# A Probabilistic Model of $(t, n)$ Visual Cryptography Scheme With Dynamic Group 

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#### Abstract

The ( $t, n$ ) visual cryptography (VC) is a secret sharing scheme where a secret image is encoded into $n$ transparencies, and the stacking of any $t$ out of $n$ transparencies reveals the secret image. The stacking of $t-1$ or fewer transparencies is unable to extract any information about the secret. We discuss the additions and deletions of users in a dynamic user group. To reduce the overhead of generating and distributing transparencies in user changes, this paper proposes a $(t, n)$ VC scheme with unlimited $n$ based on the probabilistic model. The proposed scheme allows $n$ to change dynamically in order to include new transparencies without regenerating and redistributing the original transparencies. Specifically, an extended VC scheme based on basis matrices and a probabilistic model is proposed. An equation is derived from the fundamental definitions of the $(t, n)$ VC scheme, and then the $(t, \infty)$ VC scheme achieving maximal contrast can be designed by using the derived equation. The maximal contrasts with $t=2$ to 6 are explicitly solved in this paper.


Index Terms-Contrast, random grids (RGs), secret sharing, visual cryptography (VC).

## I. Introduction

VISUAL cryptography (VC) is a branch of secret sharing. In the VC scheme, a secret image is encoded into $n$ transparencies, and the content of each transparency is noise-like so that the secret information cannot be retrieved from any one transparency via human visual observation or signal analysis techniques. In general, $\mathrm{a}(t, n)$-threshold VC scheme has the following properties: The stacking of any $t$ out of those VC generated $n$ transparencies can reveal the secret by visual perception, but the stacking of any $t-1$ or fewer number of transparencies cannot retrieve any information other than the size of the secret image. Naor and Shamir [1] proposed a $(t, n)$-threshold VC scheme based on basis matrices, and the model had been further studied and extended. The related works include the VC schemes based on probabilistic models [2]-[4], general access structures [5], [6], VC over halftone images [7], [8], VC for color images [9], cheating in VC [10], [11], the general formula of VC schemes [12], and region incrementing VC [13].

Contrast is one of the important performance metrics for VC schemes. Generally, the stacking revelation of the secret with

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higher contrast represents the better visual quality, and therefore the stacking secret with high contrast is the goal of pursuit in VC designs. Naor and Shamir [1] define a contrast formula which has been widely used in many studies. Based on the definition of contrast, there are studies attempting to achieve the contrast bound of $(t, n)$ VC scheme [4], [14]-[20]. For instance, Blundo et al. [17] give the optimal contrast of $(2, n)$ VC schemes. Hofmeister et al. [19] provide a linear program which is able to compute exactly the optimal contrast for $(t, n)$ VC schemes. Krause and Simon [20] provide the upper bound and lower bound of the optimal contrast for $(t, n)$ VC schemes. Moreover, there exist VC related researches using differential definitions of contrast [21]-[23]. Another important metric is the pixel expansion denoting the number of subpixels in transparency used to encode a secret pixel. The minimization of pixel expansions has been investigated in previous studies [24], [25].

The probabilistic model of the VC scheme was first introduced by Ito et al. [2], where the scheme is based on the basis matrices, but only one column of the matrices is chosen to encode a binary secret pixel, rather than the traditional VC scheme utilizing the whole basis matrices. The size of the generated transparencies is identical to the secret image. Yang [31] also proposed a probabilistic model of $(t, n)$ VC scheme, and the two cases $(2, n)$ and $(n, n)$ are explicitly constructed to achieve the optimal contrast. Based on Yang [31], Cimato et al. [32] proposed a generalized VC scheme in which the pixel expansion is between the probabilistic model of VC scheme and the traditional VC scheme.

Encrypting an image by random grids (RGs) was first introduced by Kafri and Keren [26] in 1987. A binary secret image is encoded into two noise-like transparencies with the same size of the original secret image, and stacking of the two transparencies reveals the content of the secret. Comparing RGs with basis matrices, one of the major advantages is that the size of generated transparencies is unexpanded. The RG scheme is similar to the probabilistic model of the VC scheme, but the RG scheme is not based on the basis matrices. The recent studies include the RG for color image [27], $(2, n)$ RG, and $(n, n)$ RG schemes [28], [29]. We also compare the proposed method with $(2, n)$ RG [29] in Section IV.

We consider the scenario of a dynamic user group, where $n^{\prime \prime}$ new participants are to join the user group with $n^{\prime}$ original participants; and the transparencies need to accommodate the new $n=n^{\prime}+n^{\prime \prime}$ users. If the transparencies are to be generated with the traditional $\left(t, n^{\prime}\right) \mathrm{VC}$ scheme, the $n^{\prime}$ original transparencies need to be discarded, and the $n^{\prime}+n^{\prime \prime}$ new transparencies need to be generated with the traditional $\left(t, n^{\prime}+n^{\prime \prime}\right) \mathrm{VC}$ scheme. The regeneration and redistribution of the whole transparencies consume computing and communication resources and may lead to
the potential security vulnerability. The addition and deletion of users are further discussed in Section IV-A.

In this paper, we propose a probabilistic model of $(t, n) \mathrm{VC}$ scheme with unlimited $n$. The major contribution is that the proposed scheme accommodates dynamic changes of users in the group sharing a VC secret. The proposed scheme allows changes of users without regeneration and redistribution of VC transparencies, which reduce the computing and communication resources in accommodating user changes. The scheme is capable of generating an arbitrary number of transparencies and the explicit algorithms are proposed to generate the transparencies. For a group with $n^{\prime}$ initial users, the proposed Algorithm 1 explicitly generates the required $n^{\prime}$ transparencies. For $n^{\prime \prime}$ newly joining participants, the $n^{\prime \prime}$ new transparencies can be explicitly generated through Algorithm 2, and the $n^{\prime \prime}$ new transparencies can be distributed to the $n^{\prime \prime}$ new participants without the need to update the original transparencies. The secondary contribution is that this paper designs an implementation of $(t, \infty)$ VC based on the probabilistic model, and the proposed scheme allows the unlimited number of users. For the conventional $(t, n)$ VC scheme to implement the case $n \rightarrow \infty$, the mathematical manipulations of infinite size of basis matrices and variables are often required, which is computationally prohibitive. Our approach designs an implementation scheme which is capable of producing a finite subset of the complete infinite transparencies through the proposed Algorithms 1 and 2 , with computationally feasible operations. We also derive an optimization problem $L(t)$ to solve the maximal contrast of the proposed $(t, \infty)$ VC scheme.

## Notations

| $t$ | The threshold of a $(t, n) \mathrm{VC}$ scheme. |
| :---: | :---: |
| $n$ | The number of generated transparencies of a $(t, n) \mathrm{VC}$ scheme. |
| $m$ | The number of subpixels to encode a secret pixel; i.e., the width of the basis matrices. |
| $\oplus$ | Stacking operation, equivalent to the bitwise operation "OR." |
| $C_{0}$ and $C_{1}$ | Two collections of $n \times m$ Boolean matrices. |
| $H(v)$ | The number of ones in a vector $v$. |
| $\alpha$ | The contrast of the VC scheme based on basis matrices. |
| $B_{0}$ and $B_{1}$ | Two $n \times m$ Boolean matrices where $B_{0} \in C_{0}$ and $B_{1} \in C_{1}$. |
| $A_{j, n}$ | A $n \times\binom{ n}{j}$ Boolean matrix containing all possible combinations of $j$ zeros and $(n-j)$ ones as the columns. |
| $H_{0}$ and $H_{1}$ | Two vectors to represent the multiplicities of $A_{j}$ in $B_{0}$ and $B_{1}$. |
| $\alpha^{\prime}$ | The contrast of the probabilistic model of VC scheme. |
| $b_{0}, b_{1}$ | A column randomly selected from $B_{0}$ and $B_{1}$. |


| $h(v)$ | The frequency probability of appearing ones in a vector $v$. |
| :---: | :---: |
| $P_{0}, P_{1}$ | The discrete probability distribution of all columns in $B_{0}$ and $B_{1}$. |
| $E(\bullet)$ | A memoryless binary sequence with values of elements as 0 or 1 , where the probability of assigning each element $e$ to 0 is $\bullet$. |
| $\Delta P$ | A vector $\Delta P=P_{0}-P_{1}$. |
| $X, Y$ | Two vectors to record the nonzero terms in $\Delta P$. |
| $L(t)$ | An optimization problem to find the optimal contrast of $(t, n \rightarrow \infty)$ VC scheme. |
| $S$ | A binary secret image. |
| $s$ | A ready-to-process pixel taken from $S$. |
| $T_{i}$ | The $i$ th generated transparency. |
| $t_{i}$ | A pixel at $T_{i}$, and the position corresponds to the position of $s$. |
| $n^{\prime}$ and $n^{\prime \prime}$ | $n^{\prime}$ is the number of original participants in the user group initially, and $n^{\prime \prime}$ is the number of new participants to join the user group. |
| Z | An index table where $Z[w, h]$ is the index of the used memoryless sequence $E\left(x_{Z[w, h]}\right)$ to encode the secret pixel $s[w, h]$. |
| $P(\bullet)$ | The probability of the event $\bullet$. |
| $G(\bullet)$ | A group which contains all columns of the matrix - as the elements. |

## II. Extended Version Based on Basis Matrices and Probabilistic Model

## A. Basis Matrices

The basis matrices of $(t, n)$ VC scheme were first introduced by Naor and Shamir [1]. In this paper, a white-and-black secret image or pixel is also described as a binary image or pixel. In the basis matrices, to encode a binary secret image $S$, each secret pixel $s \in\{$ white, black $\}$ will be turned into $n$ blocks at the corresponding position of $n$ transparencies $T_{1}, T_{2}, \ldots, T_{n}$, respectively. Each block consists of $m$ subpixels and each subpixel is opaque or transparent. Throughout this paper, we use 0 to indicate a transparent subpixel and 1 to indicate an opaque subpixel. If any two subpixels are stacked with matching positions, the representation of a stacked pixel may be transparent, when the two corresponding pixels are both transparent. Otherwise, the stacked pixel is opaque. Let $\oplus$ denote the stacking operation, defined as

$$
0 \oplus 0=0,0 \oplus 1=1,1 \oplus 0=1,1 \oplus 1=1
$$

Actually, $\oplus$ can be treated as the bitwise operation "OR." It is noted that we use the notation $T_{i} \oplus T_{j}$ to indicate the stacking of the two transparencies $T_{i}$ and $T_{j}$, since $T_{i}$ and $T_{j}$ can be treated as two Boolean matrices.

