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## Secure Two-Party Comparison Against Untrusted Participants

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**Date:** 2023-04-08T00:00:00+00:00

### Abstract

Secure two-party comparison is widely used in constructing various secure computation protocols, such as secure machine learning training and inference. Existing secure two-party comparison protocols typically have one party learn the comparison result first and then inform the other party, making it difficult to prevent the party that learns the result first from tampering with the comparison result. To address this problem, this paper first proposes a new paradigm for secure two-party comparison that is resistant to untrusted participants. Subsequently, this paper designs a secure two-party comparison protocol satisfying the new paradigm using a threshold Paillier cryptosystem. This protocol allows both parties involved in the comparison to obtain a consistent comparison result without revealing their respective data, and the protocol guarantees that no participant can tamper with the comparison result. Rigorous theoretical analysis and proofs demonstrate that the proposed protocol is correct and secure. Experimental results show that the proposed protocol outperforms existing secure two-party comparison methods in terms of computational efficiency and functionality. Compared with existing secure two-party comparison methods, the computational efficiency of the protocol in this paper is improved by 50 times.

### Full Text

#### A Novel Two-Party Comparison Protocol Against Untrusted Parties

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## Abstract

Secure two-party comparison is widely used to build various secure computing protocols (e.g., secure training, secure inference). In existing secure two-party comparison protocols, there is always one party that obtains a comparison result first, and then notifies the other party of this result. Consequently, they cannot prevent the party that obtains the result first from tampering with it. To address this problem, this paper first proposes a new paradigm for secure two-party comparison against untrusted parties. Then, we design a secure two-party comparison protocol (TOMS) satisfying this new paradigm based on the threshold Paillier cryptosystem. Each party in TOMS obtains the same comparison result without revealing their private data. Moreover, TOMS prevents any party from tampering with the comparison results. Strict theoretical analyses demonstrate the security and correctness of TOMS. Finally, experimental results show that TOMS outperforms existing secure two-party comparison methods in terms of computational efficiency and functionality, achieving a  $50\times$  speedup over previous approaches.

**Keywords:** cryptography; trusted computing; secure comparison; secure two-party computation; homomorphic encryption; threshold cryptography

This work was supported by the National Key Research and Development Program of China (No. 2022YFB3102700) and the National Natural Science Foundation of China (Nos. 62202358, 62102295, 62072109, U1804263, 61632013).

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## 1 Introduction

Secure two-party comparison enables two participants to jointly execute a comparison function  $f(x, y)$  and obtain results  $u_a$  and  $u_b$  without revealing their respective input data  $x$  and  $y$  to each other, where  $(u_a, u_b) \leftarrow f(x, y)$ . This problem originated from Yao's Millionaires' Problem proposed in 1982, which has become a key technology in secure multi-party computation. Yao's original solution required exponential numbers of decryption and verification operations, resulting in enormous time and space complexity that limited practical applicability. To improve practicality, Yao later proposed a secure two-party computation scheme based on garbled circuits in 1986.

Since then, researchers have focused on designing more efficient solutions. Ioannidis et al. addressed secure two-party comparison by parallelizing  $n$  rounds of 1-out-of-2 oblivious transfer, where  $n$  is the bit length of the input data, achieving better efficiency than exponential operations. Li et al. solved the problem using set intersection combined with symmetric encryption, which improved efficiency but performed poorly with large inputs. Damgård implemented an integer comparison protocol using a specialized DGK homomorphic encryption scheme that offered superior efficiency, though initialization took approximately 150 seconds due to security considerations. Recent work also addresses security: Li et al. used ElGamal encryption for semi-honest models and extended to malicious models using zero-knowledge proofs and cut-and-choose protocols.

Secure two-party comparison has found applications in privacy-preserving auctions, machine learning, outsourced computation, and interval computation. Damle et al. proposed a solution for combinatorial auctions (TPACAS) that protects auction-related privacy. They later developed a verifiable solution using semi-trusted third-party proxies and Ethereum. Abspoel et al. applied Legendre symbol-based comparison to secure neural network classifiers, achieving  $5\times$  efficiency improvements over MPyC's built-in protocol on MNIST. Liu et al. designed a floating-point outsourcing framework (POCF) with a sub-protocol SLT for comparing ciphertexts. Liu et al. also transformed interval computation into secure comparison using zero-knowledge proofs and Goldwasser-Micali encryption. Guo et al. combined Paillier cryptosystems with geometric theory for efficient rational interval computation. Zhao et al. constructed secure sorting protocols based on ciphertext comparison to address privacy leaks in particle swarm optimization.

