

# AN ASYMMETRIC FINGERPRINTING SCHEME BASED ON TARDOS CODES

Ana Charpentier<sup>(a)</sup>, Caroline Fontaine<sup>(b)</sup>, Teddy Furon<sup>(a)</sup>, Ingemar Cox<sup>(c)</sup>

<sup>(a)</sup>INRIA-Rennes research center, Campus de Beaulieu, Rennes, France

<sup>(b)</sup>CNRS/Lab-STICC/CID, Télécom Bretagne/ITI, Brest, France

<sup>(c)</sup>University College London, Dpt. of Computer Science, London, United Kingdom

## ABSTRACT

Tardos codes are currently the state-of-the-art in the design of practical collusion-resistant fingerprinting codes. Tardos codes rely on a secret vector drawn from a publicly known probability distribution in order to generate each Buyer's fingerprint. For security purposes, this secret vector must not be revealed to the Buyers. To prevent an untrustworthy Provider forging a copy of a Work with an innocent Buyer's fingerprint, previous asymmetric fingerprinting algorithms enforce the idea of the Buyers generating their own fingerprint. Applying this concept to Tardos codes is challenging since the fingerprint must be based on this vector secret.

This paper provides the first solution for an asymmetric fingerprinting protocol dedicated to Tardos codes. The motivations come from a new attack, in which an untrustworthy Provider by modifying his secret vector frames an innocent Buyer.

**Index Terms**— Asymmetric fingerprinting, Tardos code

## 1. INTRODUCTION

This paper considers a problem arising in the fingerprinting of digital content. In this context, a fingerprint is a binary code that is inserted into the Work for the purpose of protecting it from unauthorized use, or, more precisely, for the purpose of identifying individuals responsible for unauthorized use of a Work. In such a scenario, it is assumed that two or more users may collude in order to try to hide their identities. In this case, it is further assumed that colluders cannot alter those bits of the code that are identical for all colluders. However, where bits differ across colluders, these bits may be assigned arbitrary values. A key problem is resistance to collusion, i.e. if a coalition of  $c$  users creates a pirated copy of the Work, its tampered fingerprint (i) should not implicate innocent users, and (ii) should identify at least one of the colluders.

This problem has received considerable attention since Boneh and Shaw [1] discussed the problem. They first introduced the concept of totally  $c$ -secure codes: if a coalition of  $c$  users colludes to produce a pirate copy of the Work, the tampered fingerprint is still guaranteed to identify at least one of the colluders, with no chance of framing an innocent.

Boneh and Shaw showed that totally  $c$ -secure binary codes do not exist for  $c > 1$ . They then introduced the concept of a  $c$ -secure code such that the probability of framing an innocent is lower than  $\epsilon$ . Unfortunately, the length of their codes,  $O(c^4 \log(\frac{n}{\epsilon}) \log(\frac{1}{\epsilon}))$ , where  $n$  is the number of users, was such as to make them impractical. Following Boneh and Shaw's paper, there has been considerable effort to design shorter codes.

In 2003, Tardos [2] proposed an efficient code construction that, for the first time, reduced the code length to the lower bound,  $O(c^2 \log(\frac{n}{\epsilon}))$ , thereby making such codes practical. Tardos codes are currently the state-of-the-art for collusion-resistant fingerprinting.

Several papers have considered a scenario where the Provider is untrustworthy. Thanks to the knowledge of a Buyer's fingerprint, the Provider creates a pirated copy of a Work, implicating this innocent Buyer. To prevent this, Pfitzman [3] first introduced the concept of asymmetric fingerprinting in which the Provider doesn't need to know the Buyer's fingerprint. The Buyer first commits to a secret (the fingerprint) that only he/she knows. The Buyer and Provider then follow a protocol which results in the Buyer receiving a copy of the Work with his/her secret fingerprint (and some additional information coming from the Provider) embedded within it. The Provider did not learn the Buyer's secret, and cannot therefore create a forgery. Unfortunately, in the case of Tardos codes, fingerprints must be drawn from a particular probability distribution depending on a secret vector only known to the Provider. Thus, previous asymmetric fingerprinting methods cannot be applied to Tardos codes.