For the basis matrices, two collections of $n \times m$ Boolean matrices $C_{0}$ and $C_{1}$ are constructed to encode the binary pixel $s$, respectively. Each row of the matrix in $C_{0}$ and $C_{1}$ corresponds to an encoded block, and the elements represent the subpixels. Before describing the definitions of $C_{0}$ and $C_{1}$, we first explain how to encode $s$. For $s$ being white, the dealer randomly chooses a matrix from $C_{0}$ with uniform distribution and then sends all the rows to each $T_{i}$, respectively. For $s$ being black, the dealer randomly chooses a matrix from $C_{1}$ with uniform distribution and then sends all the rows to each $T_{i}$, respectively. The $C_{0}$ and $C_{1}$ are required to meet the conditions described in Definition 1. Let $H(v)$ denote the hamming weight of a $(0-1)$-vector $v$ (i.e., the number of ones in $v$ ).

Definition 1: A $(t, n)$ VC scheme with $m$ subpixels and contrast $\alpha>0$ can be represented as two collections of $n \times m$ Boolean matrices $C_{0}$ and $C_{1}$. Let $d$ be a constant integer. A valid VC scheme is required to meet the following conditions [1]:

1) For any matrix in $C_{0}$, the stacked $v$ of any $t$ out of the $n$ rows in the matrix satisfies $H(v) \leq d-\alpha m$.
2) For any matrix in $C_{1}$, the stacked $v$ of any $t$ out of the $n$ rows in the matrix satisfies $H(v) \geq d$.
3) For any $k$-element subset $\left\{i_{1}, i_{2}, \ldots, i_{k}\right\} \subseteq\{1,2, \ldots, n\}$ and $k<t$, the two collections of $k \times m$ matrices obtained by restricting each matrix in $C_{0}$ and $C_{1}$, to rows $i_{1}, i_{2}, \ldots, i_{k}$ are indistinguishable in the sense that they contain the same matrices with the same frequencies.
The first and the second conditions represent the contrast requirements. In general, with larger $\alpha$, the stacking result is more visually distinguishable. The third condition represents the security requirement. A valid VC must be able to prevent the secret pixels from being revealed by analyzing the patterns or probability distributions from $k$ transparencies for $k<t$.

If we find an $n \times m$ Boolean matrix $B_{0} \in C_{0}$ and an $n \times$ $m$ Boolean matrix $B_{1} \in C_{1}$, we can construct $C_{0}$ and $C_{1}$ by permuting all columns of $B_{0}$ and $B_{1}$ [1], expressed as

$$
\begin{aligned}
C_{0}= & \{\text { All the matrices obtained } \\
& \text { by permuting all columns of } \left.B_{0}\right\}, \\
C_{1}= & \{\text { All the matrices obtained } \\
& \text { by permuting all columns of } \left.B_{1}\right\} .
\end{aligned}
$$

If two $n \times m$ Boolean matrices $B_{i}^{\prime}$ and $B_{i}^{\prime \prime}$ can be adjusted to become the same matrix with reordering columns, $B_{i}^{\prime}$ and $B_{i}^{\prime \prime}$ are equivalent in terms of generating $C_{i}$. Therefore, the orders of columns of $B_{0}$ and $B_{1}$ are technically insignificant.

The previous studies indicated that the "totally symmetric" scheme [19] (also called the canonical form [15]) is only considered for constructing $B_{0}$ and $B_{1}$, since the studies [15] and [19] proved that the "totally symmetric" schemes cover the $(t, n)$ VC schemes with optimal contrast. Let $A_{j, n}$ denote an $n \times$ $\binom{n}{j}$ Boolean matrix consisting of all possible combinations of $j$ zeros and $(n-j)$ ones as columns of $A_{j, n}$. A matrix $B_{i}$ is called "totally symmetric" if $B_{i}$ can be divided into some $A_{j, n}, j \in\{0,1, \ldots, n\}$ with horizontal concatenation. Thus, the $B_{0}$ and $B_{1}$ of a "totally symmetric" VC scheme can be presented as two vectors $H_{0}=\left(h_{0,0}, h_{1,0}, \ldots, h_{n, 0}\right)$ and $H_{1}=$
$\left(h_{0,1}, h_{1,1}, \ldots, h_{n, 1}\right)$ in which $h_{j, i}$ is the multiplicity of $A_{j, n}$ in $B_{i}$. The relation between $m$ and $h_{j, i}$ is

$$
\begin{equation*}
m=\sum_{j=0}^{n}\binom{n}{j} h_{j, i} \tag{1}
\end{equation*}
$$

Example 1: An example of a $(3,4) \mathrm{VC}$ scheme is

$$
\begin{aligned}
B_{0} & =\left[\begin{array}{llllll}
0 & 0 & 0 & 1 & 1 & 1 \\
0 & 0 & 1 & 0 & 1 & 1 \\
0 & 0 & 1 & 1 & 0 & 1 \\
0 & 0 & 1 & 1 & 1 & 0
\end{array}\right] \\
\text { and } B_{1} & =\left[\begin{array}{llllll}
1 & 1 & 1 & 0 & 0 & 0 \\
1 & 1 & 0 & 1 & 0 & 0 \\
1 & 1 & 0 & 0 & 1 & 0 \\
1 & 1 & 0 & 0 & 0 & 1
\end{array}\right] .
\end{aligned}
$$

The $A_{j, n}$ are

$$
\begin{aligned}
& A_{0,4}=\left[\begin{array}{l}
1 \\
1 \\
1 \\
1
\end{array}\right], A_{1,4}=\left[\begin{array}{llll}
0 & 1 & 1 & 1 \\
1 & 0 & 1 & 1 \\
1 & 1 & 0 & 1 \\
1 & 1 & 1 & 0
\end{array}\right], \\
& A_{2,4}=\left[\begin{array}{llllll}
1 & 1 & 1 & 0 & 0 & 0 \\
1 & 0 & 0 & 1 & 1 & 0 \\
0 & 1 & 0 & 1 & 0 & 1 \\
0 & 0 & 1 & 0 & 1 & 1
\end{array}\right], A_{3,4}=\left[\begin{array}{llll}
1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 0 & 0 & 1
\end{array}\right] \\
& A_{4,4}=\left[\begin{array}{l}
0 \\
0 \\
0 \\
0
\end{array}\right] .
\end{aligned}
$$

Thus, $B_{0}$ is represented as $H_{0}=(0,1,0,0,2)$, and $B_{1}$ is represented as $H_{1}=(2,0,0,1,0)$.

## B. Probabilistic Model of VC Scheme

In the scenario where $n$ is very large, i.e., $n \rightarrow \infty$, to construct $B_{i}$ is impractical. Thus, the proposal of a new VC scheme is required in order to overcome the above problem. The probabilistic model of VC is first introduced by Ito et al. [2]. Instead of the basis matrices expanding a secret pixel into a block with $m$ subpixels in transparencies, the probabilistic model of VC only uses one subpixel to represent one secret pixel. The idea is briefly described as follows.

To encode the secret image $S$, the probabilistic model of VC constructs two basis matrices $B_{0}$ and $B_{1}$. For the secret pixel $s$ being white, the dealer randomly chooses a column $b_{0}$ from $B_{0}$ with uniform distribution and sends each element $t_{i}$ to $T_{i}$, respectively. For $s$ being black, the dealer randomly chooses a column $b_{1}$ from $B_{1}$ with uniform distribution and sends each element $t_{i}$ to $T_{i}$, respectively. To decode $S$, any $t$ out of the $n$ transparencies $T_{i_{1}}, T_{i_{2}}, \ldots, T_{i_{t}}$ are stacked $T_{i_{1}} \oplus T_{i_{2}} \oplus \cdots \oplus T_{i_{t}}$. Then the region $\{s=$ white $\mid s \in S\}$ is probabilistically more likely to appear white than the region $\{s=$ black $\mid s \in S\}$ is. Thus, we have

$$
\begin{align*}
P\left(t_{i_{1}} \oplus t_{i_{2}} \oplus\right. & \left.\cdots \oplus t_{i_{t}}=0 \mid s=\text { white }\right) \\
& >P\left(t_{i_{1}} \oplus t_{i_{2}} \oplus \cdots \oplus t_{i_{t}}=0 \mid s=\text { black }\right) \tag{2}
\end{align*}
$$

where $P(\bullet)$ is the probability of the event $\bullet$, and $t_{i_{j}} \in T_{i_{j}}$ is a binary value taken from $b_{0}$ (if $s$ is white) or $b_{1}$ (if $s$ is black). The notation $\left\{t_{i_{j}}, t_{i_{j}}, \ldots, t_{i_{t}}\right\} \subseteq b_{i}$ means that $\left\{t_{i_{j}}, t_{i_{j}}, \ldots, t_{i_{t}}\right\}$ are
any $t$ elements in $b_{i}$. As a result, $S$ can be interpreted by visual perception since the human visual system can be treated as a low pass filter. On the other hand, to protect $S$, in the case where the number of stacked transparencies $k$ is smaller than $t$, the region corresponding to the white pixels in $S$ is probabilistically identical, in terms of appearing black or white, to the region corresponding to the black secret pixels. In other words, the regions $\{s=$ white $\mid s \in S\}$ are probabilistically indistinguishable to the regions $\{s=$ black $\mid s \in S\}$ on the stacking result. The property is expressed as

$$
\begin{align*}
& P\left(t_{i_{1}} \oplus t_{i_{2}} \oplus \cdots \oplus t_{i_{k}}=0 \mid s=\text { white }\right) \\
& \quad>P\left(t_{i_{1}} \oplus t_{i_{2}} \oplus \cdots \oplus t_{i_{k}}=0 \mid s=\text { black }\right), \quad \text { for } k<t \tag{3}
\end{align*}
$$