Despite these efforts to improve performance and apply secure two-party comparison, existing protocols overlook a critical vulnerability: untrusted participants can tamper with comparison results, undermining protocol reliability. Typically, one party (e.g., Alice) obtains the result  $u_a$  first, then notifies Bob of  $u_b$ . If Alice is trusted, she sets  $u_b = u_a$  and both parties receive identical results. However, if Alice is untrusted, she can set  $u_b \neq u_a$ , causing Bob to

receive an incorrect result.

**Problem Scenario:** Two millionaires, Alice and Bob, wish to purchase a property, agreeing that only the wealthier one may buy it, yet neither wants to reveal their total assets. While existing protocols seem applicable, the millionaire who learns the result first can easily deceive the other, making it impossible for both to confidently determine who has purchasing rights.

To solve this, we propose methods for secure two-party comparison against untrusted participants. Technically, our approach ensures that Alice and Bob, with private inputs  $x$  and  $y$ , always obtain identical comparison results ( $u_a = u_b$ ) after executing protocol  $(u_a, u_b) \leftarrow f(x, y)$ , with neither party able to tamper with the result. Our contributions are threefold:

1. We propose a novel paradigm for secure two-party comparison against untrusted parties that mandates identical results for both participants and prevents result tampering.
2. We design a concrete protocol (TOMS) using threshold Paillier cryptosystem key splitting and secure two-party computation principles to prevent untrusted participant attacks.
3. The protocol is provably secure and efficient. Rigorous analysis demonstrates that result tampering is computationally infeasible, and experiments show  $50\times$  speedup over comparable methods.

The remainder is organized as follows: Section 2 surveys related work. Section 3 presents our paradigm. Section 4 details the threshold Paillier cryptosystem, system model, threat model, and protocol design. Section 5 proves CPA security and compliance with our paradigm. Section 6 evaluates efficiency and feasibility. Section 7 concludes.

## 2 Related Work

We review three categories of secure two-party comparison protocols: garbled circuit-based, homomorphic encryption-based, and secret sharing-based approaches.

**Garbled Circuit-Based Methods.** These transform comparison problems into garbled circuits. ObliVM compiles ObliVM-lang (a Java-like language) to implement efficient ORAM schemes for semi-honest models. ABY is a C++ framework that enables fine-grained efficiency control through protocol conversion mechanisms. Both provide generic solutions (optimized with free-XOR) that can handle comparison. For malicious participants, cut-and-choose methods prevent attacks by having the sender construct multiple circuits and the receiver randomly verify subsets. Canetti et al. used this approach for correctness guarantees, but constructing numerous circuits yields high computational and space overhead, limiting practicality.

**Homomorphic Encryption-Based Methods.** Homomorphic encryption's ability to map ciphertext operations to plaintext operations makes it ideal for

privacy-preserving comparison. Lin et al. combined ElGamal with set intersection methods, comparing 0/1 encodings while protecting 1-encodings. Liu et al. represented wealth values as vectors and used Paillier's homomorphic properties. Li et al. proposed max/min comparison protocols for semi-honest and malicious models using ElGamal, vectorization, and zero-knowledge proofs. Liu et al. developed POQR for natural numbers using threshold Paillier, with sub-protocol SLT for ciphertext comparison. Zhao et al. improved this work with SOCI, whose SCMP sub-protocol supports integer comparison and resists chosen-plaintext attacks. Our TOMS protocol builds upon SCMP.

**Secret Sharing-Based Methods.** Nishide proposed an efficient bit-decomposition protocol using bitwise secret sharing. Liu et al. vectorized wealth values and used secret sharing instead of Paillier to reduce complexity for multi-party comparison. Damgård et al. transformed polynomial shares of secrets into bit shares, enabling efficient comparison via multiplication protocols.

Table 1 compares our protocol with representative schemes. SemiSMC uses homomorphic encryption, semiSSC uses secret sharing, while OblivM and ABY use garbled circuits. Neither OblivM nor ABY can prevent untrusted participants from tampering with results. Our TOMS protocol achieves this protection.

### 3 Paradigm for Secure Two-Party Comparison Against Untrusted Parties

As shown in Figure 1 [Figure 1: see original paper], in an ideal secure two-party comparison, Alice and Bob with private data  $x$  and  $y$  jointly execute function  $(u_a, u_b) \leftarrow f(x, y)$  through a trusted third party to determine their relative magnitudes, without revealing inputs and with both obtaining correct, identical results  $u_a = u_b$ .

**Definition 1.** Let Alice and Bob execute two-party secure comparison protocol  $f(x, y)$  with private inputs  $x$  and  $y$ , obtaining results  $u_a$  and  $u_b$ . A paradigm for secure two-party comparison against untrusted parties satisfies:

- **Confidentiality:** After protocol execution, Alice's input  $x$  does not leak to Bob, and vice versa.
- **Correctness:** After execution, Alice and Bob obtain results  $u_a$  and  $u_b$  where  $u_a = u_b$  always holds. Both parties consistently receive identical comparison results.
- **Reliability:** During execution, the party that obtains the result first (e.g., Alice) cannot tamper with  $u_b$  to make  $u_b \neq u_a$ , and the later party (e.g., Bob) can easily detect any tampering.