The Tardos decoding is also vulnerable to an additional attack, in which the Provider does *not* need to create a forgery. Rather, given any unauthorized copy, i.e. a Work that does *not* contain the innocent Buyer's fingerprint, the Provider can alter its secret vector in order to accuse an arbitrary Buyer.

Our paper is organized as follows. We briefly introduce Tardos codes in Sec. 2. Sec. 3 describes the attack at the decoding side. In order to prevent both the Buyer and the Provider from cheating, Sec. 4 presents a new asymmetric protocol specific to Tardos codes. Sec. 5 then discusses practical aspects of the fingerprints embedding and accusation. We

finally discuss our solution in Sec. 6 before concluding.

## 2. THE TARDOS FINGERPRINTING CODE

For readers unfamiliar with Tardos codes, we now provide a brief introduction. Further details can be found in [4].

Let  $n$  denote the number of buyers, and  $m$  the length of the code. The fingerprints can then be arranged as a binary  $n \times m$  matrix  $\mathbf{X}$ , Buyer  $j$  being related to the binary fingerprint  $\mathbf{X}_j = (X_{j1}, X_{j2}, \dots, X_{jm})$ .

To generate this matrix,  $m$  real numbers  $p_i \in [t, 1 - t]$  are generated, each of them being randomly and independently drawn according to the probability density function  $f : [t, 1 - t] \rightarrow \mathbb{R}^+$  with  $f(z) = \kappa(t)(z(1 - z))^{-1/2}$  and  $\kappa(t)^{-1} = \int_t^{1-t} (z(1 - z))^{-1/2} dz$ . The parameter  $t \ll 1$  is referred to as the cutoff. We set  $\mathbf{p} = (p_1, \dots, p_m)$ . This vector  $\mathbf{p}$  is the secret key of the code only known by the Provider. Each element of the matrix  $\mathbf{X}$  is then independently randomly drawn, such that the probability that an element,  $X_{ji}$ , in the matrix is a one is given by  $\mathbb{P}(X_{ji} = 1) = p_i$ . The fingerprint is then embedded into the copy of the Work of the corresponding Buyer thanks to a watermarking technique.

If an unauthorized copy is found, its corresponding fingerprint,  $\mathbf{Y}$ , is decoded. Due to collusion, and possible distortions such as transcoding, the decoded fingerprint is unlikely to exactly equal one of the fingerprints in the matrix,  $\mathbf{X}$ . To determine if Buyer  $j$  is involved in the production of the unauthorized copy, a score, referred to as an accusation score,  $S_j$  is computed. If this score is greater than a given threshold  $Z$ , then Buyer  $j$  is considered to have colluded.

The scores are computed according to an accusation function  $g$ , reflecting the impact of the correlation between the sequence  $\mathbf{X}_j$ , associated with Buyer  $j$ , and the decoded sequence  $\mathbf{Y}$ :

$$S_j = G(\mathbf{Y}, \mathbf{X}_j, \mathbf{p}) = \sum_{i=1}^m g(Y_i, X_{ji}, p_i). \quad (1)$$

In the usual symmetric codes [4], function  $g$  is constrained (such that, for example, for an innocent, the expectation of the score is zero and its variance is  $m$ ), giving  $g(1, 1, p) = g(0, 0, 1 - p) = -g(0, 1, p) = -g(1, 0, 1 - p) = \sqrt{\frac{1-p}{p}}$ .

## 3. UNTRUSTWORTHY CONTENT PROVIDER

We now consider the case where the Provider is no longer trusted, and, as such, wishes to frame Buyer  $j$ . In such a scenario, we assume that the Provider has no prior access to an unauthorized copy, i.e. the Provider cannot insert a false fingerprint into the unauthorized copy, nor can he/she place a Buyer's copy on an unauthorized location. On receipt of an unauthorized copy, we further assume that the untrustworthy Provider to extracts the corresponding fingerprint present in

the unauthorized copy. We base this assumption on the hypothesis that the underlying watermarking algorithm comes from a technology provider and that the Provider doesn't master or has no access to this technology brick. Given the extracted fingerprint  $\mathbf{Y}$ , the Provider must now compare it to all known Buyers' fingerprints. This comparison is performed using Eq. (1). And it is here that the Provider can lie, since the probabilities,  $\mathbf{p}$ , are only known to the Provider.