For the contrast condition described in probability model [2], the contrast between white and black pixels is calculated from the differential "probability" of appearing zero and not the differential "frequency" used in basis matrices. The contrast is defined as

$$
\begin{align*}
& \alpha^{\prime}=P\left(t_{i_{1}} \oplus t_{i_{2}} \oplus \cdots \oplus t_{i_{t}}=0 \mid s=\text { white }\right) \\
&  \tag{4}\\
& \quad-P\left(t_{i_{1}} \oplus t_{i_{2}} \oplus \cdots \oplus t_{i_{t}}=0 \mid s=\text { black }\right)
\end{align*}
$$

to distinguish $\alpha$ used in Definition 1. Let $h(v)=H(v) / m$ denote the frequency probability of appearing ones in $v$. The definition of $(t, n)$ Probabilistic VC scheme based on $B_{0}$ and $B_{1}$ is given as follows.

Definition 2: A $(t, n)$ Probabilistic VC scheme with contrast $\alpha^{\prime}>0$ can be represented as two $n \times m$ Boolean matrices $B_{0}$ and $B_{1}$. Let $d^{\prime}$ be a constant integer; a valid Probabilistic VC must satisfy the following conditions:

1) The stacked $v$ of any $t$ out of the $n$ rows in $B_{0}$ satisfies $h(v) \leq d^{\prime}-\alpha^{\prime}$.
2) The stacked $v$ of any $t$ out of the $n$ rows in $B_{1}$ satisfies $h(v) \geq d^{\prime}$.
3) For any $k$-element subset $\left\{i_{1}, i_{2}, \ldots, i_{k}\right\} \subseteq\{1,2, \ldots, n\}$ and $k<t$, the two $k \times m$ matrices obtained by restricting each matrix in $B_{0}$ and $B_{1}$, to rows $i_{1}, i_{2}, \ldots, i_{k}$ are the identical.

## C. Proposed VC Scheme

The proposed method is based on the basis matrices and the idea of probabilistic model. For a $(t, n)$ VC scheme, the "totally symmetric" form of $B_{0}$ and $B_{1}$ are both constructed and described as $H_{0}$ and $H_{1}$, respectively. For a column $b_{i}$ randomly selected from $B_{i}$ with uniform distribution, let $p_{j, i}$ be the probability that $b_{i}$ consists of $j$ zeros and $(n-j)$ ones. The $p_{j, i}$ is expressed by

$$
\begin{equation*}
p_{j, i}=P\left(H\left(b_{i}\right)=n-j \mid b_{i} \in G\left(B_{i}\right)\right) \tag{5}
\end{equation*}
$$

where $G(\bullet)$ outputs a group which contains all columns of the matrix $\bullet$ as the elements, and we have

$$
\begin{align*}
\sum_{j=0}^{n} p_{j, i} & =1  \tag{6}\\
p_{j, i} & =\binom{n}{j} \frac{h_{j, i}}{m} \tag{7}
\end{align*}
$$

Thus, $P_{0}=\left(p_{0,0}, p_{1,0}, \ldots, p_{n, 0}\right)$ and $P_{1}=$ ( $p_{0,1}, p_{1,1}, \ldots, p_{n, 1}$ ) represent the probability distributions by which the columns are selected to encode white and
black secret pixels $s$, and by (7), $P_{0}$ and $P_{1}$ are homogeneous to $H_{0}$ and $H_{1}$ for describing $B_{0}$ and $B_{1}$, respectively. According to the probabilistic model, we randomly choose a column $b_{i}$ from $B_{i}$; since $B_{i}$ consists of many submatrices $A_{j, n}$, suppose a column is chosen from $A_{j, n}$. This step can be simulated by rearranging a vector consisting of $j$ zeros and ( $n-j$ ) ones randomly, and all possible combinations are equally likely to be a candidate, since all columns in $A_{j, n}$ have equal chances to be selected. This property reduces the memory requirement to store $B_{0}$ and $B_{1}$ during encoding. For the case $n \rightarrow \infty$, the generated column becomes a ( $0-1$ ) memoryless sequence $E(j / n)$ where the probability of assigning each element $e$ to 0 is $j / n$ (i.e., $P(e=0 \mid e \in E(j / n))=j / n)$. In terms of computer programming, $E(j / n)$ can be simulated with a random Boolean generator where the probability of generating 0 is $j / n$ and generating 1 is $(1-j / n)$. For convenience, in the rest of the paper, the abbreviation $(t, \infty) \operatorname{PrVC}$ scheme denotes the probabilistic model of VC with unlimited $n$.

Lemma 1: For any $k$ elements $t_{i_{1}}, t_{i_{2}}, \ldots, t_{i_{k}}$ generated from the secret pixel $s$ with $(t, \infty) \operatorname{PrVC}$ scheme, after the stacking $t_{i_{1}} \oplus t_{i_{2}} \oplus \cdots \oplus t_{i_{k}}$, the probability of the stacking result appearing zero is $P\left(t_{i_{1}} \oplus t_{i_{2}} \oplus \cdots \oplus t_{i_{k}}=0 \mid s=\right.$ white $)=$ $\sum_{j=0}^{n} p_{j, 0}\left(\frac{j}{n}\right)^{k}$ and $P\left(t_{i_{1}} \oplus t_{i_{2}} \oplus \cdots \oplus t_{i_{k}}=0 \mid s=\right.$ black $)=$ $\sum_{j=0}^{n} p_{j, 1}\left(\frac{j}{n}\right)^{k}$.

Proof: To encode the secret pixel $s$, a candidate in $\{E(j / n) \mid j=0,1, \ldots, n\}$ is chosen, and the generated elements $e_{1}, e_{2}, \ldots, e_{k}$ are assigned to $t_{i_{1}}, t_{i_{2}}, \ldots, t_{i_{k}}$, respectively. The probability of generating $E(j / n)$ as the candidate is $p_{j, i}$. Suppose $E(j / n)$ is generated to encode the secret pixel. For the case where $k$ transparencies are stacked, the probability of the secret pixel appearing zero is $(j / n)^{k}$, because the $k$ corresponding elements $e_{1}, e_{2}, \ldots, e_{k}$ at the $k$ transparencies, are required to be all zeros, and we have

$$
\begin{aligned}
P\left(e_{1} \oplus e_{2} \oplus \cdots \oplus e_{k}=0 \mid\left\{e_{1}\right.\right. & \left.\left., e_{2}, \ldots, e_{k}\right\} \subseteq E\left(\frac{j}{n}\right)\right) \\
& =\prod_{j=1}^{k} P\left(e_{j}=0\right)=\left(\frac{j}{n}\right)^{k}
\end{aligned}
$$

Therefore, the probability of a pixel appearing zero is $\sum_{j=0}^{n} p_{j, i}(j / n)^{k}$.

As stated in the security condition in Definition 2, after stacking any $k$ rows $(k<t)$ in $B_{0}$ and $B_{1}$ to obtain two vectors $v_{0}$ and $v_{1}$, we have $h\left(v_{0}\right)=h\left(v_{1}\right)$. Based on the results of Lemma 1 , the $t-1$ equations maintaining the security condition are described as

$$
\begin{align*}
\sum_{j=0}^{n} p_{j, 0}\left(\frac{j}{n}\right)^{k} & =\sum_{j=0}^{n} p_{j, 1}\left(\frac{j}{n}\right)^{k} \\
\Rightarrow \sum_{j=0}^{n} \Delta p_{j}\left(\frac{j}{n}\right)^{k} & =0 \quad \forall k=1,2, \ldots, t-1 \tag{8}
\end{align*}
$$

where $\Delta p_{j}=p_{j, 0}-p_{j, 1}$, and $\Delta P=\left(\Delta p_{0}, \Delta p_{1} \ldots, \Delta p_{n}\right)=$ $P_{0}-P_{1}$. On the other hand, by (6), we have

$$
\begin{equation*}
\sum_{j=0}^{n} \Delta p_{j}=\sum_{j=0}^{n} p_{j, 0}-\sum_{j=0}^{n} p_{j, 1}=1-1=0 \tag{9}
\end{equation*}
$$

and $\alpha^{\prime}$ is described as

$$
\begin{align*}
\alpha^{\prime} & =\sum_{j=0}^{n} p_{j, 0}\left(\frac{j}{n}\right)^{t}-\sum_{j=0}^{n} p_{j, 1}\left(\frac{j}{n}\right)^{t} \\
& =\sum_{j=0}^{n} \Delta p_{j}\left(\frac{j}{n}\right)^{t}>0 \tag{10}
\end{align*}
$$

## D. The $(t, \infty)$ PrVC Scheme With Optimal Contrast

In this section, we focus on the construction of the $(t, \infty)$ $\operatorname{PrVC}$ scheme with optimal contrast. For convenience, the abbreviation $(t, \infty)$ OptPrVC scheme is used to denote the $(t, \infty)$ PrVC scheme with optimal contrast.