## 4 Secure Two-Party Comparison Protocol Against Untrusted Parties

We first introduce the threshold Paillier cryptosystem, then describe TOMS's system model, threat model, detailed design, and correctness analysis.

### 4.1 Threshold Paillier Cryptosystem

We describe a  $(2, 2)$  threshold Paillier cryptosystem comprising key generation, encryption, decryption, key splitting, partial decryption, and threshold decryption. Unlike standard Paillier, the private key is split into two partial keys.

**Key Generation (KeyGen):** Let  $k$  be the security parameter. Choose large primes  $p, q$  with  $|p| = |q| = k$ . Compute  $N = p \cdot q$ ,  $\lambda = (p - 1)(q - 1)/2$ , and  $\mu = \lambda^{-1} \bmod N$ . Define  $L(x) = \frac{x-1}{N}$  and select generator  $g = N + 1$ . The public key is  $pk = (N, g)$  and private key is  $sk = \lambda$ .

**Encryption (Enc):** For plaintext  $m \in \mathbb{Z}_N$ , output ciphertext  $\llbracket m \rrbracket = \text{Enc}(pk, m) = g^m r^N \bmod N^2$ , where  $r$  is randomly chosen from  $\mathbb{Z}_N^*$ .

**Decryption (Dec):** For ciphertext  $\llbracket m \rrbracket$ , output plaintext  $m = \text{Dec}(sk, \llbracket m \rrbracket) = \frac{L(\llbracket m \rrbracket^\lambda \bmod N^2)}{L(g^\lambda \bmod N^2)} \bmod N$ .

**Key Splitting (KeyS):** Input private key  $sk = \lambda$ , output split keys  $sk_1 = \lambda_1$  and  $sk_2 = \lambda_2$  satisfying  $\lambda_1 + \lambda_2 \equiv 0 \pmod{\lambda}$  and  $\lambda_1 + \lambda_2 \equiv 1 \pmod{N}$ . By the Chinese Remainder Theorem, we compute  $\delta = \lambda_1 + \lambda_2 = \lambda \cdot \mu \pmod{\lambda \cdot N}$ . Choose  $\lambda_1$  as a  $\sigma$ -bit random number and set  $\lambda_2 = \lambda \cdot \mu + \eta \cdot \lambda N - \lambda_1$ , where  $\eta$  is a non-zero integer.

**Partial Decryption (PDec):** Input ciphertext  $\llbracket m \rrbracket$  and partial key  $sk_i$  ( $i \in \{1, 2\}$ ), output partial decryption result  $M_i = \text{PDec}(\llbracket m \rrbracket, sk_i) = \llbracket m \rrbracket^{\lambda_i} \bmod N^2$ .

**Threshold Decryption (TDec):** Input partial results  $M_1, M_2$ , output plaintext  $m = \text{TDec}(M_1, M_2) = L(M_1 \cdot M_2 \bmod N^2) \cdot \mu \bmod N$ .

The threshold Paillier cryptosystem provides additive and scalar multiplication homomorphisms: - **Additive:**  $\text{Dec}(sk, \llbracket m_1 \rrbracket \cdot \llbracket m_2 \rrbracket \bmod N^2) = m_1 + m_2 \bmod N$  - **Scalar Multiplication:**  $\text{Dec}(sk, \llbracket m \rrbracket^c \bmod N^2) = c \cdot m \bmod N$ , where  $c \in \mathbb{Z}_N$ .

### 4.2 System Model

As shown in Figure 2 [Figure 2: see original paper], TOMS involves two participants, Alice with private data  $x$  and Bob with private data  $y$ . To protect privacy, Alice uses a  $(2, 2)$  threshold Paillier system while Bob uses standard Paillier. Specifically:

- **Alice:** Generates key pair  $(pk_a, sk_a)$  where  $pk_a = (N_a, g_a)$ . She splits  $sk_a$  into  $(\lambda_1, \lambda_2) \leftarrow \text{KeyS}(sk_a)$  and distributes  $\lambda_2$  and  $pk_a$  to Bob.
- **Bob:** Generates Paillier key pair  $(pk_b, sk_b)$  where  $pk_b = (N_b, g_b)$  and distributes  $pk_b$  to Alice.