An untrustworthy Provider can create a fake vector of probabilities,  $\hat{\mathbf{p}}$ , that implicates Buyer  $j$ . However, the distribution,  $f(p)$ , is publicly known, so the question becomes, can the Provider generate a  $\hat{\mathbf{p}}$  that (i) implicates Buyer  $j$ , and (ii) has an arbitrarily high probability of been drawn from the distribution  $f(p)$ ?

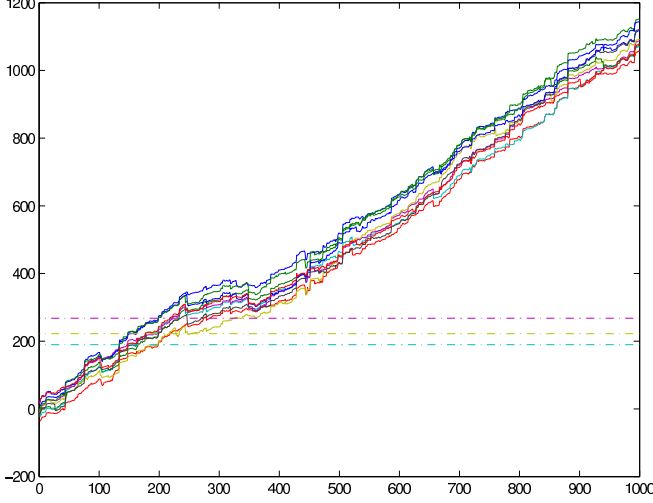
It is indeed extremely simple to do so. Let us focus on a column where  $p_i = p$  and  $Y_i = X_{j,i}$ . The true summand in Eq. (1) is  $g(1, 1, p)$  or  $g(0, 0, p)$  (with equal probability). Suppose that the content provider replaces the secret value  $p$  by a fake secret  $\hat{p}$  which is drawn independently according to  $f$ . On average, this summand takes the new value:

$$\Delta(t) = \int_t^{1-t} f(\hat{p}) \frac{g(1, 1, \hat{p}) + g(0, 0, \hat{p})}{2} d\hat{p} = \frac{1}{\pi} \ln \frac{1-t}{t}.$$

For a cutoff  $t = 1/900$  (recommended by G. Tardos to fight against 3 colluders), the numerical value is surprisingly high:  $\Delta(1/900) \approx 2.16$ . Suppose now that the content provider applies the same strategy on an index  $i$  where  $Y_i \neq X_{j,i}$ . Then the expectation is the opposite. However, in a Tardos code, even for an innocent Buyer  $j$ , the proportion  $\alpha$  of indices where symbols  $Y_i$  and  $X_{j,i}$  agree is above  $1/2$  for most of the collusion strategy. For instance, with an interleaving collusion attack,  $\alpha = 3/4$  whatever the collusion size  $c$ .

Based on this knowledge, we propose the following attack. The Provider computes the score for all Buyers, which, on average, equals 0 for innocent Buyers and  $2m/c\pi$  for the colluders [4]. The provider initializes  $\hat{\mathbf{p}} = \mathbf{p}$ . Then, he/she randomly selects a column  $i$  and randomly draws a fake secret  $\hat{p}_i \sim f$ . He/She re-computes the score of Buyer  $j$  with this fake secret and iterates selecting a different column until  $S_j$  is above the threshold  $Z$ . On average,  $m(c\pi\Delta(\alpha - 1/2))^{-1}$  secret values  $p_i$  need to be changed in this way, e.g. only 20% of the code length if the copy has been made using an interleaving attack.

Fig. 1 illustrates this attack for the case where the code length is  $m = 1000$  and the number of colluders is  $c = 3$ . The solid coloured lines depict the accusation scores of 10 randomly selected innocent buyers. We observe that after between 20-30% of the elements of  $\mathbf{p}$  have been altered, the accusation scores of the innocent Buyers exceed the *original* scores of the colluders. In fact, the colluders accusation scores also increase. However, we are not concerned with the highest score, but rather with any score exceeding the threshold. Thus, it is sufficient to raise the score of the innocent Buyer, even if this raises all other Buyers' scores as well.