Lemma 2: For $P_{0}$ and $P_{1}$ of $(t, \infty)$ OptPrVC scheme, if $p_{j, 0}>0$, then $p_{j, 1}=0$; and if $p_{j, 1}>0$, then $p_{j, 0}=0$.

Proof: Suppose $P_{0}^{\prime}=\left(p_{0,0}^{\prime}, p_{1,0}^{\prime}, \ldots, p_{l .0}^{\prime}, \ldots, p_{n, 0}^{\prime}\right)$ and $P_{1}^{\prime}=\left(p_{0,1}^{\prime}, p_{1,1}^{\prime}, \ldots, p_{l, 1}^{\prime}, \ldots, p_{n, 1}^{\prime}\right)$ are the $(t, \infty)$ OptPrVC scheme with the optimal contrast $\widehat{\alpha}^{\prime}$ but $\exists l: p_{l, 0}^{\prime}>0$ and $p_{l, 1}^{\prime}>0$. Thus, for $i \in\{0,1\}, P_{i}^{\prime}$ satisfies (6)(8)(10) which are described as

$$
\begin{gather*}
\sum_{j=0}^{n} p_{j, i}^{\prime}=1  \tag{11}\\
\sum_{j=0}^{n}\left(p_{j, 0}^{\prime}-p_{j, 1}^{\prime}\right)\left(\frac{j}{n}\right)^{k}=0 \forall k=1,2, \ldots, t-1  \tag{12}\\
\widehat{\alpha}^{\prime}=\sum_{j=0}^{n}\left(p_{j, 0}^{\prime}-p_{j, 1}^{\prime}\right)\left(\frac{j}{n}\right)^{t}>0 \tag{13}
\end{gather*}
$$

Then, a new scheme $P_{i}^{\prime \prime}=\left(p_{0, i}^{\prime \prime}, p_{1, i}^{\prime \prime}, \ldots, p_{n, i}^{\prime \prime}\right)$ with contrast $\widehat{\alpha}^{\prime \prime}$ is constructed by the following steps. First, calculate $p^{\min }=$ $\min \left\{p_{l, 0}^{\prime}, p_{l, 1}^{\prime}\right\}$; second, construct $P_{i}^{\prime \prime}$ and each term $p_{j, i}^{\prime \prime}$ is

$$
p_{j, i}^{\prime \prime}= \begin{cases}\frac{\left(p^{\prime}{ }_{l, i}-p^{\mathrm{min}}\right)}{\left(1-p^{\mathrm{min}}\right)}, & \text { if } j=l \\ \frac{p_{j, i}^{{ }^{\prime}}}{\left(1-p^{\mathrm{min}}\right)}, & \text { otherwise }\end{cases}
$$

To verify that $P_{i}^{\prime \prime}$ satisfies (6), substituting $P_{i}^{\prime \prime}$ in (6), we obtain

$$
\begin{aligned}
\sum_{j=0}^{n} p_{j, i}^{\prime \prime} & =\frac{p_{l, i}^{\prime}-p^{\min }}{1-p^{\min }}+\sum_{j=0, j \neq l}^{n} \frac{p_{j, i}^{\prime}}{1-p^{\min }} \\
& =\frac{-p^{\min }}{1-p^{\min }}+\frac{\sum_{j=0}^{n} p_{j, i}^{\prime}}{1-p^{\min }}=1
\end{aligned}
$$

To verify the security condition in Definition 2, for $k=1,2, \ldots, t-1$, substituting $P_{i}^{\prime \prime}$ in (8), we obtain

$$
\begin{align*}
\sum_{j=0}^{n} & \left(p_{j, 0}^{\prime \prime}-p_{j, 1}^{\prime \prime}\right)\left(\frac{n}{n}\right)^{k} \\
= & \left(\frac{p_{l, 0}^{\prime}-p^{\min }}{1-p^{\min }}-\frac{p_{l, 1}^{\prime}-p^{\min }}{1-p^{\min }}\right)\left(\frac{l}{n}\right)^{k} \\
& +\sum_{j=0, j \neq l}^{n}\left(\frac{p_{j, 0}^{\prime}}{1-p^{\min }}-\frac{p_{j, 1}^{\prime}}{1-p^{\min }}\right)\left(\frac{j}{n}\right)^{k} \\
= & \left(1-p^{\min }\right)^{-1} \sum_{j=0}^{n}\left(p_{j, 0}^{\prime}-p_{j, 1}^{\prime}\right)\left(\frac{j}{n}\right)^{k} \\
= & \left(1-p^{\min }\right)^{-1} \times 0=0 \tag{14}
\end{align*}
$$

The result of (14) equal to 0 is obtained by the using result of (12). To verify the contrast condition in Definition 2, substituting $P_{i}^{\prime \prime}$ in (10), we obtain

$$
\begin{align*}
\widehat{\alpha}^{\prime \prime} & =\sum_{j=0}^{n}\left(p_{j, 0}^{\prime \prime}-p_{j, 1}^{\prime \prime}\right)\left(\frac{j}{n}\right)^{t} \\
& =\left(1-p^{\min }\right)^{-1} \sum_{j=0}^{n}\left(p_{j, 0}^{\prime}-p_{j, 1}^{\prime}\right)\left(\frac{j}{n}\right)^{t} \\
& >\sum_{j=0}^{n}\left(p_{j, 0}^{\prime}-p_{j, 1}^{\prime}\right)\left(\frac{j}{n}\right)^{t}=\widehat{\alpha}^{\prime} \tag{15}
\end{align*}
$$

The contrast $\widehat{\alpha}^{\prime \prime}$ is larger than $\widehat{\alpha}^{\prime}$ since $\left(1-p^{\text {min }}\right)^{-1}$ is larger than 1 . Hence we obtain that $P_{i}^{\prime \prime}$ has better contrast than $P_{i}^{\prime}$, which incurs contradiction.

A similar result of Lemma 2 for finite $n$ had been pointed out by previous work [12]. Lemma 2 gives an immediate mapping from $\Delta P$ to $P_{0}$ and $P_{1}$ which is described as

$$
\begin{align*}
\left(p_{j, 0}, p_{j, 1}\right) & = \begin{cases}\left(\Delta p_{j}, 0\right), & \text { if } \Delta p_{j}>0 \\
\left(0,-\Delta p_{j}\right), & \text { if } \Delta p_{j}<0 \\
(0,0), & \text { if } \Delta p_{j}=0\end{cases}  \tag{16}\\
& \sum_{j=0}^{n}\left|\Delta p_{j}\right|=2 \tag{17}
\end{align*}
$$

Lemma 3: For $(t, \infty)$ OptPrVC scheme, there is a $\Delta P$ consisting of $t+1$ nonzero values and $n-t$ zeros.

Proof: By (8), (9), (10), and (17), a linear program to achieve the optimal contrast is described as

$$
\begin{equation*}
\text { Maximize : } \quad \alpha^{\prime}=\sum_{j=0}^{n} \Delta p_{j}\left(\frac{j}{n}\right)^{t} \tag{18}
\end{equation*}
$$

Subject to : $\quad \sum_{j=0}^{n}\left|\Delta p_{j}\right|=2$

$$
\begin{align*}
\sum_{j=0}^{n} \Delta p_{j} & =0  \tag{20}\\
\sum_{j=0}^{n} \Delta p_{j}\left(\frac{j}{n}\right)^{k} & =0 \quad \forall k=1,2, \ldots, t-1
\end{align*}
$$

In order to remove the absolute symbols in (19), we divide the feasible region into $2^{n+1}$ subregions according to the sign of each $\Delta p_{j}$. Then the above linear program is divided into $2^{n+1}$ subproblems, expressed as

$$
\begin{equation*}
\text { Maximize : } \quad \alpha^{\prime}=\sum_{j=0}^{n} \Delta p_{j}\left(\frac{j}{n}\right)^{t} \tag{22}
\end{equation*}
$$

$$
\begin{equation*}
\text { Subject to : } \sum_{j=0}^{n}\left(g_{j} \Delta p_{j}\right)=2 \tag{23}
\end{equation*}
$$

$$
\begin{equation*}
\sum_{j=0}^{n} \Delta p_{j}\left(\frac{j}{n}\right)^{k}=0 \forall k=1,2, \ldots, t-1 \tag{24}
\end{equation*}
$$

$$
\begin{equation*}
\left(g_{j} \Delta p_{j}\right) \geq 0 \forall j=0,1, \ldots, n \tag{26}
\end{equation*}
$$

where $g_{j}= \pm 1$ is the sign of $\Delta p_{j}$ in each subregion. There are one formula (23), one formula (24), and $t-1$ formulas (25)
as the constraints of the linear program, and those formulas are linearly independent as long as the subregion is not empty, and the maximal $\alpha^{\prime}$ lies at the boundary sets (26). Thus, $\Delta P$ consists of $t+1$ nonzero values and $n-t$ zeros.