### 4.3 Threat Model

TOMS involves two entities: Alice and Bob, both assumed **semi-honest**. They follow the protocol faithfully but may attempt to infer the other's input from intermediate values. Their goals are to: (1) learn the other's input, or (2) tamper with the other's final result. Specific adversarial behaviors include:

1. Semi-honest Alice inferring Bob's input  $y$  from messages received.
2. Semi-honest Bob inferring Alice's input  $x$  from messages received.
3. Semi-honest Alice (who obtains results first) tampering with Bob's result  $u_b$  (making  $u_b \neq u_a$ ).

### 4.4 Detailed Design

Assume  $x, y \in [-2^\ell, 2^\ell]$  where  $\ell$  is an integer satisfying  $2^\ell \ll N$  (e.g.,  $\ell = 32$ ). The TOMS protocol, shown in Figure 3 [Figure 3: see original paper], takes private inputs  $x$  and  $y$  and outputs comparison results  $u_a$  and  $u_b$ , formally  $(u_a, u_b) \leftarrow F(x, y)$  where  $F$  satisfies our paradigm. The five-step process is:

**Step 1:** Bob encrypts  $y$  as  $\llbracket y \rrbracket_{pk_b}$  and sends it to Alice.

**Step 2:** Alice encrypts  $x$  as  $\llbracket x \rrbracket_{pk_b}$ . She randomly selects  $s \in \{0, 1\}$  and computes:

$$D = \begin{cases} \llbracket r_1(x - y + 1) + r_2 \rrbracket_{pk_b} & \text{if } s = 0 \\ \llbracket r_1(y - x) + r_2 \rrbracket_{pk_b} & \text{if } s = 1 \end{cases}$$

where  $r_1 \in \{0, 1\}^\sigma \setminus \{0\}$  and  $r_2 \in \{0, 1\}^\kappa \setminus \{0\}$  are random numbers ( $\sigma, \kappa$  are security parameters with  $\ell < \sigma < \kappa < |N_b|/2$ ) satisfying  $r_2 \leq N_b$  and  $r_1 + r_2 > N_b$ .

**Step 3:** Bob receives  $D$  and decrypts it to get  $d = \text{Dec}(sk_b, D)$ . He sets  $u_1 = 0$  if  $d > N_b/2$ , otherwise  $u_1 = 1$ . He encrypts  $u_1$  as  $\llbracket u_1 \rrbracket_{pk_a}$  and sends it to Alice. Alice encrypts  $s$  as  $\llbracket s \rrbracket_{pk_a}$ , partially decrypts it to get  $M_1 = \text{PDec}(\llbracket s \rrbracket_{pk_a}, \lambda_1)$ , and sends  $D, \llbracket s \rrbracket_{pk_a}, M_1$  to Bob.

**Step 4:** Alice decrypts  $\llbracket u_1 \rrbracket_{pk_a}$  to obtain  $u_1$ , computes  $u_a = s - u_1$ , and sends  $\lambda_2$  to Bob.

**Step 5:** Bob receives  $\lambda_2$ , partially decrypts  $\llbracket s \rrbracket_{pk_a}$  to get  $M_2 = \text{PDec}(\llbracket s \rrbracket_{pk_a}, \lambda_2)$ , then computes  $s = \text{TDec}(M_1, M_2)$ . Finally, Bob computes  $u_b = s - u_1$ .

### 4.5 Correctness Analysis

**Alice obtains correct results.** In Step 2, with  $x, y \in [-2^\ell, 2^\ell]$ , random  $r_1 \in \{0, 1\}^\sigma \setminus \{0\}$  and  $r_2 \in \{0, 1\}^\kappa \setminus \{0\}$  satisfy  $r_2 \leq N_b$  and  $r_1 + r_2 > N_b/2$ . We analyze cases based on  $s$ :

- **If  $s = 0$ :**  $x - y + 1 \in [-2^{\ell+1} + 1, 2^{\ell+1} + 1]$ , so  $d = r_1(x - y + 1) + r_2 \in (N_b/2 + 2 - 2^{\sigma+\ell+1} - 2^\sigma, N_b/2 + 2^{\sigma+\ell+1} + 2^\sigma)$ . Since  $N_b/2 \gg 2^{\sigma+\ell+1}$  and

$0 < d < N_b$ , Alice obtains  $u_1$  in Step 4 (if  $d \geq N_b/2$ ,  $u_1 = 0$ ; otherwise  $u_1 = 1$ ).

- When  $x \geq y$ :  $d \geq N_b/2$ , so  $u_1 = 0$  and  $u_a = s - u_1 = 0$ . TOMS outputs  $u_a = 0$  representing  $x \geq y$ .
- When  $x < y$ :  $d < N_b/2$ , so  $u_1 = 1$  and  $u_a = 0 - 1 = -1 \equiv N_b - 1 \pmod{N_b}$ . Since  $u_a > N_b/2$ , we compute  $u_a = N_b - u_a = 1$ , representing  $x < y$ .

