

# More Efficient Secure Outsourcing Methods for Bilinear Maps

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**Abstract**—Bilinear maps are popular cryptographic primitives which have been commonly used in various modern cryptographic protocols. However, the cost of computation for bilinear maps is expensive because of their realization using variants of Weil and Tate pairings of elliptic curves. Due to increasing availability of cloud computing services, devices with limited computational resources can outsource this heavy computation to more powerful external servers. Currently, the checkability probability of the most efficient outsourcing algorithm is  $1/2$  and the overall computation requires 4 point addition in the preimage and 3 multiplications in the image of the bilinear map under the one-malicious version of a two-untrusted-program model. In this paper, we propose two efficient new algorithms which decrease not only the memory requirement but also the overall communication overhead.

**Index Terms**—Outsourcing computation, Bilinear maps, Secure delegation, Secure Cloud Computing.

## I. INTRODUCTION

THE improvements in the cloud computing services result in variety of new security and privacy challenges. Many cryptographic mechanisms involving complex computations such as bilinear maps are proposed to overcome these challenges [1], [2], [3]. Since speeding up the computation of bilinear maps is crucial in real-life applications, many schemes are suggested to reduce the computational cost of pairing computation [4], [5], [2], [6], [7], [8]. Especially, Hess introduced a general framework encompassing different types of pairing functions giving optimum numbers of computation steps [9]. However, these computations are still infeasible or unaffordable for resource constrained devices including mobile phones, tablets, smart or RFID cards.

Since Hohenberger and Lysyanskaya stated the question of how a computationally limited device may outsource its computation to another, potentially malicious, but much more computationally powerful device [10], it has been studied extensively. Outsourcing the complex computations to external powerful devices dates back to Matsumoto, in which the RSA signature generation problem is considered [11]. More specifically, Chevallier-Mames *et al.* proposed a protocol enabling a computationally limited device to outsource the computation of bilinear maps into a more resourceful device [12]. However, this delegation process brought new concerns. Firstly, the external device should learn nothing about the secrets. Also, the computationally-limited device should be able to check

whether the external device computed correctly, at least with certain probability. These two concerns can be eliminated by masking the secret values with the cost of some extra computations before sending to the external server, and then removing the masking values together with a way of validating the outsourced computation.

Besides the efficiency constraints, secrecy is the main objective of the security model, since the input and output pair of a client is used for cryptographic purposes. Henceforth, outsource mechanisms surely follow a security model in which the client (the energy limited trusted device that needs to delegate the computation) does not trust the servers (which perform the needed computations). Thus, in the security model, it is assumed that the client is honest but the servers are untrusted. Furthermore, checkability, validation of the computation processes, should be also addressed.

As simulated in Figure 1, outsource computation protocols may utilize one or more servers. Based on the number of servers utilized Tian *et al.* classified them as follows [13]:

- One-Untrusted Program (OUP): One malicious server performs the computation.
- One-Malicious version of a Two-Untrusted Program (OMTUP): Two untrusted servers perform the computation but only one of them may behave maliciously.
- Two-Untrusted Program (TUP): Two untrusted servers perform the computation and both of them may behave maliciously, but they do not maliciously collude.

Following the work of Chen *et al.*, [14], Tian *et al.* proposed two algorithms [13] in the OMTUP setting. First algorithm achieves less computational complexity and the second one improves the checkability at the cost of some additional computations. These outsource protocols are composed of both offline and online computation steps. In the offline phase, the client prepares the necessary values. During the online phase, the client creates masked values based on the precomputed offline values and requests the bilinear map computation.

In this paper, we propose two new algorithms following the steps of Chen *et al.* and Tian *et al.*'s work. We further analyze the protocols under the OMTUP assumption and reduce not only the computational complexity of the offline computations, but also the memory needed to store the values resulting from the offline computations together with the communication overhead. While doing so, we do not increase the computation costs that need to be handled by the client.