Since the $t+1$ nonzero terms in $\Delta P$ represent the critical values, those nonzero terms are recorded by two vectors $X=$ $\left(x_{0}, x_{1}, \ldots, x_{t}\right)$ and $Y=\left(y_{0}, y_{1}, \ldots, y_{t}\right)$, where $\left|y_{i}\right|$ is the probability of using the memoryless sequence $E\left(x_{i}\right)$ to encode the secret pixel $s$ taken from the secret image $S$, with $y_{i}>0$ for $s=$ white and $y_{i}<0$ for $s=$ black. Without loss of generality, $X$ is arranged in increasing order (i.e., $x_{i}<x_{i+1}$ ). In other words, the $i$ th nonzero term $\Delta p_{j}$ is recorded in $x_{i}$ and $y_{i}$ such that $x_{i}=j / n$ and $y_{i}=\Delta p_{j}$. The equations of $x_{i}$ and $y_{i}$ are described as

$$
\begin{equation*}
0 \leq x_{0}<x_{1}<\cdots<x_{t} \leq 1 \tag{27}
\end{equation*}
$$

Example 2: For $H_{0}=(0,1,0,0,2)$ and $H_{1}=(2,0,0,1,0)$ obtained from Example 1, the $P_{0}, P_{1}, \Delta P, X$, and $Y$ are

$$
\begin{aligned}
P_{0} & =\left(0, \frac{1}{3}, 0,0, \frac{2}{3}\right), P_{1}=\left(\frac{2}{3}, 0,0, \frac{1}{3}, 0\right) \\
\Delta P & =\left(-\frac{2}{3}, \frac{2}{3}, 0,-\frac{1}{3}, \frac{1}{3}\right) \\
X & =\left(\frac{0}{4}, \frac{1}{4}, \frac{3}{4}, \frac{4}{4}\right), Y=\left(-\frac{2}{3}, \frac{2}{3},-\frac{1}{3}, \frac{1}{3}\right) .
\end{aligned}
$$

By (8), (9), and (10), the $(t+1)$ equations substituted with $X$ and $Y$ are expressed as

$$
\begin{gather*}
\sum_{j=0}^{t} y_{j} x_{j}^{k}=0 \quad \forall k=1,2, \ldots, t-1  \tag{28}\\
\sum_{j=0}^{t} y_{j}=0  \tag{29}\\
\alpha^{\prime}=\sum_{j=0}^{t} y_{j} x_{j}^{t} \tag{30}
\end{gather*}
$$

Then, (28)-(30) are expressed in the matrix form as

$$
\left[\begin{array}{cccc}
1 & 1 & \cdots & 1  \tag{31}\\
x_{0} & x_{1} & \cdots & x_{t} \\
\vdots & \vdots & \ddots & \vdots \\
x_{0}^{t} & x_{1}^{t} & \cdots & x_{t}^{t}
\end{array}\right]\left[\begin{array}{l}
y_{0} \\
y_{1} \\
\vdots \\
y_{t}
\end{array}\right]=\left[\begin{array}{l}
0 \\
0 \\
\vdots \\
\alpha^{\prime}
\end{array}\right] .
$$

The $(t+1) \times(t+1)$ matrix $M$ is the transpose of a Vandermonde matrix. By the property of $M^{-1}=\left(\left(M^{T}\right)^{-1}\right)^{\mathrm{T}}$, we use the inverse Vandermonde matrix to obtain $M^{-1}$. In fact, the coefficients of inverse Vandermonde matrix can be expressed as Lagrange polynomials[30]. We fill $M^{-1}$ with those coefficients to obtain

$$
\left[\begin{array}{l}
y_{0}  \tag{32}\\
y_{1} \\
\vdots \\
y_{t}
\end{array}\right]=\left[\begin{array}{ccc}
\bullet & \cdots & \left(\prod_{j=0, j \neq 0}^{t}\left(x_{0}-x_{j}\right)\right)^{-1} \\
\bullet & \cdots & \left(\prod_{j=0, j \neq 1}^{t}\left(x_{1}-x_{j}\right)\right)^{-1} \\
\vdots & \ddots & \vdots \\
\bullet & \cdots & \left(\prod_{j=0, j \neq t}^{t}\left(x_{t}-x_{j}\right)\right)^{-1}
\end{array}\right]\left[\begin{array}{l}
0 \\
0 \\
\vdots \\
\alpha^{\prime}
\end{array}\right]
$$

where - are don't-care terms, since those values will be multiplied by zeros in the later stage. Then $y_{i}$ is described as

$$
\begin{equation*}
y_{i}=\alpha^{\prime}\left(\prod_{j=0, j \neq i}^{t}\left(x_{i}-x_{j}\right)\right)^{-1} \tag{33}
\end{equation*}
$$

Lemma 4: For an $(t, \infty)$ OptPrVC scheme, if $t+i$ is even, then $y_{i}>0$; otherwise, if $t+i$ is odd, then $y_{i}<0$.

Proof: The result is easily verified by (33).
By Lemma 4, all values of $y_{i}$ are grouped into two sets $\left\{y_{i} \mid t+\right.$ $i \in$ Even integer $\}$ and $\left\{y_{i} \mid t+i \in\right.$ Odd integer $\}$; and the sums of all elements in both set are 1 and -1 , respectively. By (33), their mathematical formulations are expressed as

$$
\begin{gather*}
\sum_{j=0}^{\lfloor t / 2\rfloor} y_{2 i}=\alpha^{\prime} \sum_{i=0}^{\lfloor t / 2\rfloor}\left(\prod_{j=0, j \neq i}^{t}\left(x_{2 i}-x_{j}\right)\right)^{-1}=(-1)^{t}  \tag{34}\\
\sum_{j=0}^{\lfloor(t-1) / 2\rfloor} y_{2 i+1}=\alpha^{\prime} \sum_{i=0}^{\lfloor(t-1) / 2\rfloor}\left(\prod_{j=0, j \neq i}^{t}\left(x_{2 i+1}-x_{j}\right)\right)^{-1} \\
=(-1)^{t+1} . \tag{35}
\end{gather*}
$$

By (34) and (35), $\alpha^{\prime}$ is expressed as

$$
\begin{align*}
\alpha^{\prime} & =(-1)^{t}\left[\sum_{i=0}^{\lfloor t / 2\rfloor}\left(\prod_{j=0, j \neq i}^{t}\left(x_{2 i}-x_{j}\right)\right)^{-1}\right]^{-1} \\
& =(-1)^{t+1}\left[\sum_{i=0}^{\lfloor(t-1) / 2\rfloor}\left(\prod_{j=0, j \neq i}^{t}\left(x_{2 i+1}-x_{j}\right)\right)^{-1}\right]^{-1} . \tag{36}
\end{align*}
$$

Lemma 5: For the $X$ of a $(t, \infty)$ OptPrVC scheme, $x_{0}=0$ and $x_{t}=1$.

Proof: Suppose $\left(X^{\prime}, Y^{\prime}\right)$ is a $(t, \infty)$ OptPrVC scheme but $x_{0}^{\prime}>0$ or $x_{t}^{\prime}<1$. We generate a new $\left(X^{\prime \prime}, Y^{\prime \prime}\right)$ where each term $x_{j}^{\prime \prime}$ in $X$ is

$$
x_{j}^{\prime \prime}= \begin{cases}0, & \text { if } j=0 \\ 1, & \text { if } j=t \\ x_{j}^{\prime}, & \text { otherwise }\end{cases}
$$

Then the contrast of $\left(X^{\prime \prime}, Y^{\prime \prime}\right)$ is better than $\left(X^{\prime}, Y^{\prime}\right)$ by substituting $X^{\prime}$ and $X^{\prime \prime}$ in (36).

Finally, the optimization problem $L(t)$ is derived by combining (27) and (36) with Lemma 5. The optimization problem $L(t)$ is described as

$$
\begin{align*}
\text { Minimize }: \alpha^{\prime-1}=(-1)^{t} \sum_{i=0}^{\lfloor t / 2\rfloor}\left(\prod_{j=0, j \neq i}^{t}\left(x_{2 i}-x_{j}\right)\right)^{-1} \\
=(-1)^{t+1} \sum_{i=0}^{\lfloor(t-1) / 2\rfloor}\left(\prod_{j=0, j \neq i}^{t}\left(x_{2 i+1}-x_{j}\right)\right)^{-1} \tag{37}
\end{align*}
$$

Subject to : $0=x_{0}<x_{1}<\cdots<x_{t}=1$.
The target function is the reciprocal of $\alpha^{\prime}$. Krause and Simon [20] proved that the contrast of $(k, n) \mathrm{VC}$ scheme is no less than $4^{-(t-1)}$, so the outcome of $L(t)$ is expected to be $4^{(t-1)}$.

