end{bmatrix} \right\}$$

Table A2 (*continued*)

$$C_{\circ}^{\bullet\bullet} = \left\{ \begin{bmatrix} R\bullet \\ R\bullet \end{bmatrix} \begin{bmatrix} G\bullet \\ G\bullet \end{bmatrix} \begin{bmatrix} B\bullet \\ B\bullet \end{bmatrix} \begin{bmatrix} C\bullet \\ C\bullet \end{bmatrix} \begin{bmatrix} M\bullet \\ M\bullet \end{bmatrix} \begin{bmatrix} Y\bullet \\ Y\bullet \end{bmatrix} \begin{bmatrix} \bullet R \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet G \\ \bullet G \end{bmatrix} \begin{bmatrix} \bullet B \\ \bullet B \end{bmatrix} \begin{bmatrix} \bullet C \\ \bullet C \end{bmatrix} \begin{bmatrix} \bullet M \\ \bullet M \end{bmatrix} \begin{bmatrix} \bullet Y \\ \bullet Y \end{bmatrix} \right\}$$

$$C^{\circ\circ} = \left\{ \begin{array}{cccccccccccccccccccccccc} [RR] & [RG] & [RB] & [RC] & [RM] & [RY] & [RR] & [GR] & [BR] & [CR] & [MR] & [YR] & [GR] & [GG] & [GB] & [GC] & [GM] & [GY] \\ [CC] & [CM] & [CY] & [CR] & [CG] & [CB] & [CC] & [MC] & [YC] & [RC] & [GC] & [BC] & [MC] & [MM] & [MY] & [MR] & [MG] & [MB] \\ [CM] & [GG] & [BG] & [CG] & [MG] & [YG] & [BC] & [BG] & [YB] & [YY] & [BM] & [BY] & [RB] & [GB] & [BB] & [CB] & [GY] & [YB] \\ [MM] & [YM] & [YM] & [RM] & [GM] & [BM] & [YC] & [YM] & [YY] & [YR] & [YG] & [YB] & [CY] & [MY] & [YY] & [RY] & [CB] & [BY] \\ [CR] & [CG] & [CB] & [CC] & [CM] & [CY] & [RC] & [GC] & [BC] & [CC] & [MC] & [YC] & [MR] & [MG] & [MB] & [MC] & [MM] & [MY] \\ [RC] & [RM] & [RY] & [RR] & [RG] & [RB] & [CR] & [MR] & [YR] & [RR] & [GR] & [BR] & [GC] & [GM] & [GY] & [GR] & [GG] & [GB] \\ [RM] & [GM] & [BM] & [CM] & [MM] & [YM] & [YR] & [YG] & [YB] & [YC] & [YM] & [YY] & [RY] & [GY] & [BY] & [CY] & [MY] & [YY] \\ [CG] & [MG] & [YG] & [RG] & [GG] & [BG] & [BC] & [BM] & [BY] & [BR] & [BG] & [BB] & [CB] & [MB] & [YB] & [RB] & [GB] & [BB] \end{array} \right\}$$

$$C_{\bullet}^{\circ} = \left\{ \begin{array}{cccccccccccccccccccccc} [R_o] & [G_o] & [G_o] & [G_o] & [G_o] \\ [CR] & [CG] & [CB] & [CC] & [CM] & [CY] & [RC] & [GC] & [BC] & [CC] & [MC] & [YC] & [MR] & [MG] & [MB] & [MC] & [MM] \\ [\bullet G] & [\bullet B] & [\bullet B] & [\bullet B] & [\bullet B] \\ [RM] & [GM] & [BM] & [CM] & [MM] & [YM] & [B_o] & [BY] & [B_o] & [YB] & [B_o] & [YC] & [B_o] & [YY] & [B_o] & [RY] & [\bullet B] \\ [Co] & [M_o] & [M_o] & [M_o] & [M_o] \\ [RR] & [RG] & [RB] & [RC] & [RM] & [RY] & [RR] & [GR] & [BR] & [CR] & [MR] & [YR] & [GR] & [GG] & [GB] & [GC] & [GM] \\ [\bullet M] & [Y_o] & [\bullet Y] & [\bullet Y] & [\bullet Y] & [\bullet Y] \\ [RG] & [GG] & [BG] & [CG] & [MG] & [YG] & [BR] & [BG] & [BB] & [BC] & [BM] & [BY] & [RB] & [GB] & [BB] & [CB] & [MB] & [YB] \end{array} \right\}$$