**Fig. 1.** Accusation score as a function of the number of changed elements of the vector  $\mathbf{p}$  for the case where  $m = 1000$  and  $c = 3$ . The solid coloured lines show how the accusation score of 10 randomly selected innocent buyers increases as more of the elements are modified. The dotted horizontal lines show the original scores for the colluders before the modification.

Randomly selecting some  $p_i$ 's (independently from  $\mathbf{X}_j$  and  $\mathbf{Y}$ ) and re-drawing them according to the same law ensures that  $\hat{p}_i \sim f, \forall i$ . Therefore, a judge observing  $\hat{\mathbf{p}}$  cannot distinguish the forgery. For this reason, the judge might request to see the matrix  $\mathbf{X}$  to statistically test whether the elements of  $\mathbf{X}$  are drawn from the distribution  $\hat{\mathbf{p}}$ . In this case, the Provider can give a fake matrix  $\tilde{\mathbf{X}}$  where the columns whose  $p_i$  have been modified are re-drawn such that  $\mathbb{P}(X_{ki} = 1) = \hat{p}_i, \forall k \neq j$ . The only way to prevent this deception would be if the judge asked an innocent User  $k \neq j$  for his copy in order to verify the authenticity of  $\tilde{\mathbf{X}}$ . This latter step seems somewhat odd.

## 4. AN ASYMMETRIC TARDOS CODE CONSTRUCTION

In previous asymmetric fingerprinting schemes, it is up to the Buyer to generate his or her fingerprint. The Buyer then sends a commitment to the Provider, which prevents the Buyer from changing the fingerprint during the protocol. Unfortunately, this cannot be done with a Tardos code since the fingerprint must follow a given statistical distribution controlled by  $\mathbf{p}$ , and  $\mathbf{p}$  is only known to the Provider. This section proposes a solution to this problem, which consists of two phases. We first review its main building blocks.

### 4.1. Building blocks

There are two key building blocks to the proposed protocol. The first is a block involving encryption primitives, while the second involves double-blind random selection.

#### 4.1.1. Encryption Primitives

We need two cryptographic primitives: a regular symmetric cryptosystem  $\mathbf{E}$  (e.g. AES) and a commutative encryption scheme  $\mathbf{CE}$  (e.g. in [5, 6]). This latter primitive has the following property. For every key  $k_1$  and  $k_2$ , and for every message  $m$ , ciphering twice with  $k_1$  and then  $k_2$ , or  $k_2$  and then  $k_1$  leads to the same result:

$$\mathbf{CE}(k_1, \mathbf{CE}(k_2, m)) = \mathbf{CE}(k_2, \mathbf{CE}(k_1, m)). \quad (2)$$

#### 4.1.2. Pick a card, any card!

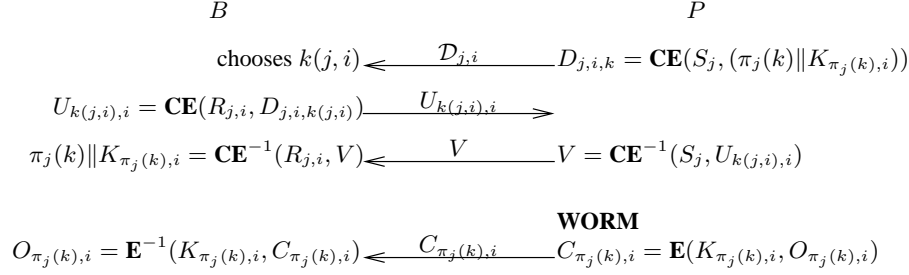
Here we introduce a double-blind random selection protocol between two entities A and B, based on [5]. Let  $\{O_k\}_{k=1}^N$  be a list of  $N$  objects offered by entity A. We now explain how entity B selects an item from this list without actually seeing the list and entity A does not know which item entity B picked.