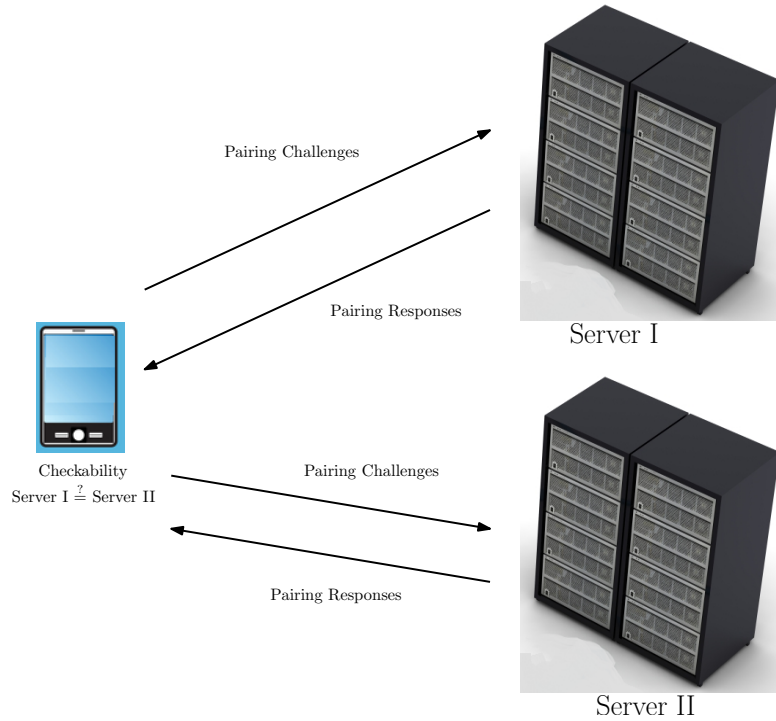


Fig. 1. Outsourcing Bilinear Maps with Two Untrusted Cloud Servers

#### A. Related work

Weil and Tate pairings are firstly used as cryptanalytic tools for reducing the discrete logarithm problem (DLP) on some elliptic curves to DLP on finite fields [15]. Later, Boneh *et al.* and Joux constructed new cryptographic protocols based on bilinear maps [1], [2], [3]. Reducing the computational cost of bilinear maps are suggested in [4], [5], [2], [6], [7], [8], [16], [9].

First protocol for secure outsourcing of elliptic curve pairings were proposed by Chevallier-Mames *et al.* [12]. The algorithm assumes the OUP setting and it is 1-checkable. However, the algorithm requires expensive computations, namely multiple membership test operation which is equivalent to an exponentiation over the finite field and inversion on the exponents. Later on, under the same OUP assumption, Kang *et al.* [17] and Canard *et al.* [18] improved the computational complexity results. However, the solutions were not feasible since exponentiation, membership test, and inversion were still required. Tsang *et al.* made a taxonomy for pairing based computations and constructed a batch pairing delegation mechanism [19]. Chow *et al.* studied server aided signature verification [20].

Chen *et al.* broke the paradigm by utilizing two servers under the OMTUP assumption and by performing some computations during an idle time of the resource-limited device [14]. As a result, this outsourcing mechanism of bilinear maps was the first one which does not depend on the membership test operations and exponentiations over the finite field. Additionally, this scheme decreased the online computations on the client side. The user had to perform only 5 point additions in  $\mathbb{G}_1$  and  $\mathbb{G}_2$ , and 4 multiplications in  $\mathbb{G}_3$ , where  $e : \mathbb{G}_1 \times \mathbb{G}_2 \rightarrow \mathbb{G}_3$  is the underlying bilinear map. Later, Tian

*et al.* proposed a more efficient algorithm [13] and reduced the online computation phase to 4 point additions in  $\mathbb{G}_1$  and  $\mathbb{G}_2$ , and 3 multiplications in  $\mathbb{G}_3$ .

#### B. Our contributions

In this paper, we propose two efficient new algorithms for secure outsourcing of bilinear maps. Compared to the state of the art algorithms (especially Tian *et al.*'s [13]), our algorithms need less offline computations, less memory, and less queries to the servers. In order to manage that, different from the previous studies, we use negation of an input value (it is almost for free since it is located over the elliptic curve), and we also send the same checking computation to both servers. Since it is assumed that the two servers do not collude, we reduce the computation costs without affecting the checkability of the system. The second algorithm may for instance be utilized in signature verification applications, in which we evade from at least one multiplication. For both propositions, we also provide the security model following exactly the lines of the security model of Hohenberger and Lysyanskaya [10]. We conclude the paper by comparing the efficiency of the system with the very recent work of Tian *et al.* [13].

#### C. Roadmap

In Section II, we give the security definitions for the outsourcing algorithm. Then, we present some background and preliminaries that will be needed throughout the manuscript, and we propose our two main algorithms together with their security analysis in Section III. Next, in Section IV, we analyze complexity of our new algorithms and compare it to the complexity of the best known algorithm [13]. Finally, we conclude the paper in Section V.

## II. SECURITY MODEL

Chen *et al.* [14] and Tian *et al.* [13] follow the security model proposed by Hohenberger and Lysyanskaya [10]. We remark especially that we also follow exactly their security model [10].

*Definition 1:* An algorithm is said to obey the outsource input/output specification if it takes five inputs, and produces three outputs. The first three inputs are generated by an honest party, and are classified by how much the adversary  $A = (E, U')$  knows about them. The first input is called the honest, secret input, which is unknown to both  $E$  and  $U'$ ; the second is called the honest, protected input, which may be known by  $E$ , but is protected from  $U'$ ; and the third is called the honest, unprotected input, which may be known by both  $E$  and  $U'$ . In addition, there are two adversarially-chosen inputs generated by the environment  $E$ : the adversarial, protected input, which is known to  $E$ , but protected from  $U'$ ; and the the adversarial.