$$C_{\bullet}^{\bullet\bullet} = \left\{ \begin{bmatrix} R\bullet \\ C\bullet \end{bmatrix}, \begin{bmatrix} G\bullet \\ M\bullet \end{bmatrix}, \begin{bmatrix} B\bullet \\ Y\bullet \end{bmatrix}, \begin{bmatrix} C\bullet \\ R\bullet \end{bmatrix}, \begin{bmatrix} M\bullet \\ G\bullet \end{bmatrix}, \begin{bmatrix} Y\bullet \\ B\bullet \end{bmatrix}, \begin{bmatrix} \bullet R \\ \bullet C \end{bmatrix}, \begin{bmatrix} \bullet G \\ \bullet M \end{bmatrix}, \begin{bmatrix} \bullet B \\ \bullet Y \end{bmatrix}, \begin{bmatrix} \bullet C \\ \bullet R \end{bmatrix}, \begin{bmatrix} \bullet M \\ \bullet G \end{bmatrix}, \begin{bmatrix} \bullet Y \\ \bullet B \end{bmatrix} \right\}$$

Table A3

The collections of $C_o^{c_1 c_2}$ and $C_\bullet^{c_1 c_2}$ in Construction 3.

$$C_{\circ}^{\infty} = \left\{ \begin{bmatrix} RR \\ RG \\ RG \end{bmatrix} \begin{bmatrix} RG \\ RR \\ RR \end{bmatrix} \begin{bmatrix} RR \\ GR \\ GR \end{bmatrix} \begin{bmatrix} GR \\ GG \\ GG \end{bmatrix} \begin{bmatrix} GR \\ GG \\ GG \end{bmatrix} \begin{bmatrix} GG \\ RG \\ RG \end{bmatrix} \begin{bmatrix} RG \\ GG \\ RG \end{bmatrix} \right\} \quad C_{\circ}^{\bullet} = \left\{ \begin{bmatrix} RR \\ R\bullet \\ R\bullet \end{bmatrix} \begin{bmatrix} RG \\ R\bullet \\ R\bullet \end{bmatrix} \begin{bmatrix} RR \\ \bullet R \\ \bullet R \end{bmatrix} \begin{bmatrix} GR \\ \bullet R \\ \bullet R \end{bmatrix} \begin{bmatrix} GR \\ G\bullet \\ G\bullet \end{bmatrix} \begin{bmatrix} GG \\ G\bullet \\ G\bullet \end{bmatrix} \begin{bmatrix} RG \\ \bullet G \\ \bullet G \end{bmatrix} \begin{bmatrix} GG \\ \bullet G \\ \bullet G \end{bmatrix} \right\}$$

$$C_{\circ}^{\bullet\bullet} = \left\{ \begin{bmatrix} R\bullet \\ RR \end{bmatrix}, \begin{bmatrix} R\bullet \\ RG \end{bmatrix}, \begin{bmatrix} \bullet R \\ RR \end{bmatrix}, \begin{bmatrix} \bullet R \\ GR \end{bmatrix}, \begin{bmatrix} G\bullet \\ GR \end{bmatrix}, \begin{bmatrix} G\bullet \\ GG \end{bmatrix}, \begin{bmatrix} \bullet G \\ RG \end{bmatrix}, \begin{bmatrix} \bullet G \\ GG \end{bmatrix} \right\} \quad C_{\circ}^{\bullet\bullet} = \left\{ \begin{bmatrix} R\bullet \\ R\bullet \end{bmatrix}, \begin{bmatrix} G\bullet \\ G\bullet \end{bmatrix}, \begin{bmatrix} \bullet R \\ \bullet R \end{bmatrix}, \begin{bmatrix} \bullet G \\ \bullet G \end{bmatrix} \right\}$$