- **If  $s = 1$ :**  $y - x \in [-2^{\ell+1}, 2^{\ell+1}]$ , so  $d = r_1(y - x) + r_2 \in (N_b/2 + 2 - 2^{\sigma+\ell+1} - 2^\sigma, N_b/2 + 2^{\sigma+\ell+1} + 2^\sigma)$ . Similar analysis shows Alice always obtains correct results.

**Bob obtains correct results.** Bob sends  $\llbracket y \rrbracket_{pk_b}$  in Step 1. In Step 3, he decrypts  $D$  to get  $d$ , derives  $u_1$  from  $d$  versus  $N_b/2$ , and obtains encrypted  $\llbracket u_1 \rrbracket_{pk_a}$ . In Step 5, after receiving  $\lambda_2$ , he partially decrypts  $\llbracket s \rrbracket_{pk_a}$  to get  $M_2$ , then fully decrypts to obtain  $s = \text{TDec}(M_1, M_2)$ . Since Alice correctly obtains  $u_a$  by Step 4 and both parties use identical  $s$  and  $u_1$ , Bob correctly computes  $u_b = s - u_1$ .

Since  $u_a = s - u_1$  and  $u_b = s - u_1$ , we have  $u_a = u_b$  always holds.

## 5 Security Analysis

We first define CPA security and prove that the encryption scheme  $r_1(x - y + 1) + r_2$  (or  $r_1(y - x) + r_2$ ) used in Step 2 is CPA-secure. We then define semi-honest security and prove TOMS's security via simulation paradigm. Finally, we verify TOMS satisfies our paradigm.

### 5.1 CPA Security Definition

The experiment  $\text{PubK}_{A, r_{1m} + r_2}^{\text{CPA}}(\sigma, \kappa)$  involves adversary  $A$  and challenger  $C$ :

1.  $A$  selects messages  $m_0, m_1$  and sends them to  $C$ .
2.  $C$  randomly chooses  $b \in \{0, 1\}$ , generates random  $r_1, r_2$ , and sends  $r_{1m}b + r_2$  to  $A$ .
3.  $A$  outputs guess  $b' \in \{0, 1\}$ .
4. If  $b = b'$ , the experiment returns 1 (adversary succeeds); otherwise 0.

**Definition 2.** The encryption scheme  $r_{1m} + r_2$  is CPA-secure if for any probabilistic polynomial-time adversary  $A$ , there exists a negligible function  $\text{negl}(\sigma, \kappa)$  such that:

$$\Pr[\text{PubK}_{A, r_{1m} + r_2}^{\text{CPA}}(\sigma, \kappa) = 1] \leq \frac{1}{2} + \text{negl}(\sigma, \kappa)$$

**Theorem 1.** For  $m_0, m_1 \in [-2^\ell, 2^\ell]$ ,  $r_{1,0}, r_{1,1} \in [2^{\sigma-1}, 2^\sigma - 1]$  and  $r_{2,0}, r_{2,1} \in [2^{\kappa-1}, 2^\kappa - 1]$ , the values  $r_{1,0}m_0 + r_{2,0}$  and  $r_{1,1}m_1 + r_{2,1}$  are computationally indistinguishable. Formally, challenger  $C$  randomly selects  $m_b \in [-2^\ell, 2^\ell]$ ,  $r_1 \in$

$[2^{\sigma-1}, 2^{\sigma} - 1]$ ,  $r_2 \in [2^{\kappa-1}, 2^{\kappa} - 1]$  and computes  $r_{1m}b + r_2$ . For adversary  $A$  inferring  $b' = b$  from  $r_{1m}b + r_2$ , we have  $\Pr[b' = b \mid r_{1m}b + r_2] \leq \frac{1}{2} + \text{negl}(\kappa)$ .

**Proof.** Since  $r_1, r_2$  are randomly selected,  $r_{1m}b + r_2$  is uniformly distributed in  $[2^{\kappa-1} - 2^{\ell+\sigma} + 2^{\ell}, 2^{\kappa} + 2^{\ell+\sigma} - 2^{\ell} - 1]$ . The adversary's success probability is maximized when  $m_0 = -2^{\ell}$ ,  $m_1 = 2^{\ell}$ :

- If  $b = 0$  and  $r_{1m}b + r_2 \in [2^{\kappa-1} - 2^{\ell+\sigma} + 2^{\ell}, 2^{\kappa-1} + 2^{\ell+\sigma-1} - 1]$ ,  $A$  guesses correctly with probability 1.
- If  $b = 1$  and  $r_{1m}b + r_2 \in [2^{\kappa} - 2^{\ell+\sigma-1}, 2^{\kappa} + 2^{\ell+\sigma} - 2^{\ell} - 1]$ ,  $A$  guesses correctly with probability 1.