TABLE I
$(X, Y)$ OF $\operatorname{Some}(t, \infty)$ OptPrVC SChEMES FOR $2 \leq t \leq 6$

| $i$ | $X$ and $Y$ | $\alpha^{\prime}$ |
| :--- | :--- | :--- |
| 2 | $Y=(1 / 2,-1,1 / 2)$ | $4^{-1}$ |
|  | $X=(0,1 / 2,1)$ |  |
| 3 | $Y=(-1 / 3,2 / 3,-2 / 3,1 / 3)$ | $4^{-2}$ |
|  | $X=(0,1 / 4,3 / 4,1)$ | $4^{-3}$ |
| 4 | $Y=(1 / 4,-1 / 2,1 / 2,-1 / 2,1 / 4)$ |  |
|  | $X=(0,(2-\sqrt{2}) / 4,1 / 2,(2+\sqrt{2}) / 4,1)$ | $4^{-4}$ |
| 5 | $Y=(-1 / 5,2 / 5,-2 / 5,2 / 5,-2 / 5,1 / 5)$ |  |
|  | $X=(0,(3-\sqrt{5}) / 8,(5-\sqrt{5}) / 8$, | $4^{-5}$ |
|  | $(3+\sqrt{5}) / 8,(5+\sqrt{5}) / 8,1)$ |  |
| 6 | $Y=(1 / 6,-1 / 3,1 / 3,-1 / 3,1 / 3,-1 / 3,1 / 6)$ |  |
|  | $X=(0,(2-\sqrt{3}) / 4,1 / 4,1 / 2,3 / 4,(2+\sqrt{3}) / 4,1)$ |  |

We use Maple to solve $L(t)$ for $2 \leq t \leq 6$, and the results are listed in Table I.

## E. Encoding Algorithm

For a given value of $t$, the transparencies can be continuously generated with the $(t, \infty)$ OptPrVC scheme. However, practical applications require the algorithm to terminate within finite steps. To meet the requirement, a finite number $n^{\prime}$ is used to specify the number of transparencies in the algorithm. The algorithm requires $(X, Y)$, obtained by solving $L(t)$ or by looking up Table I. The outputs of Algorithm 1 are $n^{\prime}$ transparencies $T_{1}, T_{2}, \ldots, T_{n}^{\prime}$ and an index table $Z$, where $Z[w, h]$ is the index of the used memoryless sequence $E\left(x_{Z[w, h]}\right)$ to encode the secret pixel $s[w, h]$.

```
Algorithm 1. The algorithm of \((t, \infty)\) OptPrVC scheme
Input: A binary secret image \(S\), two positive integers \(t, n^{\prime}\),
and two vectors \((X, Y)\).
Output: \(n^{\prime}\) transparencies \(T_{1}, T_{2}, \ldots, T_{n}^{\prime}\); an index table \(Z\).
    for each pixel \(s[w, h]\) in \(S\) do
        if \(s[w, h]=\) white then
            Generate an integer \(z \in\{t-2 k \mid k=0,1, \ldots,\lfloor t / 2\rfloor\}\)
            and \(P(z=t-2 k)=y_{t-2 k}\).
        else
            Generate an integer \(z \in\{t-1-2 k \mid k=\)
            \(0,1, \ldots,\lfloor(t-1) / 2\rfloor\}\) and \(P(z=t-1-2 k)=\)
            \(-y_{t-1-2 k}\).
        end if
        \(Z[w, h]=z\).
        for \(k=1\) to \(n^{\prime}\) do
            Assign randomly \(T_{k}[w, h]\) to 0 or 1 where
            \(P\left(T_{k}[w, h]=0\right)=x_{z}\).
        end for
    end for
```

In the first round, we use Algorithm 1 to generate $n^{\prime}$ transparencies and $Z$. If we need not to generate more transparencies in the future, $Z$ is not required and discarded. Otherwise, $Z$ has to be stored in a safe place, and we can generate more transparencies $T_{1}^{\prime}, T_{2}^{\prime}, \ldots, T_{n^{\prime \prime}}^{\prime}$ by utilizing $Z$.


Fig. 1. Binary secret image.


(b)

Fig. 2. Results of $(2, \infty)$ OptPrVC scheme. (a) $T_{1}$. (b) $T_{2}$. (c) $T_{1} \oplus T_{2}$.

Algorithm 2. The algorithm of $(t, \infty) \mathrm{OptPrVC}$ scheme by the index table $Z$

Input: An index table $Z$, a positive integer $n^{\prime \prime}$, and a vector $X$.
Output: $n^{\prime \prime}$ transparencies $T_{1}^{\prime}, T_{2}^{\prime}, \ldots, T_{n^{\prime \prime}}^{\prime}$.
for each $z[w, h]$ in $Z$ do
for $k=1$ to $n^{\prime \prime}$ do
Assign randomly $T_{k}^{\prime}[w, h]$ to 0 or 1 where
$P\left(T_{k}^{\prime}[w, h]=0\right)=x_{z}$.
end for
end for

## III. EXperimental Results

Three experiments are performed for $t=2,3$, and 4. Fig. 1 shows the binary secret image $S$. Fig. 2 shows the first experiment for $t=2$. Fig. 2(a)-(b) shows two generated transparencies $T_{1}$ and $T_{2}$, and the stacking result $T_{1} \oplus T_{2}$ is shown in Fig. 2(c). We observe that the characters on the stacking result are clear. Based on Lemma 1, we verify the security and contrast of the three experiments. We use $T_{i_{1}}, T_{i_{2}}, \ldots, T_{i_{k}}$ to indicate any $k$ generated transparencies with the proposed scheme, and $t_{i_{j}}$ is a pixel at $T_{i_{j}}$ and the position corresponds to the secret pixel $s$.

For a single transparency $T_{i_{1}}$, the probabilities of a white secret pixel appearing zero and a black secret pixel appearing zero are

$$
\begin{aligned}
& P\left(t_{i_{1}}=0 \mid s=\text { white }\right)=0 \times 1 / 2+1 \times 1 / 2=1 / 2 \\
& P\left(t_{i_{1}}=0 \mid s=\text { black }\right)=1 \times 1 / 2=1 / 2
\end{aligned}
$$

Therefore, the white pixels and black pixels cannot be distinguished by observing only one transparency. In the case where two transparencies are stacked $T_{i_{1}} \oplus T_{i_{2}}$, the contrast


Fig. 3. Results of $(3, \infty)$ OptPrVC scheme. (a) $T_{1}$. (b) $T_{2}$. (c) $T_{3}$. (d) $T_{1} \oplus T_{2}$. (e) $T_{1} \oplus T_{3}$. (f) $T_{2} \oplus T_{3}$. (g) $T_{1} \oplus T_{2} \oplus T_{3}$.
verification is
$P\left(t_{i_{1}} \oplus t_{i_{2}}=0 \mid s=\right.$ white $)=0 \times(1 / 2)^{2}+1 \times(1 / 2)^{2}=1 / 2$, $P\left(t_{i_{1}} \oplus t_{i_{2}}=0 \mid s=\right.$ black $)=1 \times(1 / 2)^{2}=1 / 4$.

Thus, the contrast is $\alpha^{\prime}=1 / 2-1 / 4=1 / 4$.
Fig. 3 shows the second experiment for $t=3$. Fig. 3(a)-(c) show three generated transparencies $T_{1}, T_{2}$, and $T_{3}$, and Fig. 3(d)-(f) show the results of stacking any two transparencies $T_{1} \oplus T_{2}, T_{1} \oplus T_{3}$, and $T_{2} \oplus T_{3}$. Fig. $3(\mathrm{~g})$ is the result of stacking the three transparencies $T_{1} \oplus T_{2} \oplus T_{3}$, and we can see contours of the secret on stacking result. The verifications of security and contrast are described as follows.

For a single transparency $T_{i_{1}}$, the security verification is

$$
\begin{aligned}
& P\left(t_{i_{1}}=0 \mid s=\text { white }\right)=2 / 3 \times 1 / 4+1 / 3 \times 1=1 / 2 \\
& P\left(t_{i_{1}}=0 \mid s=\text { black }\right)=1 / 3 \times 0+2 / 3 \times 3 / 4=1 / 2
\end{aligned}
$$

In the case where arbitrary two transparencies are stacked $T_{i_{1}} \oplus$ $T_{i_{2}}$, the security verification is
$P\left(t_{i_{1}} \oplus t_{i_{2}}=0 \mid s=\right.$ white $)=2 / 3 \times(1 / 4)^{2}+1 / 3 \times 1^{2}=3 / 8$, $P\left(t_{i_{1}} \oplus t_{i_{2}}=0 \mid s=\right.$ black $)=1 / 3 \times 0^{2}+2 / 3 \times(3 / 4)^{2}=3 / 8$.

In the case where arbitrary three transparencies are stacked $T_{i_{1}} \oplus$ $T_{i_{2}} \oplus T_{i_{3}}$, the contrast verification is
$P\left(t_{i_{1}} \oplus t_{i_{2}} \oplus t_{i_{3}}=0 \mid s=\right.$ white $)=\frac{2}{3} \times\left(\frac{1}{4}\right)^{3}+\frac{1}{3} \times 1^{3}=\frac{11}{32}$,
$P\left(t_{i_{1}} \oplus t_{i_{2}} \oplus t_{i_{3}}=0 \mid s=\right.$ black $)=\frac{1}{3} \times 0^{3}+\frac{2}{3} \times\left(\frac{3}{4}\right)^{3}=\frac{9}{32}$.
Thus, the contrast is $\alpha^{\prime}=11 / 32-9 / 32=1 / 16$.
Fig. 4 shows the third experiment for $t=4$. Fig. 4(a)-(d) show four generated transparencies $T_{1}, T_{2}, T_{3}$, and $T_{4}$. It is noted that we skip the stacking results of arbitrary two or three transparencies due to the large number of combinations. Fig. 4(e) shows the stacking result $T_{1} \oplus T_{2} \oplus T_{3} \oplus T_{4}$, where


Fig. 4. Results of $(4, \infty)$ OptPrVC scheme. (a) $T_{1}$. (b) $T_{2}$. (c) $T_{3}$. (d) $T_{4}$. (e) $T_{1} \oplus T_{2} \oplus T_{3} \oplus T_{4}$.
the characters on the stacking result are barely visible. The verifications of security and contrast are described as follows.