$$C_{\bullet}^{\circ\circ} = \left\{ \begin{bmatrix} RR \\ GG \end{bmatrix}, \begin{bmatrix} RG \\ GR \end{bmatrix}, \begin{bmatrix} RR \\ GG \end{bmatrix}, \begin{bmatrix} GR \\ RG \end{bmatrix}, \begin{bmatrix} GR \\ RG \end{bmatrix}, \begin{bmatrix} GG \\ RR \end{bmatrix}, \begin{bmatrix} RG \\ GR \end{bmatrix}, \begin{bmatrix} GG \\ RR \end{bmatrix} \right\} \quad C_{\bullet}^{\bullet} = \left\{ \begin{bmatrix} RR \\ G\bullet \end{bmatrix}, \begin{bmatrix} RG \\ G\bullet \end{bmatrix}, \begin{bmatrix} RR \\ \bullet G \end{bmatrix}, \begin{bmatrix} GR \\ \bullet G \end{bmatrix}, \begin{bmatrix} GR \\ R\bullet \end{bmatrix}, \begin{bmatrix} GG \\ R\bullet \end{bmatrix}, \begin{bmatrix} RG \\ \bullet R \end{bmatrix}, \begin{bmatrix} GG \\ \bullet R \end{bmatrix} \right\}$$

$$C_{\bullet}^{\bullet \circ} = \left\{ \begin{bmatrix} R \bullet \\ GR \end{bmatrix}, \begin{bmatrix} R \bullet \\ GG \end{bmatrix}, \begin{bmatrix} \bullet R \\ RG \end{bmatrix}, \begin{bmatrix} \bullet R \\ GG \end{bmatrix}, \begin{bmatrix} G \bullet \\ RR \end{bmatrix}, \begin{bmatrix} G \bullet \\ RG \end{bmatrix}, \begin{bmatrix} \bullet G \\ RR \end{bmatrix}, \begin{bmatrix} \bullet G \\ GR \end{bmatrix} \right\} \quad C_{\bullet}^{\bullet \bullet} = \left\{ \begin{bmatrix} R \bullet \\ G \bullet \end{bmatrix}, \begin{bmatrix} \bullet R \\ G \bullet \end{bmatrix}, \begin{bmatrix} G \bullet \\ R \bullet \end{bmatrix}, \begin{bmatrix} \bullet G \\ R \bullet \end{bmatrix} \right\}$$

Table A4

The matrices of cover images of $A^{c_1 c_2 c_3}$ of (2, 3)-CBW-EVCS in Construction 4.