*Definition 2:* Let  $\text{Alg}(\cdot, \cdot, \cdot, \cdot, \cdot)$  be an algorithm with outsource-IO. A pair of algorithms  $(T, U)$  is said to be an outsource-secure implementation of an algorithm  $\text{Alg}$  if: **Correctness.**  $T^U$  is a correct implementation of  $\text{Alg}$ .

**Security.** For all probabilistic polynomial-time adversaries  $A = (E, U')$ , there exist probabilistic expected polynomial-time simulators  $(S_1, S_2)$  such that the following pairs of random variables are computationally indistinguishable. Let us say that the honestly-generated inputs are chosen by a process  $I$ .

- **Pair One:**  $EVIEW_{real}^i \sim EVIEW_{ideal}^i$  (The external adversary,  $E$ , learns nothing.)
- The view that the adversarial environment  $E$  obtains by participating in the following REAL process:

$$\begin{aligned} EVIEW_{real}^i = \{ & (istate^i, x_{hs}^i, x_{hp}^i, x_{hu}^i) \leftarrow \\ & I(1^k, istate^{i-1}); \\ & (estate^i, j^i, x_{ap}^i, x_{au}^i, stop^i) \leftarrow \\ & E(1^k, EVIEW_{real}^{i-1}, x_{hp}^i, x_{hu}^i); \\ & (tstate^i, ustate^i, y_s^i, y_p^i, y_u^i) \leftarrow \\ & T^{U'}(ustate^{i-1})(tstate^{i-1}, x_{hs}^i, x_{hp}^i, x_{hu}^i, x_{ap}^i, x_{au}^i) : \\ & (estate^i, y_p^i, y_u^i) \} \end{aligned}$$

$EVIEW_{real} = EVIEW_{real}^i$  if  $stop^i = TRUE$ .

The real process proceeds in rounds. In round  $i$ , the honest (secret, protected, and unprotected) inputs  $(x_{hs}^i, x_{hp}^i, x_{hu}^i)$  are picked using an honest, stateful process  $I$  to which the environment does not have access. Then the environment, based on its view from the last round, chooses (0) the value of its  $estate_i$  variable as a way of remembering what it did next time it is invoked; (1) which previously generated honest inputs  $(x_{hs}^i, x_{hp}^i, x_{hu}^i)$  to give to  $T^{U'}$  (note that the environment can specify the index  $j^i$  of these inputs, but not their values); (2) the adversarial, protected input  $x_{ap}^i$ ; (3) the adversarial, unprotected input  $x_{au}^i$ ; (4) the Boolean variable  $stop^i$  that determines whether  $round_i$  is the last round in this process. Next, the algorithm  $T^{U'}$  is run on the inputs  $(tstate^{i-1}, x_{hs}^i, x_{hp}^i, x_{hu}^i, x_{ap}^i, x_{au}^i)$ , where  $tstate^{i-1}$

is  $T$ 's previously saved state, and produces a new state  $tstate^i$  for  $T$ , as well as the secret  $y_s^i$ , protected  $y_p^i$  and unprotected  $y_u^i$  outputs. The oracle  $U'$  is given its previously saved state,  $ustate^{i-1}$ , as input, and the current state of  $U'$  is saved in the variable  $ustate^i$ . The view of the real process in  $round_i$  consists of  $estate^i$ , and the values  $y_p^i$  and  $y_u^i$ . The overall view of the environment in the real process is just its view in the last round (i.e.,  $i$  for which  $stop^i = TRUE$ ).

- The IDEAL process:

$$\begin{aligned} EVIEW_{ideal}^i = \{ & (istate^i, x_{hs}^i, x_{hp}^i, x_{hu}^i) \leftarrow \\ & I(1^k, istate^{i-1}); \\ & (estate^i, j^i, x_{ap}^i, x_{au}^i, stop^i) \leftarrow \\ & E(1^k, EVIEW_{ideal}^{i-1}, x_{hp}^i, x_{hu}^i); \\ & (astate^i, y_s^i, y_p^i, y_u^i) \leftarrow \\ & \text{Alg}(astate^{i-1}, x_{hs}^i, x_{hp}^i, x_{hu}^i, x_{ap}^i, x_{au}^i); \\ & (sstate^i, ustate^i, Y_p^i, Y_u^i, replace^i) \leftarrow \\ & S_1^{U'}(ustate^{i-1})(sstate^{i-1}, x_{hp}^i, x_{hu}^i, x_{ap}^i, x_{au}^i, y_p^i, y_u^i); \\ & (z_p^i, z_u^i) = \text{replace}^i(Y_p^i, Y_u^i) + (1 - \text{replace}^i)(y_p^i, y_u^i) : \\ & (estate^i, z_p^i, z_u^i) \} \end{aligned}$$

$EVIEW_{ideal} = EVIEW_{ideal}^i$  if  $stop^i = TRUE$ .