For a single transparency $T_{i_{1}}$, the security verification is

$$
\begin{aligned}
& P\left(t_{i_{1}}=0 \mid s=\text { white }\right)=0 \times \frac{1}{4}+\frac{1}{2} \times \frac{1}{2}+\frac{1}{4} \times 1=\frac{1}{2} \\
& P\left(t_{i_{1}}=0 \mid s=\text { black }\right)=\frac{1}{2} \times \frac{(2-\sqrt{2})}{4}+\frac{1}{2} \times \frac{(2+\sqrt{2})}{4}=\frac{1}{2} .
\end{aligned}
$$

In the case where two transparencies are stacked $T_{i_{1}} \oplus T_{i_{2}}$, the security verification is

$$
\begin{aligned}
& P\left(t_{i_{1}} \oplus t_{i_{2}}=0 \mid s=\text { white }\right) \\
& =1 / 4 \times 0^{2}+1 / 2 \times(1 / 2)^{2}+1 / 4 \times 1^{2}=3 / 8 \\
& P\left(t_{i_{1}} \oplus t_{i_{2}}=0 \mid s=\text { black }\right) \\
& =1 / 2 \times((2-\sqrt{2}) / 4))^{2}+1 / 2 \times((2+\sqrt{2}) / 4)^{2}=3 / 8
\end{aligned}
$$

In the case where three transparencies are stacked $T_{i_{1}} \oplus T_{i_{2}} \oplus$ $T_{i_{3}}$, the security verification is

$$
\begin{aligned}
& P\left(t_{i_{1}} \oplus t_{i_{2}} \oplus t_{i_{3}}=0 \mid s=\text { white }\right) \\
& =1 / 4 \times 0^{3}+1 / 2 \times(1 / 2)^{3}+1 / 4 \times 1^{3}=5 / 16 \\
& P\left(t_{i_{1}} \oplus t_{i_{2}} \oplus t_{i_{3}}=0 \mid s=\text { black }\right) \\
& =1 / 2 \times\left(((2-\sqrt{2}) / 4)^{3}+1 / 2 \times((2+\sqrt{2}) / 4)^{3}=5 / 16\right.
\end{aligned}
$$

In the case where four transparencies are stacked $T_{i_{1}} \oplus T_{i_{2}} \oplus$ $T_{i_{3}} \oplus T_{i_{4}}$, the contrast verification is

$$
\begin{aligned}
& P\left(t_{i_{1}} \oplus t_{i_{2}} \oplus t_{i_{3}} \oplus t_{i_{4}}=0 \mid s=\text { white }\right) \\
& =1 / 4 \times 0^{4}+1 / 2 \times(1 / 2)^{4}+1 / 4 \times 1^{4}=9 / 32 \\
& P\left(t_{i_{1}} \oplus t_{i_{2}} \oplus t_{i 3} \oplus t_{i_{4}}=0 \mid s=\text { black }\right) \\
& =1 / 2 \times((2-\sqrt{2}) / 4)^{4}+1 / 2 \times((2+\sqrt{2}) / 4)^{4}=17 / 64
\end{aligned}
$$

Thus, the contrast is $9 / 32-17 / 64=1 / 64$.

## IV. Discussions and Applications

## A. Addition and Deletion of Users in the Dynamic User Group

Since the proposed scheme allows dynamic changes of users in the user group, the operations to add and delete users are elaborated in this section. For the addition of $n^{\prime \prime}$ new participants to

TABLE II
Optimal Contrast $(t, n)$ VC Schemes for $2 \leq t \leq 4$

| $\operatorname{tnn}$ | 2 | 3 | 4 | 5 | 6 | 7 | 8 | $\ldots$ | 50 | $\ldots$ | 100 | $\ldots$ | $\infty$ |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| 2 | $1 / 2$ | $1 / 3$ | $1 / 3$ | $3 / 10$ | $3 / 10$ | $2 / 7$ | $2 / 7$ | $25 / 98$ | $25 / 99$ |  | $1 / 4$ |  |  |
| 3 |  | $1 / 4$ | $1 / 6$ | $1 / 8$ | $1 / 10$ | $1 / 10$ | $2 / 21$ | $13 / 196$ |  | $625 / 9702$ |  | $1 / 16$ |  |
| 4 |  |  | $1 / 8$ | $1 / 15$ | $1 / 18$ | $3 / 70$ | $3 / 80$ | $1161 / 65800$ | $425 / 25608$ | $1 / 64$ |  |  |  |

the user group with $n^{\prime}$ original participants, the transparencies need to be able to accommodate the new $n=n^{\prime}+n^{\prime \prime}$ users. In the case where the $n^{\prime}$ original transparencies are produced through the traditional $\left(t, n^{\prime}\right) \mathrm{VC}$ scheme, the $n^{\prime}$ original transparencies need to be eliminated, and then the new set of $n^{\prime}+n^{\prime \prime}$ transparencies need to be generated through the $\left(t, n^{\prime}+n^{\prime \prime}\right)$ VC scheme. The frequent regeneration and redistribution of the whole set of transparencies consume huge computing and communication resources for a dynamically changing user group, and may lead to potential security risks if certain original transparencies are not discarded completely. For example in the case of $t=2$, an original participant holding one of original $\left(2, n^{\prime}\right)$ VC transparencies and obtaining one of the new $\left(2, n^{\prime}+n^{\prime \prime}\right)$ VC transparencies might potentially be able to decode the secret by stacking the two transparencies. In the applications with a known maximal number of participants $n^{\text {max }}$, the better strategy is to apply the $\left(t, n^{\max }\right) \mathrm{VC}$ scheme to generate $n^{\max }$ transparencies; and each authorized participant is assigned a transparency. By applying the $\left(t, n^{\max }\right)$ VC scheme, the addition and departure of users can be accommodated without frequent regeneration and redistributions of transparencies. However, in many applications, the $n^{\max }$ is unknown and applying a larger number as $n^{\text {max }}$ is a feasible option to avoid the overflow of users over the $n^{\text {max }}$. The proposed $(t, \infty)$ OptPrVC scheme accommodates the addition and departure of users without the problem of user overflow. In addition, with very large $n^{\max }$, the contrast of the $(t, \infty)$ OptPrVC scheme is similar to the $\left(t, n^{\max }\right)$ VC scheme addressed in Section IV-B.

For the deletion of users, we suppose that $n^{\prime \prime}$ out of $n^{\prime}\left(n^{\prime \prime}<\right.$ $n^{\prime}$ ) participants are to leave the user group. If the contrast is to be optimized, the $n^{\prime}$ original transparencies need to be eliminated, and the new $n^{\prime}-n^{\prime \prime}$ transparencies need to be generated through the $\left(t, n^{\prime}-n^{\prime \prime}\right)$ VC scheme; the process consumes computing and communication resources for a frequently changing user group. Another strategy is to retrieve the transparencies from the departing participants; those retrieved transparencies can be assigned to the new participants. For the proposed $(t, \infty)$ OptPrVC scheme to accommodate the scenarios of departing users, the departing users can simply discard the transparencies, and the new transparencies for new users can simply be generated through Algorithm 2; the retrieval of the transparencies from the departing users is not required. The authentication and security of the retrieved transparencies present a more challenging problem and relevant studies can be found in [11] and [33].

## B. Contrast Comparisons of the Proposed Scheme With Traditional ( $t, n$ ) VC Schemes

In general, the contrast of the $(t, \infty)$ OptPrVC scheme is lower than certain $(t, n)$ VC schemes with a finite value of $n$. Nevertheless, if $n$ is very large, the $(t, \infty) \operatorname{OptPrVC}$ has some


Fig. 5. Experimental result of $(2,100)$ PrVC. (a) $T_{1}$. (b) $T_{2}$. (c) $T_{1} \oplus T_{2}$.
advantages over the traditional $(t, n) \mathrm{VC}$ scheme. In this case, the contrast of using the $(t, n) \mathrm{VC}$ scheme is similar to the $(t, \infty)$ OptPrVC scheme. The $(t, n)$ VC schemes with optimal contrast for $2 \leq t \leq 4$ and various values of $n$ are listed in Table II, which is duplicated for reference from [19], except for $n=\infty$. For $(t, n)=(2,100)$, the contrast is $25 / 99 \approx 0.2525$ which is very closes to $1 / 4=0.25$. In addition, For $(t, n)=$ $(3,100)$, the contrast is $625 / 9702 \approx 0.064419$ which is also very closes to $1 / 16=0.0625$. Thus, the stacking results of $(t, \infty) \mathrm{OptPrVC}$ scheme and $(t, n)$ VC scheme are very close to each other for large $n$. Fig. 5 is an example of the $(2,100)$ VC scheme based on probabilistic model with contrast 25/99. We can see similar qualities in Figs. 5(c) and 2(c). However, the traditional $(t, n)$ VC scheme requires larger amounts of computations and memory space to construct basis matrices. A simple analysis is discussed as follows.
For two $n \times m$ basis matrices $B_{0}$ and $B_{1}$ of the traditional $(t, n) \mathrm{VC}$ scheme, the $2 m n$ memory space is needed to store $B_{0}$ and $B_{1}$. To encode a secret pixel, first, the arithmetic sequence $1,2, \ldots, m$ is permutated to obtain $d_{1}, d_{2}, \ldots, d_{m}$ with $\Theta(m)$ operations; second, each row of $B_{0}$ or $B_{1}$ is duplicated to the corresponding transparency, and the elements of each row are arranged by following the orders $d_{1}, d_{2}, \ldots, d_{m}$ with $\Theta(n m)$ I/O operations. Thus, the encoding includes $\Theta(m)$ computations and $\Theta(n m)$ I/O operations, and the $\Theta(n m)$ dominates the traditional $(t, n)$ VC scheme due to the essentiality of the duplicating operations during the production of transparencies. It is noted that the encoding complexity is $\Theta(m)$ if the I/O operation is excluded from the complexity measure. On the other hand, the proposed probabilistic $(t, n)$ VC scheme requires the values of two vectors $X$ and $Y$, whose lengths are both $t+1$. To encode a secret pixel, an $n$-element memoryless sequence is generated, which takes $\Theta(n)$ operations. The value of $m$ depends on the construction method of $B_{0}$ and $B_{1}$. Blundo et al. [17] proved that a traditional $(2, n)$ VC scheme with $\alpha>1 / 4$ holds the property $m \geq n-1$. Thus, the space complexity and the time complexity of the $(2, n)$ VC scheme are both $\Theta\left(n^{2}\right)$. On the other hand, the proposed scheme takes $\Theta(t=2)=\Theta(1)$ space and $\Theta(n)$ operations. Thus, the proposed method substantially reduces the computations and memory space for large $n$.