$$\begin{aligned} A^{\circ\circ\circ} &= \left\{ \begin{bmatrix} R \\ G \\ B \end{bmatrix} \begin{bmatrix} B \\ R \\ G \end{bmatrix} \begin{bmatrix} G \\ B \\ R \end{bmatrix} \right\}, & A^{\circ\circ\bullet} &= \left\{ \begin{bmatrix} R \\ G \\ \bullet \end{bmatrix} \begin{bmatrix} B \\ R \\ \bullet \end{bmatrix} \begin{bmatrix} G \\ B \\ \bullet \end{bmatrix} \right\}, & A^{\circ\bullet\circ} &= \left\{ \begin{bmatrix} R \\ \bullet \\ B \end{bmatrix} \begin{bmatrix} B \\ \bullet \\ G \end{bmatrix} \begin{bmatrix} G \\ \bullet \\ R \end{bmatrix} \right\}, & A^{\circ\bullet\bullet} &= \left\{ \begin{bmatrix} R \\ \bullet \\ \bullet \end{bmatrix} \begin{bmatrix} B \\ \bullet \\ \bullet \end{bmatrix} \begin{bmatrix} G \\ \bullet \\ \bullet \end{bmatrix} \right\} \\ A^{\bullet\circ\circ} &= \left\{ \begin{bmatrix} \bullet \\ G \\ B \end{bmatrix} \begin{bmatrix} \bullet \\ R \\ G \end{bmatrix} \begin{bmatrix} \bullet \\ B \\ R \end{bmatrix} \right\}, & A^{\bullet\circ\bullet} &= \left\{ \begin{bmatrix} \bullet \\ G \\ \bullet \end{bmatrix} \begin{bmatrix} \bullet \\ R \\ \bullet \end{bmatrix} \begin{bmatrix} \bullet \\ B \\ \bullet \end{bmatrix} \right\}, & A^{\bullet\bullet\circ} &= \left\{ \begin{bmatrix} \bullet \\ \bullet \\ B \end{bmatrix} \begin{bmatrix} \bullet \\ \bullet \\ G \end{bmatrix} \begin{bmatrix} \bullet \\ \bullet \\ R \end{bmatrix} \right\}, & A^{\bullet\bullet\bullet} &= \left\{ \begin{bmatrix} \bullet \\ \bullet \\ \bullet \end{bmatrix} \begin{bmatrix} \bullet \\ \bullet \\ \bullet \end{bmatrix} \begin{bmatrix} \bullet \\ \bullet \\ \bullet \end{bmatrix} \right\} \end{aligned}$$

Table A5

The matrices of cover images of $A^{c_1 c_2 c_3 c_4}$ in Example 4.