The ideal process also proceeds in rounds. In the ideal process, we have a stateful simulator  $S_1$  who, shielded from the secret input  $x_{hs}$ , but given the non-secret outputs that  $\text{Alg}$  produces when run all the inputs for round  $i$ , decides to either output the values  $(y_p^i, y_u^i)$  generated by  $\text{Alg}$ , or replace them with some other values  $(Y_p^i, Y_u^i)$  (Notationally, this is captured by having the indicator variable  $replace^i$  be a bit that determines whether  $y_p^i$  will be replaced with  $Y_p^i$ .) In doing so, it is allowed to query the oracle  $U'$ ; moreover,  $U'$  saves its state as in the real experiment.

- **Pair Two:**  $UVIEW_{real} \sim UVIEW_{ideal}$  (The untrusted software,  $(U_1, U_2)$ , learns nothing.)
- The view that the untrusted software  $U'$  obtains by participating in the REAL process described in Pair One.  $UVIEW_{real} = ustate^i$  if  $stop^i = TRUE$ .
- The IDEAL process:

$$\begin{aligned} UVIEW_{ideal}^i = \{ & (istate^i, x_{hs}^i, x_{hp}^i, x_{hu}^i) \leftarrow \\ & I(1^k, istate^{i-1}); \\ & (estate^i, j^i, x_{ap}^i, x_{au}^i, stop^i) \leftarrow \\ & E(1^k, estate^{i-1}, x_{hp}^i, x_{hu}^i, y_p^{i-1}, y_u^{i-1}); \\ & (astate^i, y_s^i, y_p^i, y_u^i) \leftarrow \\ & \text{Alg}(astate^{i-1}, x_{hs}^i, x_{hp}^i, x_{hu}^i, x_{ap}^i, x_{au}^i); \\ & (sstate^i, ustate^i) \leftarrow \\ & S_2^{U'}(ustate^{i-1})(sstate^{i-1}, x_{hu}^i, x_{au}^i) : \\ & (ustate^i) \} \end{aligned}$$

$UVIEW_{ideal} = UVIEW_{ideal}^i$  if  $stop^i = TRUE$ .

In the ideal process, we have a stateful simulator  $S_2$  who, equipped with only the unprotected inputs  $(x_{hu}^i, x_{au}^i)$ , queries  $U'$ . As before,  $U'$  may maintain state.

In our security model we assume one-malicious version of a two-untrusted program (OMTUP) model. More concretely, there are two untrusted cloud servers in this model perform-

ing the outsourced computation, where only one of them is assumed to be malicious.

**Definition 3:** A pair of algorithms  $(T, U_1, U_2)$  are an  $\alpha$ -efficient implementation of an algorithm **Alg** if (1) they are an outsource-secure implementation of **Alg**, and (2)  $\forall$  inputs  $x$ , the running time of  $T$  is  $\leq$  an  $\alpha$ -multiplicative factor of the running time of **Alg**( $x$ ).

**Definition 4:** A pair of algorithms  $(T, U_1, U_2)$  are an  $\beta$ -checkable implementation of an algorithm **Alg** if (1) they are an outsource-secure implementation of **Alg**, and (2)  $\forall$  inputs  $x$ , if  $U'_i, i = 1, 2$  deviates from its advertised functionality during the execution of  $T^{(U'_1, U'_2)}(x)$ ,  $T$  will detect the error with probability  $\geq \beta$ .

**Definition 5:** A pair of algorithms  $(T, U_1, U_2)$  are an  $(\alpha, \beta)$ -outsource secure implementation of an algorithm **Alg** if they are both  $\alpha$ -efficient and  $\beta$ -checkable.

### III. ALGORITHMS FOR OUTSOURCING OF BILINEAR MAPS

#### A. Preliminaries: Bilinear Maps

Let  $(\mathbb{G}_1, +)$  and  $(\mathbb{G}_2, +)$  be two additive cyclic groups of order  $q$  with  $\mathbb{G}_1 = \langle Q \rangle$  and  $\mathbb{G}_2 = \langle P \rangle$ ,  $(\mathbb{G}_3, \cdot)$  be a multiplicative cyclic group of order  $q$ , where  $q$  is a prime number and  $0_{\mathbb{G}_1}, 0_{\mathbb{G}_2}$  and  $1_{\mathbb{G}_3}$  are the identity elements of the groups  $\mathbb{G}_1, \mathbb{G}_2$  and  $\mathbb{G}_3$ , respectively. Assume that Discrete Logarithm Problem (DLP) is hard in both  $\mathbb{G}_1$  and  $\mathbb{G}_2$  (i.e., given a random  $y \in \mathbb{G}_1$  (or  $\in \mathbb{G}_2$ ), it is computationally infeasible to find an integer  $x \in \mathbb{Z}$  such that  $y = g^x$ ). If it is clear from the context we write 0 for the identity elements of  $\mathbb{G}_1, \mathbb{G}_2$  and 1 for  $\mathbb{G}_3$ . A *bilinear map* is a map  $e: \mathbb{G}_1 \times \mathbb{G}_2 \rightarrow \mathbb{G}_3$  satisfying the following properties [15]:

- **Bilinearity:** For all  $P_1, Q_1 \in \mathbb{G}_1, P'_1, Q'_1 \in \mathbb{G}_2$ ,  $e$  is a group homomorphism in each component, i.e.
  - 1)  $e(P_1 + Q_1, P'_1) = e(P_1, P'_1) \cdot e(Q_1, P'_1)$ ,
  - 2)  $e(P_1, P'_1 + Q'_1) = e(P_1, P'_1) \cdot e(P_1, Q'_1)$ .
- **Non-degeneracy:**  $e$  is non-degenerate in each component, i.e.,
  - 1) For all  $P \in \mathbb{G}_1, P \neq 0$ , there is an element  $Q \in \mathbb{G}_2$  such that  $e(P, Q) \neq 1$ ,
  - 2) For all  $Q \in \mathbb{G}_2, Q \neq 0$ , there is an element  $P \in \mathbb{G}_1$  such that  $e(P, Q) \neq 1$ .
- **Computability:** There exists an algorithm which computes the bilinear map  $e$  efficiently.

#### B. Algorithm 1

1) **Precomputations:** Like all existing outsourcing algorithms, some precomputations are performed to speed up the proposed algorithms following the method of [13]. It includes a static table ST and a dynamic table DT. The values stored in the dynamic table are replaced while they are used, and then the table is reconstructed in an idle time of the device. We next describe the steps of the Rand1 algorithm to generate random group elements.

Rand1

- **Preprocessing Step:** Let  $P_1$  and  $P_2$  be generators of  $\mathbb{G}_1$  and  $\mathbb{G}_2$ , respectively. Generate  $n$  random elements  $\alpha_1, \dots, \alpha_n \in \mathbb{Z}/q\mathbb{Z}$ . For  $j = 1, \dots, n$  compute  $\beta_{j1} =$

$\alpha_j \cdot P_1$  and  $\beta_{j2} = \alpha_j \cdot P_2$ , and store the values of  $\alpha_j, \beta_{j1}$  and  $\beta_{j2}$  in ST. Compute  $e(P_1, P_2) \in \mathbb{G}_3$  and store it in ST.

- **Generation of Precomputed Values:** A new entry in DT is computed as follows: Generate randomly  $S \in \{1, \dots, n\}$  such that  $|S| = k$ . For each  $j \in S$ , select randomly  $K_j \in \{1, \dots, h-1\}$ , where  $h > 1$  is a small integer. Compute

$$x_1 \equiv \sum_{j \in S} \alpha_j K_j \pmod{q}.$$

If  $x_1 \equiv 0 \pmod{q}$ , start again. Otherwise, compute

$$x_1 \cdot P_1 \equiv \sum_{j \in S} K_j \cdot \beta_{j1} \pmod{q}.$$

Following the above procedure, compute similarly the elements  $(x_2, x_2 \cdot P_2), (x_3, x_3 \cdot P_1)$ , and  $(x_4, x_4 \cdot P_2)$ . Then compute

- 1)  $2x_1 \cdot P_1$ ,
- 2)  $-2x_2 \cdot P_2$ ,
- 3)  $e(P_1, P_2)^{2x_1 x_2}$ ,
- 4)  $e(P_1, P_2)^{x_3 x_4}$ .

The entry

$$(x_1 \cdot P_1, 2x_1 \cdot P_1, x_3 \cdot P_1, x_2 \cdot P_2, -2x_2 \cdot P_2, x_4 \cdot P_2, e(P_1, P_2)^{2x_1 x_2}, e(P_1, P_2)^{x_3 x_4})$$

is stored in DT. On each invocation of Rand1, an entry is returned and removed from DT. Further, a new set of values is used as fresh random values.

2) **Proposed Algorithm 1:** Our algorithm takes  $A \in \mathbb{G}_1, B \in \mathbb{G}_2$  as inputs and produces  $e(A, B)$  as output. In what follows,  $T$  denotes a trusted device with limited computation resources, and  $U_i(A, B) \rightarrow e(A, B), i \in \{1, 2\}$  denotes party  $U_i$  taking  $(A, B)$  as inputs and returning  $e(A, B)$  as output.

- **Initialization:**  $T$  calls Rand1 to get random values

$$(x_1 \cdot P_1, 2x_1 \cdot P_1, x_3 \cdot P_1, x_2 \cdot P_2, -2x_2 \cdot P_2, x_4 \cdot P_2, \lambda = e(P_1, P_2)^{2x_1 x_2}, e(P_1, P_2)^{x_3 x_4}).$$

- **Computation:** In random orders,  $T$  sends the following values to  $U_1$

- 1)  $U_1(A + 2x_1 \cdot P_1, -B - 2x_2 \cdot P_2) \rightarrow \alpha_1$ ,
- 2)  $U_1(x_3 \cdot P_1, x_4 \cdot P_2) \rightarrow \alpha'_1$ .