Entity A chooses  $N$  secret keys for the  $\mathbf{E}$  cryptosystem called  $\{K_k\}_{k=1}^N$  and computes the cipher texts  $C_k = \mathbf{E}(K_k, O_k)$ . Entity A also chooses a secret key  $S$  for the  $\mathbf{CE}$  cryptosystem and encrypts the previous keys such that  $D_k = \mathbf{CE}(S, K_k)$ . He sends B the lists  $\mathcal{C} = \{C_k\}_{k=1}^N$  and  $\mathcal{D} = \{D_k\}_{k=1}^N$ . Entity B chooses an index  $k \in [N]$  (with the notation  $[N] = \{1, \dots, N\}$ ), a secret key  $R$  for the  $\mathbf{CE}$  cryptosystem, and sends A the cipher  $U_k = \mathbf{CE}(R, D_k)$ . Entity A decrypts  $U$  with his key  $S$  and sends B the result. Thanks to the commutative property, this message indeed equals  $\mathbf{CE}(R, K_k)$ , which B is able to decrypt thanks to his/her key  $R$ . The result is the key  $K_k$  which deciphers  $C_k$  onto the object  $O_k$ .

### 4.2. Phase 1: Generation of the fingerprint

We use the above protocol  $m$  times to generate the fingerprint of the  $j$ -th Buyer  $\mathbf{X}_j = (X_{j,1}, \dots, X_{j,m})$ . In this generation phase, A is the Provider, and B is Buyer  $j$ . The Provider generates a secret vector  $\mathbf{p}$  for a Tardos code. Each  $p_i$  is quantized such that  $p_i = L_i/N$  with  $L_i \in [N-1]$ .

For a given index  $i$ , the objects are the concatenation of a binary symbol and a text string. There are only two versions of an object in list  $\mathcal{C}_i$ . For  $L_i$  objects,  $O_{k,i} = (1 \parallel \text{ref}_{1,i})$ , and  $O_{k,i} = (0 \parallel \text{ref}_{0,i})$  for the  $N - L_i$  remaining ones. The use of the text strings  $\{\text{ref}_{X,i}\}$  depends on the content distribution mode as detailed in Sec. 5.1. The object  $O_{k,i}$  is encrypted with key  $K_{k,i}$  and stored in the list  $\mathcal{C}_i = \{C_{k,i}\}_{k=1}^N$ . There are thus as many different lists  $\mathcal{C}_i$  as the length  $m$  of the fingerprint. These lists are published in a public Write Once Read Many (WORM) directory [7] whose access is granted to all



**Fig. 2.** Generation of a fingerprint bit.

users. As explicitly stated in its name, nobody can modify or erase what has been put the first time in a WORM directory; beside, anybody can check its integrity.

On the contrary, the  $\mathcal{D}$ -lists are made specific to a given Buyer  $j$ . The provider picks a secret key  $S_j$  and a permutation  $\pi_j(\cdot)$  over  $[N]$ . This Buyer is proposed a list  $\mathcal{D}_{j,i}$  of  $N$  items as  $D_{j,i,k} = \mathbf{CE}(S_j, (\pi_j(k) || K_{\pi_j(k),i}))$ . Therefore, the lists  $\mathcal{C}_i$  are common for all users, whereas the lists  $\mathcal{D}_{j,i}$  are specific to Buyer  $j$ . We have introduced here a slight change wrt to protocol 4.1.2, i.e. the permutation  $\pi_j$  whose role is explained below. Buyer  $j$  chooses a secret  $R_{j,i}$  and one object in the list, say the  $k(j, i)$ -th object. He/she sends the corresponding ciphertext  $U_{k(j,i),i} = \mathbf{CE}(R_{j,i}, D_{j,i,k(j,i)})$  decrypted by the provider with  $S_j$  and sent back to the Buyer who, at the end, gets the index  $\text{ind}(j, i) = \pi_j(k(j, i))$  and the key  $K_{\text{ind}(j,i),i}$ , which grants him/her the access to the object  $O_{\text{ind}(j,i),i}$ , store encrypted in the WORM. It contains the symbol  $b_{\text{ind}(j,i),i}$ . This will be the value of the  $i$ -th bit of his/her fingerprint,  $X_{j,i} = b_{\text{ind}(j,i),i}$ , which equals '1' with probability  $p_i$ .