The maximum success probability is:

$$\Pr[b' = b \mid r_{1m}b + r_2] \leq \frac{3 \cdot 2^{\ell+\sigma} - 2^{\ell+1}}{2^{\kappa}}$$

Given  $\ell < \sigma < \kappa$ , we have  $2^{\kappa} \gg 3 \cdot 2^{\ell+\sigma} - 2^{\ell+1}$ , so there exists negligible function  $\text{negl}(\kappa) = \frac{3 \cdot 2^{\ell+\sigma} - 2^{\ell+1}}{2^{\kappa}}$  satisfying the CPA security bound.

## 5.2 Semi-Honest Security Definition

**Ideal Model:** Let  $f : \{0, 1\}^* \times \{0, 1\}^* \rightarrow \{0, 1\}^* \times \{0, 1\}^*$  be a probabilistic polynomial-time function representing the secure computation, where  $f(x, y) = (f_1(x, y), f_2(x, y))$ . In the ideal model, Alice and Bob send inputs  $x$  and  $y$  to a Trusted Third Party (TTP), which computes  $f(x, y)$  and returns  $f_1(x, y)$  to Alice and  $f_2(x, y)$  to Bob. The TTP reveals nothing beyond these outputs, making the ideal model maximally secure.

**Simulation Paradigm:** A real protocol  $\pi$  is secure if it leaks no more information than the ideal model. Specifically, a participant's view during protocol execution should be simulatable from only its input and output.

Let  $\text{view}_1^{\pi}(x, y) = (x, r, m_1, \dots, m_t)$  and  $\text{view}_2^{\pi}(x, y) = (y, r', m'_1, \dots, m'_t)$  denote Alice's and Bob's views during protocol  $\pi$ , where  $r, r'$  are random coin tosses and  $m_i, m'_i$  are received messages. Their outputs are  $\text{output}_1^{\pi}(x, y)$  and  $\text{output}_2^{\pi}(x, y)$ .

**Definition 3.** Protocol  $\pi$  securely computes  $f$  if there exist probabilistic polynomial-time simulators  $S_1$  and  $S_2$  such that:

$$\begin{aligned} \{(S_1(x, f_1(x, y)), f_2(x, y))\}_{x,y} &\equiv \{(\text{view}_1^{\pi}(x, y), \text{output}_2^{\pi}(x, y))\}_{x,y} \\ \{(S_2(y, f_2(x, y)), f_1(x, y))\}_{x,y} &\equiv \{(\text{view}_2^{\pi}(x, y), \text{output}_1^{\pi}(x, y))\}_{x,y} \end{aligned}$$

where  $\equiv$  denotes computational indistinguishability.

## 5.3 Security Proof

**Theorem 2.** The secure two-party comparison protocol against untrusted parties is secure under semi-honest conditions.

**Proof.** We construct simulators  $S_1$  and  $S_2$  per Definition 3.

**Constructing  $S_1$ :** 1. Randomly selects  $y' \in [-2^\ell, 2^\ell]$  such that  $f_1(x, y) = f_1(x, y') = u_a$ . Encrypts  $y'$  as  $\llbracket y' \rrbracket_{pk_b}$ . 2. Randomly selects  $r_1 \in \{0, 1\}^\sigma \setminus \{0\}$ ,  $r_2 \in \{0, 1\}^\kappa \setminus \{0\}$ , and  $s \in \{0, 1\}$ . Encrypts  $s$  as  $\llbracket s \rrbracket_{pk_a}$  and partially decrypts to get  $M_1 = \text{PDec}(\llbracket s \rrbracket_{pk_a}, \lambda_1)$ . Computes  $D' = \llbracket r_1(x - y' + 1) + r_2 \rrbracket_{pk_b}$  if  $s = 0$ , else  $D' = \llbracket r_1(y' - x) + r_2 \rrbracket_{pk_b}$ . 3. Decrypts  $D'$  to get  $d' = \text{Dec}(sk_b, D')$ . Sets  $u'_1 = 0$  if  $d' > N_b/2$ , else  $u'_1 = 1$ . Encrypts  $u'_1$  as  $\llbracket u'_1 \rrbracket_{pk_a}$ . 4. Decrypts  $\llbracket u'_1 \rrbracket_{pk_a}$  to get  $u'_1$ , computes  $u'_a = s - u'_1$ . 5. Partially decrypts  $\llbracket s \rrbracket_{pk_a}$  to get  $M_2 = \text{PDec}(\llbracket s \rrbracket_{pk_a}, \lambda_2)$ , fully decrypts to get  $s = \text{TDec}(M_1, M_2)$ , and computes  $u'_b = s - u'_1$ .