Similarly, in random orders,  $T$  sends the following values to  $U_2$

- 1)  $U_2(A + x_1 \cdot P_1, B + x_2 \cdot P_2) \rightarrow \alpha_2$ ,
- 2)  $U_2(x_3 \cdot P_1, x_4 \cdot P_2) \rightarrow \alpha'_2$ .

- **Recover:**  $T$  checks whether  $\alpha'_1 \stackrel{?}{=} \alpha'_2 \stackrel{?}{=} e(P_1, P_2)^{x_3 x_4}$ . If the verifications are successful then it computes

$$\beta = \alpha_1 \alpha_2^2 \lambda$$

and produces  $\beta$  as output. Otherwise, it rejects and gives an "Error".

### 3) Security Analysis:

**Theorem 6:** Under the OMTUP assumption, the algorithms  $(T, (U_1, U_2))$  of “Algorithm 1” are an outsource-secure implementation of a pairing evaluation, where the input  $(A, B)$  may be honest, secret, honest, protected, or adversarial, protected.

*Proof:* The correctness follows easily by bilinear property of  $e$ :

$$\begin{aligned} \beta &= \alpha_1 \cdot \alpha_2 \cdot \lambda \\ &= e(A + 2x_1 \cdot P_1, -B - 2x_2 \cdot P_2) \cdot \\ &e(A + x_1 \cdot P_1, B + x_2 \cdot P_2)^2 \cdot e(P_1, P_2)^{2x_1x_2} \\ &= e(A, B)^{-1} \cdot e(P_1, P_2)^{-4x_1x_2} \cdot e(A, B)^2 \cdot \\ &e(P_1, P_2)^{2x_1x_2} \cdot e(P_1, P_2)^{2x_1x_2} \\ &= e(A, B). \end{aligned}$$

We next prove the security of the algorithm. Let  $(E, U'_1, U'_2)$  be a PPT adversary that interacts with a PPT algorithm  $T$  in the OMTUP model.

**Pair One:**  $EVIEW_{real}^i \sim EVIEW_{ideal}^i$  (The external adversary,  $E$ , learns nothing.):

For a round  $i$ , if the input  $(A, B)$  is other than secret, honest, (i.e., honest, protected or adversarial, protected) the simulator  $S_1$  behaves as in the real round.  $S_1$  never requires to access the secret input, since there is none. So suppose that the input  $(A, B)$  is honest, secret. In that case, the simulator  $S_1$  behaves as follows: On receiving input in the  $i$ th round,  $S_1$  ignores it and instead make random queries to the servers  $U'_1, U'_2$ :

- $U'_1(x_1 \cdot P_1, x_2 \cdot P_2) \rightarrow \alpha_1$ ,
- $U'_1(x_3 \cdot P_1, x_4 \cdot P_2) \rightarrow \alpha'_1$ ,
- $U'_2(x_5 \cdot P_1, x_6 \cdot P_2) \rightarrow \alpha_2$ ,
- $U'_2(x_3 \cdot P_1, x_4 \cdot P_2) \rightarrow \alpha'_2$ .

After getting responses from  $U'_1$  and  $U'_2$ ,  $S_1$  checks:

- If  $\alpha'_1 \neq \alpha_2$  or  $e(x_3 \cdot P_1, X_4 \cdot P_2) \neq \alpha'_1$ ,  $S_1$  produces  $Y_p^i = \text{“Error”}$ ,  $Y_u^i = \emptyset$  and  $rep^i = 1$ .
- If all responses are correct,  $S_1$  sets  $Y_p^i = \emptyset$ ,  $Y_u^i = \emptyset$ , and  $rep^i = 0$ .
- Otherwise,  $S_1$  selects a random value  $s_r \in \mathbb{G}_3$  and sets  $Y_p^i = s_r$ ,  $Y_u^i = \emptyset$  and  $rep^i = 1$ .

For all cases,  $S_1$  saves the appropriate states.

The input distributions in the real and ideal experiments are computationally indistinguishable for  $U'_1$  and  $U'_2$ . The inputs to  $U'_1$  and  $U'_2$  are chosen uniformly at random in the ideal experiment. In a real experiment, each part of all queries that  $T$  makes to any one program in the computation step is independently re-randomized, and the re-randomization factors (i.e., outputs of Rand1) are either truly randomly generated by naive table-lookup approach or computationally indistinguishable from random by the assumption of the EBPV generator [21]. Now, there are three possibilities to consider.

- If  $U'_1$  and  $U'_2$  behave honestly in the round  $i$ ,  $S_1$  gives the correct output, using Alg, which is the same as the output of  $T^{U_1, U_2}$ .
- If one of  $(U'_1, U'_2)$  give an incorrect output in the  $i$ th round and it has been detected by both  $T$  and  $S_1$  with probability  $1/2$ , then it will result in an “Error”.

- Finally, we consider the case, where one of  $(U'_1, U'_2)$  give an incorrect output in the  $i$ th round and it is not caught with probability  $1/2$ . In the real experiment, the two outputs generated by  $(U'_1, U'_2)$  are multiplied together along with a random value  $\lambda$ .