## C. Comparisons of $L(t)$ With Two Linear Programs

Hofmeister et al. [19] provide an elegant linear program to calculate the optimal contrast of $(t, n) \mathrm{VC}$ scheme. In addition, the basis matrices can be constructed by the solution of the linear

TABLE III
Comparisons of $L(t)$ With Other Optimization Equations

|  | Hofmeister et al. [19] | Bose and Mukerjee [14] | The proposed $L(t)$ |
| :--- | :--- | :--- | :--- |
| Goal | Calculate the optimal <br> contrast of $(t, \quad n)$ VC <br> schemes. | Calculate the optimal <br> contrast of $\quad t, \quad n) \quad$ VC <br> schemes. | Calculate the optimal <br> contrast of $\quad t, \quad \infty) \quad$ VC <br> schemes. |
| The contrast <br> definition | Naor and Shamir [1] | Naor and Shamir [1] | Ito et al. [2] |
| The model | Basis matrices with the <br> totally symmetric scheme | Basis matrices with the <br> totally symmetric scheme | Probabilistic model |
| The number of <br> Tariables | $2 n+2$ | $n-t$ | $t-1$ |

program. We restate the linear program $M(t, n)$ as follows:

$$
\begin{equation*}
\text { Minimize }: \sum_{j=0}^{n-t} \frac{\binom{n-t}{j}}{\binom{n}{j}} \times\left(p_{j, 0}-p_{j, 1}\right) \tag{39}
\end{equation*}
$$

Subject to : $p_{j, 0}, p_{j, 1} \geq 0 \forall j=0,1, \ldots, n$

$$
\begin{gather*}
\sum_{j=0}^{n} p_{j, 0}=\sum_{j=0}^{n} p_{j, 1}=1  \tag{40}\\
\sum_{j=l}^{n-t+l+1} \frac{\binom{n-t+1}{j-l}}{\binom{n}{j}} \times\left(p_{j, 0}-p_{j, 1}\right)  \tag{41}\\ \tag{42}
\end{gather*}
$$

$M(t, n)$ requires $2 n+2$ variables $\left\{p_{j, 0} \mid j=0,1, \ldots, n\right\}$ to calculate the optimal contrast of $(t, n)$ VC schemes. Obviously, for $n \rightarrow \infty$, (39)-(42) cannot be explicitly formulated due to the infinite terms of the sigma summation, so $M(t, \infty)$ is not computationally feasible. The proposed $L(t)$ is a specialized optimization problem under $n \rightarrow \infty$, and (37) and (38) are well formulated capable of solving the $t-1$ variables $\left\{x_{1}, x_{2}, \ldots, x_{t-1}\right\}$. On the other hand, for a finite integer of $n, M(t, n)$ is more suitable than $L(t)$.

Moreover, Bose and Mukerjee [14] also provide the $L_{1}$-norm optimization equation to calculate the optimal contrast of the $(t, \infty)$ VC scheme; the equation $N(t, n)$ is restated as follows:

$$
\begin{equation*}
\delta=\min _{p_{n}, p_{n-1}, \ldots, p_{t+1}}\left\{\sum_{j=0}^{n}\left|q(j, n-t)-\sum_{k=0}^{n-t-1} q(j, k) p_{n-k}\right|\right\} \tag{43}
\end{equation*}
$$

where $q(j, k)=(-1)^{j-k}\binom{n}{j}\binom{j}{n-k}$, and the contrast is given as $\alpha^{\prime}=2 \delta^{-1}$. For $n \rightarrow \infty,(43)$ also cannot be explicitly formulated due to the infinite terms of the sigma summation. Table III lists the comparisons.

## D. Comparisons of the Proposed Method With RG Scheme

Table IV shows the comparisons of the $(t, \infty)$ OptPrVC scheme with RGs [26]-[29]. Rather than the basis matrices, the size of RG's transparencies is identical to the secret image, and this property is similar to the proposed $(t, \infty) \mathrm{OptPrVC}$ scheme. It is noted that certain studies [26]-[29] propose various encoding algorithms to achieve various stacking results and contrasts, and therefore, the contrast is not unique. The definition of the contrast follows (4). The proposed $(t, \infty)$ Opt$\operatorname{PrVC}$ scheme is a generalized scheme for arbitrary $t$, and the contrast is optimal under the definition of Ito et al. [2]. In order to give a detailed comparison, Table V shows the $(2, n) \mathrm{RG}$ proposed by Chen and Tsao [29]. As stated in [29], for a white secret pixel $s[w, h]=$ white, we have $P\left(T_{k}[w, h]=0\right)=1 / 2$,

TABLE IV
Comparisons of the $(t, \infty)$ OptPrVC Scheme With RGs

| Methods | Kafri and Keren [26] | Shyu [27] | Shyu [28] |
| :--- | :--- | :--- | :--- |
| $(t, n)$ | $(2,2)$ | $(2,2)$ | $(n, n)$ |
| Contrast | $2^{-1}$ or $4^{-1}$ | $2^{-1}$ or $4^{-1}$ | $2^{1-n},\left(2^{n}+1\right)^{-1}$ or $2^{-n}$ |


| Methods | Chen and Tsao [29] |  | The $(t, \infty)$ OptPrVC scheme |
| :--- | :--- | :--- | :--- |
| $(t, n)$ | $(n, n)$ | $(2, n)$ | $(t, \infty)$ |
| Contrast | $2^{1-n},\left(2^{n}+1\right)^{-1}$ <br> or $2^{-n}$ | $4^{-1}$ | $4^{1-t}$ |

TABLE V
Encoding Algorithm of $(2, n)$ RG Scheme Proposed by Chen and Tsao [29]

```
Input: A binary secret image S, and a positive integers n.
Output: }n\mathrm{ transparencies }\mp@subsup{T}{1}{},\mp@subsup{T}{2}{},\ldots,\mp@subsup{T}{n}{}\mathrm{ .
1 Create }\mp@subsup{T}{1}{}\mathrm{ by assigning each pixel to 0 or 1 and P(T, [w,
    h]=0)=1/2.
    for }k=2\mathrm{ to }n\mathrm{ do
        for each pixel s[w,h] in S do
            if s[w,h]= white then
                    Tk[w,h]= Tk-1 [w,h].
            end if
            if s[w,h]= black then
                    Assign randomly Tk[w,h] to 0 or 1 where P(T, [ [w,
                    h]=0)=1/2.
            end if
    10 end for
    llend for
```

and $P\left(T_{k}[w, h] \oplus T_{l}[w, h]=0\right)=1 / 2, \forall k \neq l$; and for $s[w, h]=$ black, we have $P\left(T_{k}[w, h]=0\right)=1 / 2$, and $P\left(T_{k}[w, h] \oplus T_{l}[w, h]=0\right)=1 / 4, \forall k \neq l$. The probabilities are identical to Algorithm 1 for $t=2$. Thus, the $(2, n) \mathrm{RG}$ [29] performs the same distribution in generating transparencies through the Algorithm 1 for $t=2$; and based on the observation, the proposed $(t, \infty) \mathrm{OptPrVC}$ scheme is also a generalized version of $(t, n) \mathrm{RG}$ for general $t$ and unlimited $n$.

## V. Conclusion

We have proposed a $(t, n)$ VC scheme with flexible value of $n$. From the practical perspective, the proposed scheme accommodates the dynamic changes of users without regenerating and redistributing the transparencies, which reduces computation and communication resources required in managing the dynamically changing user group. From the theoretical perspective, the scheme can be considered as the probabilistic model of $(t, n) \mathrm{VC}$ with unlimited $n$. Initially, the proposed scheme is based on basis matrices, but the basis matrices with infinite size cannot be constructed practically. Therefore, the probabilistic
model is adopted in the scheme. As the results listed in Table I, the proposed scheme also provides the alternate verification for the lower bound proved by Krause and Simon [20]. For $t=4$, the contrast $4^{-3}$ is very low so that the secret is visually insignificant. Therefore, in practical applications, the values of 2 or 3 for $t$ are empirically suggested for the proposed scheme.

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