The simulated view is  $S_1(x, f_1(x, y)) = (x, pk_a, pk_b, \llbracket y' \rrbracket_{pk_b}, s, \llbracket s \rrbracket_{pk_a}, M_1, M_2, D', d', u'_1, f_1(x, y'))$ . Since  $f_1(x, y) = f_1(x, y')$ , we have  $u_a = u'_a$ , implying  $u_1 = u'_1$ ,  $d = d'$ , and  $D \equiv D'$ . By Paillier's semantic security,  $\llbracket y \rrbracket_{pk_b} \equiv \llbracket y' \rrbracket_{pk_b}$  and  $\llbracket s \rrbracket_{pk_a} \equiv \llbracket s \rrbracket_{pk_a}$ . Thus:

$$\{(S_1(x, f_1(x, y)), f_2(x, y))\}_{x, y} \equiv \{(\text{view}_1^\pi(x, y), \text{output}_2^\pi(x, y))\}_{x, y}$$

**Constructing  $S_2$ :** 1. Randomly selects  $\bar{x} \in [-2^\ell, 2^\ell]$  such that  $f_2(x, y) = f_2(\bar{x}, y) = u_b$ . Encrypts  $y$  as  $\llbracket y \rrbracket_{pk_b}$ . 2. Randomly selects  $\bar{r}_1 \in \{0, 1\}^\sigma \setminus \{0\}$ ,  $\bar{r}_2 \in \{0, 1\}^\kappa \setminus \{0\}$ , and  $\bar{s} \in \{0, 1\}$ . Encrypts  $\bar{s}$  as  $\llbracket \bar{s} \rrbracket_{pk_a}$ , partially decrypts to  $\bar{M}_1 = \text{PDec}(\llbracket \bar{s} \rrbracket_{pk_a}, \lambda_1)$ . Computes  $\bar{D} = \llbracket \bar{r}_1(\bar{x} - y + 1) + \bar{r}_2 \rrbracket_{pk_b}$  if  $\bar{s} = 0$ , else  $\bar{D} = \llbracket \bar{r}_1(y - \bar{x}) + \bar{r}_2 \rrbracket_{pk_b}$ . 3. Decrypts  $\bar{D}$  to  $\bar{d} = \text{Dec}(sk_b, \bar{D})$ . Sets  $\bar{u}_1 = 0$  if  $\bar{d} > N_b/2$ , else  $\bar{u}_1 = 1$ . Encrypts  $\bar{u}_1$  as  $\llbracket \bar{u}_1 \rrbracket_{pk_a}$ . 4. Decrypts  $\llbracket \bar{u}_1 \rrbracket_{pk_a}$  to  $\bar{u}_1$ , computes  $\bar{u}_a = \bar{s} - \bar{u}_1$ . 5. Partially decrypts  $\llbracket \bar{s} \rrbracket_{pk_a}$  to  $\bar{M}_2 = \text{PDec}(\llbracket \bar{s} \rrbracket_{pk_a}, \lambda_2)$ , fully decrypts to  $\bar{s} = \text{TDec}(\bar{M}_1, \bar{M}_2)$ , computes  $\bar{u}_b = \bar{s} - \bar{u}_1$ .

The simulated view is  $S_2(y, f_2(x, y)) = (y, pk_a, pk_b, \llbracket \bar{s} \rrbracket_{pk_a}, \bar{M}_1, \bar{M}_2, \bar{D}, \bar{d}, \bar{u}_1, f_1(\bar{x}, y))$ . Since  $f_2(x, y) = f_2(\bar{x}, y)$ , we have  $u_b = \bar{u}_b$ , implying  $\bar{s} \equiv s$ ,  $\bar{u}_1 \equiv u_1$ , and by Theorem 1,  $d \equiv \bar{d}$  and  $D \equiv \bar{D}$ . By Paillier's semantic security,  $M_1 \equiv \bar{M}_1$  and  $M_2 \equiv \bar{M}_2$ . Thus:

$$\{(S_2(y, f_2(x, y)), f_1(x, y))\}_{x, y} \equiv \{(\text{view}_2^\pi(x, y), \text{output}_1^\pi(x, y))\}_{x, y}$$

**Reliability:** Since  $d = r_1(y - x) + r_2$  (or  $d = r_1(x - y + 1) + r_2$ ), adversary Alice cannot change Bob's final result  $u_b$  by selecting a specific  $d'$ . By Theorem 1,  $d$  and  $d'$  are computationally indistinguishable, so Alice cannot tamper with  $u_b$ . Therefore, TOMS satisfies the reliability requirement of our paradigm.

## 6 Experimental Evaluation

We verify that TOMS satisfies our paradigm (confidentiality, correctness, reliability) and compare its performance with OblivM, ABY, and semiSMC.