The provider keeps in a log file the values of  $S_j$  and  $U_{k(j,i),i}$ , the user keeps  $R_{j,i}$  in his/her records.

### 4.3. Phase 2: Disclosure of the halfword

For a more practical accusation process (see Sec. 5.2), the Provider will order Buyer  $j$  to reveal  $m_h < m$  bits of his fingerprint (phase 1 has been completed). This is done in order to build the so-called halfword [3] allowing the Provider to list a bunch of suspected users to be forwarded to the judge (See Sec. 5.2). The following facts must be enforced: Buyer  $j$  doesn't know which bits of his/her fingerprint are disclosed, and the Provider asks for the same bit indices to all the users.

Again, we propose to use the double-blind random selection protocol of Sec. 4.1.2. Now, Buyer  $j$  plays the role of A, and the Provider the role of B,  $N = m$ , and object  $O_i = (R_{i,j} || \text{alea}_{i,j})$ . These items are the  $m$  secret keys selected by Buyer  $j$  during Sec. 4.2 concatenated with random strings  $\text{alea}_{i,j}$  to be created by Buyer  $j$ . This alea finds its use during the personalization of the content (see Sec. 5.1). Following the protocol, the Provider selects  $m_h$  such object. The decryption of message  $U_{k(i,j),j}$  received during the construction phase of Sec. 4.2 thanks to the disclosure of the key

$R_{i,j}$  yields  $D_{i,j,k(i,j)}$  which in turn decrypted with key  $S_j$  provides the index of the selected object, otherwise the protocol stops. This prevents a colluder from denying the symbol of his fingerprint and from copying the symbol of an accomplice. At the end, the Provider learns which item was picked by Buyer  $j$  at index  $i$ . Therefore, he/she ends up with  $m_h$  couples  $(X_{j,i}, \text{alea}_{k(i,j),i})$  associated to a given Buyer  $j$ .

Thanks to this second part of our protocol, the Provider discloses  $m_h$  bits of the fingerprints without revealing any knowledge about the others, and Buyer  $j$  doesn't know which bits of his fingerprint were disclosed even if the Provider always chooses the same indices from a user to another. Of course, Buyer  $j$  refuses to follow this part of the protocol for more than  $m_h$  objects.

## 5. OTHER IMPLEMENTATION DETAILS

At this point, we have both introduced a new attack and a new asymmetric fingerprinting algorithm that are both specific to Tardos codes. The astute reader will be aware that our asymmetric fingerprinting protocol does not constitute a complete system. Here we briefly touch upon other implementation issues.

### 5.1. Watermarking

First, we need an algorithm so that the Provider sends the Buyer a copy of the Work with his/her fingerprint embedded, given the Provider does not know this fingerprint. There exist buyer-seller protocols for embedding a sequence  $\mathbf{X}_j$  into a content  $c_o$  without disclosing  $\mathbf{X}_j$  to the seller and  $c_o$  to the buyer. They are based on homomorphic encryption scheme and work with some specific implementations of spread spectrum [8] or Quantization Index Modulation watermarking [9]. The reader is directed to [8, 9] for further details. These methods can be adapted to embed the Tardos codes, but due to space limitations, a brief sketch of the adaptation of [9] is presented hereafter.

We adapt the secure embedding proposed in the last cited work as follows. Let  $\mathbf{c}_i^{(0)} = (c_{i,1}^{(0)}, \dots, c_{i,Q}^{(0)})$  be the  $Q$  quantized components (like pixels, DCT coefficients, portion of



streams etc) of the  $i$ -th content block watermarked with symbol ‘0’ (resp.  $\mathbf{c}_i^{(1)}$  with symbol ‘1’). Denote  $\mathbf{d}_i = \mathbf{c}_i^{(1)} - \mathbf{c}_i^{(0)}$ . Assume as in [9, Sect. 5], an additive homomorphic and probabilistic encryption  $E[\cdot]$  such as the Pallier cryptosystem. Buyer  $j$  has a pair of public/private keys  $(pk_j, sk_j)$  and sends  $(E_{pk_j}[X_{j,1}], \dots, E_{pk_j}[X_{j,m}])$ . The provider sends him/her the ciphers

$$E_{pk_j}[c_{i,\ell}^{(0)}] \cdot E_{pk_j}[X_{j,i}]^{d_{i,\ell}}, \forall (i, \ell) \in [m] \times [Q].$$