Thus, a corrupted output looks random to the environment  $E$ , in the real experiment,  $S_1$  also simulates with a random value in  $\mathbb{G}_3$  as the output. So,  $EVIEW_{real}^i \sim EVIEW_{ideal}^i$  even when one of  $(U'_1, U'_2)$  is dishonest. Now, by the hybrid argument, we conclude that  $EVIEW_{real} \sim EVIEW_{ideal}$ .

**Pair Two:**  $UVIEW_{real} \sim UVIEW_{ideal}$  (The untrusted software,  $(U_1, U_2)$ , learns nothing.):

Here, regardless of the input type, the simulator  $S_2$  always behaves the same way. Upon receiving an input on round  $i$ ,  $S_2$  ignores it and instead makes four random queries to  $U'_1$  and  $U'_2$ . Then  $S_2$  saves its own state and the states of  $(U'_1, U'_2)$ .  $E$  can easily distinguish between these real and ideal experiments (output of the ideal experiment is never corrupted), but we want to show that  $E$  cannot share this information with  $(U'_1, U'_2)$ . This happens because in the  $i$ th round of the real experiment,  $T$  always re-randomizes the inputs to  $(U'_1, U'_2)$ , and in the ideal experiment  $S_2$  creates random, independent queries for  $(U'_1, U'_2)$ . So, for each round  $i$ , we have  $UVIEW_{real}^i \sim UVIEW_{ideal}^i$ . Then, by the hybrid argument, we get the desired result  $UVIEW_{real} \sim UVIEW_{ideal}$ .

**Theorem 7:** The algorithms  $(T, (U_1, U_2))$  of “Algorithm 1” are a  $(\mathcal{O}(\frac{1}{\log q}), 1/2)$ -outsource secure implementation of a pairing evaluation under the OMTUP assumption.

*Proof:* By the above theorem,  $U_1$  and  $U_2$  cannot distinguish a test query from a real query. Without loss of generality, assume that  $U_1$  is honest while  $U_2$  is dishonest (since we are under the OMTUP assumption). Thus,  $U_2$  fails with a probability  $1/2$ .

## C. Algorithm 2

### 1) Precomputations: Rand2

- **Preprocessing Step:** Generate  $n$  random elements  $\alpha_1, \dots, \alpha_n \in \mathbb{Z}_q$ . For  $j = 1, \dots, n$  compute  $\beta_{j_1} = \alpha_j \cdot P_1$  and  $\beta_{j_2} = \alpha_j \cdot P_2$ , and store the values of  $\alpha_j, \beta_{j_1}$  and  $\beta_{j_2}$  in a static table ST. Compute  $e(P_1, P_2)$  and store the value in ST.
- **Generation of Precomputed Values:** When a table DT needs a new entry, it is produced as follows. Randomly generate  $S \in \{1, \dots, n\}$  such that  $|S| = k$ . For each  $j \in S$ , randomly select  $K_j \in \{1, \dots, h-1\}$ , where  $h > 1$  is a small integer. Compute

$$x_1 \equiv \sum_{j \in S} \alpha_j K_j \pmod{q}.$$

If  $x_1 \equiv 0 \pmod{q}$ , start again. Otherwise, compute

$$x_1 P_1 \equiv \sum_{j \in S} K_j \cdot \beta_{j_1} \pmod{q}.$$

Following the above procedure, compute similarly the elements  $(x_2, x_2 \cdot P_2), (x_3, x_3 \cdot P_1), (x_4, x_4 \cdot P_1)$ , and  $(x_5, x_5 \cdot P_2)$ . Then compute

	Algorithm [13]	Algorithm 1	Algorithm 2
SM	3	2	1
ME	2	2	1
PA	5(k+h-3)	4(k+h-3)	5(k+h-3)

TABLE I  
COMPARISON OF PRECOMPUTATION

	Algorithm [13]	Algorithm 1	Algorithm 2
PA	4	4	4
FM	3	3	2

TABLE II  
COMPARISON OF CLIENT'S WORKLOAD

$$\begin{aligned} 1) & (x_2 - x_1^{-1}x_2x_3) \cdot P_2, \\ 2) & e(P_1, P_2)^{x_4x_5}. \end{aligned}$$

The entry

$$\begin{aligned} & (x_1 \cdot P_1, x_3 \cdot P_1, x_4 \cdot P_1, x_2 \cdot P_2, x_5 \cdot P_2, \\ & (x_2 - x_1^{-1}x_2x_3) \cdot P_2, e(P_1, P_2)^{x_4x_5}) \end{aligned}$$

is stored in DT. On each invocation of Rand2, an entry is returned and removed from DT. Further, a new set of values is used as fresh random values.