$$\begin{aligned} A^{\circ\circ\circ\circ} &= \left\{ \begin{bmatrix} R\bullet \\ G\bullet \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet R \\ R\bullet \\ \bullet R \\ B\bullet \end{bmatrix} \begin{bmatrix} B\bullet \\ \bullet R \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} G\bullet \\ B\bullet \\ G\bullet \\ R\bullet \end{bmatrix} \right\}, & A^{\circ\circ\circ\bullet} &= \left\{ \begin{bmatrix} R\bullet \\ G\bullet \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet R \\ R\bullet \\ \bullet R \\ B\bullet \end{bmatrix} \begin{bmatrix} B\bullet \\ \bullet R \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} G\bullet \\ B\bullet \\ G\bullet \\ R\bullet \end{bmatrix} \right\}, & A^{\circ\circ\bullet\circ} &= \left\{ \begin{bmatrix} R\bullet \\ G\bullet \\ \bullet R \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet R \\ R\bullet \\ \bullet R \\ \bullet R \end{bmatrix} \begin{bmatrix} B\bullet \\ \bullet R \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} G\bullet \\ B\bullet \\ G\bullet \\ R\bullet \end{bmatrix} \right\} \\ A^{\circ\circ\bullet\bullet} &= \left\{ \begin{bmatrix} R\bullet \\ G\bullet \\ \bullet R \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet R \\ R\bullet \\ \bullet R \\ \bullet R \end{bmatrix} \begin{bmatrix} B\bullet \\ \bullet R \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} G\bullet \\ B\bullet \\ \bullet R \\ \bullet R \end{bmatrix} \right\}, & A^{\circ\bullet\circ\circ} &= \left\{ \begin{bmatrix} R\bullet \\ \bullet R \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet R \\ \bullet R \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} B\bullet \\ \bullet R \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} G\bullet \\ \bullet R \\ G\bullet \\ \bullet R \end{bmatrix} \right\}, & A^{\circ\bullet\circ\bullet} &= \left\{ \begin{bmatrix} R\bullet \\ \bullet R \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet R \\ \bullet R \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} B\bullet \\ \bullet R \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} G\bullet \\ \bullet R \\ G\bullet \\ \bullet R \end{bmatrix} \right\} \\ A^{\circ\bullet\bullet\circ} &= \left\{ \begin{bmatrix} R\bullet \\ \bullet R \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet R \\ \bullet R \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} B\bullet \\ \bullet R \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} G\bullet \\ \bullet R \\ G\bullet \\ \bullet R \end{bmatrix} \right\}, & A^{\circ\bullet\bullet\bullet} &= \left\{ \begin{bmatrix} R\bullet \\ \bullet R \\ \bullet R \\ B\bullet \end{bmatrix} \begin{bmatrix} \bullet R \\ \bullet R \\ \bullet R \\ B\bullet \end{bmatrix} \begin{bmatrix} B\bullet \\ \bullet R \\ \bullet R \\ B\bullet \end{bmatrix} \begin{bmatrix} G\bullet \\ \bullet R \\ \bullet R \\ B\bullet \end{bmatrix} \right\}, & A^{\bullet\circ\circ\circ} &= \left\{ \begin{bmatrix} \bullet \bullet \\ G\bullet \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet R \\ R\bullet \\ G\bullet \\ B\bullet \end{bmatrix} \begin{bmatrix} \bullet R \\ G\bullet \\ R\bullet \\ B\bullet \end{bmatrix} \begin{bmatrix} \bullet R \\ B\bullet \\ G\bullet \\ R\bullet \end{bmatrix} \right\} \\ A^{\bullet\circ\circ\bullet} &= \left\{ \begin{bmatrix} \bullet \bullet \\ G\bullet \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet R \\ R\bullet \\ G\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet R \\ G\bullet \\ R\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet R \\ B\bullet \\ G\bullet \\ \bullet R \end{bmatrix} \right\}, & A^{\bullet\circ\bullet\circ} &= \left\{ \begin{bmatrix} \bullet \bullet \\ G\bullet \\ \bullet R \\ B\bullet \end{bmatrix} \begin{bmatrix} \bullet R \\ R\bullet \\ \bullet R \\ B\bullet \end{bmatrix} \begin{bmatrix} \bullet R \\ \bullet R \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet R \\ \bullet R \\ G\bullet \\ R\bullet \end{bmatrix} \right\}, & A^{\bullet\circ\bullet\bullet} &= \left\{ \begin{bmatrix} \bullet \bullet \\ G\bullet \\ \bullet R \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet R \\ R\bullet \\ \bullet R \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet R \\ \bullet R \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet R \\ \bullet R \\ G\bullet \\ \bullet R \end{bmatrix} \right\} \\ A^{\bullet\bullet\circ\circ} &= \left\{ \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ G\bullet \\ B\bullet \end{bmatrix} \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ R\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ G\bullet \\ R\bullet \end{bmatrix} \right\}, & A^{\bullet\bullet\circ\bullet} &= \left\{ \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ B\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ G\bullet \\ R\bullet \end{bmatrix} \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ R\bullet \\ \bullet R \end{bmatrix} \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ R\bullet \\ \bullet R \end{bmatrix} \right\}, & A^{\bullet\bullet\bullet\circ} &= \left\{ \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ \bullet R \\ B\bullet \end{bmatrix} \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ R\bullet \\ G\bullet \end{bmatrix} \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ G\bullet \\ R\bullet \end{bmatrix} \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ G\bullet \\ R\bullet \end{bmatrix} \right\} \\ A^{\bullet\bullet\bullet\bullet} &= \left\{ \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ \bullet \bullet \\ \bullet \bullet \end{bmatrix} \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ \bullet \bullet \\ \bullet \bullet \end{bmatrix} \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ \bullet \bullet \\ \bullet \bullet \end{bmatrix} \begin{bmatrix} \bullet \bullet \\ \bullet \bullet \\ \bullet \bullet \\ \bullet \bullet \end{bmatrix} \right\} \end{aligned}$$

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