**Confidentiality:** Neither Alice's nor Bob's inputs leak. Alice's  $x$  is protected because Bob receives only  $D = \llbracket r_1(x - y + 1) + r_2 \rrbracket_{pk_b}$  or  $\llbracket r_1(y - x) + r_2 \rrbracket_{pk_b}$ ,

which is CPA-secure per Theorem 1. Bob's  $y$  is protected because Alice only receives  $\llbracket y \rrbracket_{pk_b}$  without  $sk_b$ .

**Correctness:** Proven in Section 4.5.

**Reliability:** Alice cannot tamper with Bob's result  $u_b = s - u_1$  because  $d$  and  $d'$  are computationally indistinguishable (Theorem 1), making it infeasible for Alice to manipulate  $u_1$  by choosing a specific  $d'$ .

## 6.1 Efficiency Analysis

We measure efficiency via computational and communication complexity. Since modular exponentiation dominates, we count exponentiations while ignoring lighter operations. Pre-computation phases are excluded.

**Computational Complexity:** For TOMS with  $N_a = N_b = N$ : 1. Bob encrypts  $y$ :  $1.5|N|$  modular multiplications. 2. Alice encrypts  $x$  and  $s$ :  $3|N|$  multiplications; generates  $D$ :  $1.5|N| + 3\sigma$ ; partial decryption:  $1.5\sigma$ . 3. Bob decrypts  $D$ :  $1.5|N|$ ; encrypts  $u_1$ :  $1.5|N|$ . 4. Alice decrypts  $u_1$ :  $1.5|N|$ . 5. Bob partial decryption:  $1.5\sigma$ .

Total:  $10.5|N| + 6\sigma$  modular multiplications.

For semiSMC (configured for two parties), encryption requires  $6v|N|$  multiplications and decryption averages  $3v|N|$ , totaling  $9v|N|$ , where  $v$  is the input domain size. Since  $v \gg 2$  and  $|N| \gg \sigma$ , TOMS is more efficient.

OblivM and ABY use garbled circuits; their complexity depends on AND gate count (denoted as  $\perp$  in Table 2), making direct comparison difficult. Experiments confirm TOMS's superior efficiency.

**Communication Complexity:** OblivM and ABY require 4 rounds (garbled table, OT, result). semiSMC needs  $n(n-1)$  rounds for  $n$  parties. TOMS uses 4 rounds. Communication overhead: TOMS (1.25 KB)  $\ll$  semiSMC (49.9 KB)  $<$  ABY (10.29 KB)  $<$  OblivM (14.8 KB).

## 6.2 Experimental Tests

Experiments run on Intel Core i5-8300H CPU @ 2.30GHz, 16GB RAM, Windows 11 64-bit, using C++ and GMP 6.2. Pre-computation optimizes TOMS by pre-calculating values like  $\llbracket s \rrbracket_{pk_a}$  before execution.

**Test 1:** Varying security parameter  $N$  (512–1024 bits) with  $\ell = 32$  and  $\ell = 64$  bits. As shown in Figures 4 [Figure 4: see original paper] and 5 [Figure 5: see original paper], OblivM and ABY's runtime is independent of  $N$  (200.8 ms and 13.9 ms for  $\ell = 32$ ; 340.1 ms and 17.5 ms for  $\ell = 64$ ). TOMS and semiSMC runtime increase with  $N$ . At  $N = 1024$  bits, TOMS outperforms OblivM by 189 ms (32-bit) and 327.6 ms (64-bit), ABY by 2.1 ms (32-bit) and 5 ms (64-bit), and semiSMC by 592 ms (51.2 $\times$  speedup for 32-bit) and 598 ms (48.8 $\times$  speedup for 64-bit).

**Test 2:** Varying plaintext length  $\ell$  (8–64 bits) with fixed  $\sigma = 128$  and  $N \in \{512, 1024, 2048, 4096\}$ . As shown in Figure 6 [Figure 6: see original paper], TOMS runtime is independent of  $\ell$  for a given  $N$  (e.g.,  $\sim 12.8$  ms at  $N = 1024$  across all  $\ell$ ). This is because Paillier’s random factor  $r$  makes ciphertext size independent of plaintext length.

Results confirm TOMS achieves both our paradigm’s security guarantees and superior efficiency.

## 7 Conclusion

To address the limitation that existing secure two-party comparison protocols cannot resist untrusted participants, this paper proposes a novel protocol paradigm and a concrete instantiation, TOMS, based on threshold Paillier. Compared to existing protocols, TOMS ensures that all participants obtain identical, tamper-proof comparison results while significantly improving efficiency. Future work will extend this paradigm to general secure multi-party computation against untrusted participants.

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*Note: Figure translations are in progress. See original paper for figures.*

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