2) *Proposed Scheme 2*: Algorithm 2 takes  $A \in \mathbb{G}_1, B \in \mathbb{G}_2$  as inputs and produces  $e(A, B)$  as output. In what follows,  $T$  denotes a trusted device with limited computation resources, and  $U_i(A, B) \rightarrow e(A, B), i \in \{1, 2\}$  denotes party  $U_i$  taking  $(A, B)$  as inputs and giving  $e(A, B)$  as output.

- **Init**:  $T$  calls Rand2 to get random values

$$\begin{aligned} & (x_1 \cdot P_1, x_3 \cdot P_1, x_4 \cdot P_1, x_2 \cdot P_2, x_5 \cdot P_2, \\ & (x_2 - x_1^{-1}x_2x_3) \cdot P_2, e(P_1, P_2)^{x_4x_5}). \end{aligned}$$

- **Computation**: In random orders,  $T$  sends the following to  $U_1$

$$\begin{aligned} 1) & U_1(A + x_1 \cdot P_1, B + x_2 \cdot P_2) \rightarrow \alpha_1, \\ 2) & U_1(x_4 \cdot P_1, x_5 \cdot P_2) \rightarrow \alpha'_1. \end{aligned}$$

Similarly, in random orders,  $T$  sends the following to  $U_2$

$$\begin{aligned} 1) & U_2(A + x_3 \cdot P_1, -x_2 \cdot P_2) \rightarrow \alpha_2, \\ 2) & U_2(-x_1 \cdot P_1, B + (x_2 - x_1^{-1}x_2x_3) \cdot P_2) \rightarrow \alpha_3 \\ 3) & U_2(x_4 \cdot P_1, x_5 \cdot P_2) \rightarrow \alpha'_2. \end{aligned}$$

- **Recover**:  $T$  checks whether  $\alpha'_1 \stackrel{?}{=} \alpha'_2 \stackrel{?}{=} e(P_1, P_2)^{x_4x_5}$ . If the verifications are successful then it computes

$$\beta = \alpha_1\alpha_2\alpha_3$$

and produces  $\beta$  as output. Otherwise, it rejects and gives an "Error".

### 3) Security Analysis:

*Theorem 8*: Under the OMTUP assumption, the algorithms  $(T, (U_1, U_2))$  of "Algorithm 1" are an outsource-secure implementation of a pairing evaluation, where the input  $(A, B)$  may be honest, secret, honest, protected, or adversarial, protected.

	Algorithm A [13]	Algorithm 1	Algorithm 2
PC	6	4	5

TABLE III  
COMPARISON OF SERVER'S WORKLOAD

	Algorithm A [13]	Algorithm 1	Algorithm 2
PC	$\approx 0,117$ KB	$\approx 0,078$ KB	0,098 KB

TABLE IV  
COMPARISON OF COMMUNICATION OVERHEAD FOR 80-BIT SECURITY

*Proof*: The correctness is straight forward:

$$\begin{aligned} \beta &= \alpha_1 \cdot \alpha_2 \cdot \alpha_3 \\ &= e(A + x_1 \cdot P_1, B + x_2 \cdot P_2) \cdot e(A + x_3 \cdot P_1, -x_2 \cdot P_2) \cdot \\ & e(-x_1 \cdot P_1, B + (x_2 - x_1^{-1}x_2x_3) \cdot P_2) \\ &= e(A, B)e(P_1, P_2)^{x_1x_2} \cdot e(P_1, P_2)^{-x_2x_3} \cdot \\ & e(P_1, P_2)^{-x_1(x_2 - x_1^{-1}x_2x_3)} \\ &= e(A, B). \end{aligned}$$

The proof of the security part follows analogously to the proof of Theorem 6.

## IV. COMPLEXITY ANALYSIS

### A. Comparisons

We compare the precomputation algorithms of our proposed schemes with Tian *et al.*'s algorithm [13]. In the following tables, SM denotes scalar multiplication in  $\mathbb{G}_1, \mathbb{G}_2$ , ME modular exponentiation, PC pairing computation on the server side, FM field multiplication, and PA point addition in  $\mathbb{G}_1, \mathbb{G}_2$ . Furthermore,  $k$  denotes the size of the set  $S$  in the algorithms Rand1 and Rand2. Table I compares the precomputation, Table II compares the client's workload, and Table III compares the server's workload. Table IV compares the communication overhead between the client and the servers, and finally Table V gives the memory requirements for ST and DT by means of counting the number of group elements.

## V. CONCLUSION

In this paper, we studied outsourcing the computation of bilinear maps and proposed two new efficient algorithms decreasing both the memory requirement and the overall communication overhead. We defined the necessary security model, and proved the correctness and the security of the proposed secure outsourcing algorithms. We further gave the comparisons of our algorithms with a very recent outsourcing mechanism of Tian *et al.* [13] with respect to the offline and online computations, and the memory to be used. In this

	ST	DT
Algorithm 1	3	8
Algorithm 2	3	7

TABLE V  
MEMORY REQUIREMENTS FOR Rand ALGORITHMS.

way, we show that our algorithms are more efficient than all previously proposed solutions.

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