Thanks to the homomorphism, Buyer  $j$  decrypts this with  $sk_j$  into  $c_{i,\ell}^{(0)}$  if  $X_{j,i} = 0$ ,  $c_{i,\ell}^{(1)}$  if  $X_{j,i} = 1$ . Since  $X_{j,i}$  is constant for the  $Q$  components of the  $i$ -th block, a lot of bandwidth and computer power will be saved with a composite signal representation as detailed in [9, Sect. 3.2.2].

A crucial step in these buyer-seller protocols is to prove to the seller that what is sent by the Buyer is indeed the encryption of bits, and moreover bits of the Buyer’s fingerprint. To do so usually involves complex zero-knowledge subprotocols [8, 9]. We believe we can avoid this complexity by taking advantage of the fact that the Provider already knows some bits of the fingerprint  $\mathbf{X}_j$ , i.e. those belonging to the halfword (see Sect. 4.3), and the Buyers do not know the indices of these bits. Therefore, in  $m_v$  random indices of the halfword, the Provider asks the Buyer  $j$  to open his/her commitment. For one such index  $i_v$ , Buyer  $j$  reveals the random value  $r_{i_v}$  of the probabilistic Pallier encryption (with the notation of [9]). The Provider computes  $g^{X_{j,i_v}} h^{r_{i_v}} \bmod N$  and verifies it equals the  $i_v$ -th cipher, which Buyer  $j$  pretended to be  $E_{pk_j}[X_{j,i_v}]$ .

One drawback of this simple verification scheme is that the Buyer discovers  $m_v$  indices of the halfword. This may give rise to more elaborated collusion attacks. For example, Buyer  $j$ , as a colluder, could try to enforce  $Y_{i_v} \neq X_{j,i_v}$  when attempting to forge a pirated copy. Further discussion of this is beyond the scope of this paper.

This approach may also introduce a threat to the Buyer. An untrustworthy Provider can ask to open the commitments of non-halfword bits in order to disclose bits he/she is not supposed to know. For this reason, the Provider needs to send  $\text{alea}_{k(i_v,j),i_v}$  as defined in Sec. 4.3 to show Buyer  $j$  that his/her verification occurs on a halfword bit.

## 5.2. The accusation procedure

When an unauthorized copy is found, the Provider decodes the watermark and extracts the sequence  $\mathbf{Y}$  from the pirated content. The Provider computes the halfscores by applying Eq. (1) only on the halfwords. This produces a list of suspects, e.g. those users whose score is above a threshold, or those users with the highest scores.

Of course, this list cannot be trusted, since the Provider may be untrustworthy. The list is therefore sent to a third party, referred to as the Judge, who first verifies the computation of the halfscores. If different values are found, the

Provider is black-listed. Otherwise, the Judge computes the scores of the full fingerprint.

To do so, the Judge needs the secret  $\mathbf{p}$ : he/she asks the Provider for the keys  $\{K_{k,i}\}, \forall (k, i) \in [N] \times [m]$  and thereby obtains from the WORM all the objects  $\{O_{k,i}\}$ , and therefore the true values of  $(p_1, \dots, p_m)$ . The Judge must also request suspected Buyer  $j$  for the keys  $R_{j,i}$  in order to decrypt the messages  $U_{k(j,i),i}$  in  $D_{i,j,k(i,j)}$  which reveal which object Buyer  $j$  picked during the  $i$ -th round of Sec. 4.2 and whence  $X_{j,i}$ . Finally, the Judge accuses the user whose score over the full length fingerprint is above a given threshold (related to a probability of false alarm).

## 5.3. Security

Suppose first that the Provider is honest and denote by  $c$  the collusion size. A reliable tracing capability on the halfwords is needed to avoid false alarms. Therefore, as proven by G. Tardos,  $m_h = O(c^2 \log n \epsilon^{-1})$ , where  $\epsilon$  is the probability of suspecting some innocent Buyers. Moreover, successful collusions are avoided if there are secret values such that  $p_i < c^{-1}$  or  $p_i > 1 - c^{-1}$  (see [10]). Therefore,  $N$  should be sufficiently big, around a hundred, to resist against collusion of size of the order of ten. During the generation of the fingerprint in Sec. 4.2, permutation  $\pi_j(\cdot)$  makes sure that Buyer  $j$  randomly picks up a bit ‘1’ with probability  $p_i = L_i/N$  as needed in the Tardos code. In particular, a colluder cannot benefit from the discoveries made by his accomplices.

We now analyze why colluders would cheat during the watermarking of their version of the Work described in Sec. 5.1. By comparing their fingerprints, they see indices where they all have the same symbols, be it ‘0’ or ‘1’. As explained in the introduction, they won’t be able to alter those bits in the tampered fingerprint except if they cheat during the watermarking: If their fingerprint bits at index  $i$  all equal ‘1’, one of them must pretend he/she has a ‘0’ in this position. If they succeed to do so for all these positions, they will be able to forge a pirated copy with a null fingerprint for instance.

How many times do the colluders need to cheat? With probability  $p_i^c$  (resp.  $(1 - p_i)^c$ ), they all have bit ‘1’ (resp. ‘0’) at index  $i$ . Thus, there are on average  $m_c(c) = m \int_0^{1-c} (p^c + (1 - p)^c) f(p) dp$  such indices. The Provider asks for a bit verification with probability  $m_v/m_h$ . The probability of a successful attack for a collusion of size  $c$  is therefore  $(1 - m_v/m_h)^{m_c(c)}$ . Our numerical simulations have shown that  $m_v$  shouldn’t be more than 50 bits for typical code length and collusion size below a hundred. Thus,  $m_v$  is well below  $m_h$ .

Suppose now that the Provider is dishonest. The fact that the  $m$  lists  $C_i, \forall i \in [m]$  are public and not modifiable prevents the Provider from altering them for a specific Buyer in order to frame him/her afterwards. Moreover, it will raise the Judge’s suspicion if the empirical distribution of the  $p_i$  is not close to the pdf  $f$ . Yet, biases can be introduced on the probabilities for the symbols of the colluders’ fingerprint

only if there is a coalition between them and the untrustworthy Provider. For instance, the Provider can choose a permutation such that by selecting the first item (resp. the last one) in the list  $\mathcal{D}_{j,i}$  an accomplice colluder is sure to pick up a symbol '1' (resp. '0'). This ruins the tracing property of the code, but this does not allow the Provider to frame an innocent. First, it is guaranteed that  $\mathbf{p}$  used in Eq. 1 is the one which generated the code. Second, the Provider and his accomplices colluders must ignore a significant part of the fingerprints of innocent Buyers. To this end,  $m - m_h$  must also be in order of  $O(c^2 \log n \epsilon^{-1})$ . If this holds, the Judge is able to take a reliable decision while discarding the half-word part of the fingerprint. Consequently,  $m \approx 2m_h$ , our protocol has doubled the typical code length, which is still in  $O(c^2 \log n \epsilon^{-1})$ .

## 6. DISCUSSION AND SUMMARY

Tardos codes are currently the state-of-the-art in collusion-resistant fingerprinting. However, the previous asymmetric fingerprint protocols cannot be applied to this particular construction. There are mainly two difficulties. First, the Buyer has to generate his/her secret fingerprint but according to vector  $\mathbf{p}$ , which is kept secret by the Provider. Second, the vector  $\mathbf{p}$  used in the accusation process must be the same as the one which generated the fingerprints.

We have proposed a new asymmetric fingerprinting protocol dedicated to Tardos codes. We believe that this is the first such protocol, and that it is practically efficient.

The construction of the fingerprints and their embedding within pieces of Work do not need a trusted third party.

Note, however, that during the accusation stage, a trusted third party is necessary like in any asymmetric fingerprinting scheme we are aware of. Further work is needed to determine if such a third-party can be eliminated. In particular, we anticipate that some form of secure multi-party computation can be applied.

Other extensions to this work include (i) non-binary Tardos codes, and (ii) implementation on compliant consumer devices such as Blu-Ray players. We also plan to develop this as part of future work.

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