

# Secure Computation - I

CS 600.442 Modern Cryptography

Fall 2016

## Motivating Example

Consider two billionaires Alice and Bob with net worths  $x$  and  $y$ , respectively:

- They want to find out who is richer by computing the following function

$$f(x, y) = \begin{cases} 1 & \text{if } x > y \\ 0 & \text{otherwise} \end{cases}$$

- Potential Solution: Alice sends  $x$  to Bob, who sends  $y$  to Alice. They each compute  $f$  on their own.
- Problem: Alice learns Bob's net worth (and vice-versa). No privacy!
- Main Question: Can Alice and Bob compute  $f$  in a “secure manner” s.t. they only learn the output of  $f$ , and *nothing more*?

# General Setting

Two parties  $A$  and  $B$ , with private inputs  $x$  and  $y$ , respectively:

- They want to “securely” compute a function  $f$
- If both  $A$  and  $B$  are honest, then they should learn the output  $f(x, y)$
- Even if one party is adversarial, it should not learn anything beyond the output (and its own input)
- Think: How to formalize this security requirement?

# Types of Adversaries

Two types of adversaries:

- **Honest but curious (a.k.a. semi-honest):** Such an adversary follows the instructions of the protocol, but will later analyze the protocol transcript to learn any “extra information” about the input of the other party
- **Malicious:** Such an adversary can deviate from the protocol instructions and follow an arbitrary strategy

**Note:** We will only consider *semi-honest* adversaries

# Secure Computation: Intuition

- Want to formalize that no semi-honest adversary learns anything from the protocol execution beyond its input and the (correct) output
- Idea: Use simulation paradigm, as in zero-knowledge proofs
- View of adversary in the protocol execution can be efficiently simulated given only its input and output, and without the input of the honest party

# Secure Computation: Definition

## Definition (Semi-honest Secure Computation)

A protocol  $\pi = (A, B)$  securely computes a function  $f$  in the semi-honest model if there exists a pair of non-uniform PPT simulator algorithms  $\mathcal{S}_A, \mathcal{S}_B$  such that for every security parameter  $n$ , and all inputs  $x, y \in \{0, 1\}^n$ , it holds that:

$$\left\{ \mathcal{S}_A(x, f(x, y)), f(x, y) \right\} \approx \left\{ e \leftarrow [A(x) \leftrightarrow B(y)] : \text{View}_A(e), \text{Out}_B(e) \right\},$$

$$\left\{ \mathcal{S}_B(y, f(x, y)), f(x, y) \right\} \approx \left\{ e \leftarrow [A(x) \leftrightarrow B(y)] : \text{View}_B(e), \text{Out}_A(e) \right\}.$$

## Remarks on Definition

- Recall: In zero-knowledge, we only require indistinguishability of simulated view and adversary's view in the real execution
- Here, indistinguishability is w.r.t. the *joint distribution* over the adversary's view and the honest party's output
- This is necessary for **correctness**: it implies that output of the honest party in the protocol execution must be indistinguishable from the correct output  $f(x, y)$
- If we remove this requirement, then a clearly wrong protocol where parties are instructed to output  $y$  would be trivially secure!

# Oblivious Transfer

Consider the following functionality, called, 1-out-of-2 oblivious transfer (OT):

- Two parties: Sender  $A$ , and Receiver  $B$
- Inputs:  $A$ 's input is a pair of bits  $(a_0, a_1)$ , and  $B$ 's input is a bit  $b$
- Outputs:  $B$ 's output is  $a_b$ , and  $A$  receives no output

**Note:** Definition of secure computation promises that in a secure OT protocol,  $A$  does not learn  $b$  and  $B$  does not learn  $a_{1-b}$



# Importance of Oblivious Transfer

- Can be realized from physical channels [Wiener,Rabin]
- **OT is complete:** given a secure protocol for OT, any function can be securely computed
- **OT is necessary:** OT is the minimal assumption for secure computation

## Oblivious Transfer: Construction

Let  $\{f_i\}_{i \in \mathcal{I}}$  be a family of trapdoor permutations with sampling algorithm  $\text{Gen}$ . Let  $h$  be a hardcore predicate for any  $f_i$ .

**Sender's input:**  $(a_0, a_1)$  where  $a_i \in \{0, 1\}$

**Receiver's input:**  $b \in \{0, 1\}$

**Protocol OT =  $(A, B)$ :**

$A \rightarrow B$ :  $A$  samples  $(f_i, f_i^{-1}) \leftarrow \text{Gen}(1^n)$  and sends  $f_i$  to  $B$

$B \rightarrow A$ :  $B$  samples  $x \xleftarrow{\$} \{0, 1\}^n$  and computes  $y_b = f_i(x)$ . It also samples  $y_{1-b} \xleftarrow{\$} \{0, 1\}^n$ .  $B$  sends  $(y_0, y_1)$  to  $A$

$A \rightarrow B$ :  $A$  computes the inverse of each value  $y_j$  and XORs the hard-core bit of the result with  $a_j$ :

$$z_j = h(f_i^{-1}(y_j)) \oplus a_j$$

$A$  sends  $(z_0, z_1)$  to  $B$

$B(x, b, z_0, z_1)$ :  $B$  outputs  $h(x) \oplus z_b$

## OT = (A, B) is Semi-honest Secure : Intuition

- Security against  $A$ : Both  $y_0$  and  $y_1$  are uniformly distributed and therefore independent of  $b$ . Thus,  $b$  is hidden from  $A$
- Security against  $B$ : If  $B$  could learn  $a_{1-b}$ , then it would be able to predict the hardcore predicate

**Note:** A *malicious*  $B$  can easily learn  $a_{1-b}$  by deviating from the protocol strategy

## OT = (A, B) is Semi-honest Secure : Simulator $\mathcal{S}_A$

**Simulator  $\mathcal{S}_A((a_0, a_1), \perp)$ :**

- 1 Fix a random tape  $r_A$  for  $A$ . Run honest emulation of  $A$  using  $(a_0, a_1)$  and  $r_A$  to obtain the first message  $f_i$
- 2 Choose two random strings  $y_0, y_1 \in \{0, 1\}^n$  as  $B$ 's message
- 3 Run honest emulation of  $A$  using  $(y_0, y_1)$  to obtain the third message  $(z_0, z_1)$
- 4 Stop and output  $\perp$

**Claim:** The following two distributions are identical:

$$\left\{ \mathcal{S}_A((a_0, a_1), \perp), a_b \right\} \text{ and } \left\{ e \leftarrow [A(a_0, a_1) \leftrightarrow B(b)] : \text{View}_A(e), \text{Out}_B(e) \right\}$$

**Proof:** The only difference between  $\mathcal{S}_A$  and real execution is in step 2. However, since  $f$  is a permutation,  $y_0, y_1$  are identically distributed in both cases.

## OT = (A, B) is Semi-honest Secure : Simulator $\mathcal{S}_B$

**Simulator  $\mathcal{S}_B(b, a_b)$ :**

- 1 Sample  $f_i$
- 2 Choose random tape  $r_B$  for  $B$ . Run honest emulation of  $B$  using  $(b, r_B, f_i)$  to produce  $(x, y_0, y_1)$  s.t.  $y_b = f_i(x)$  and  $y_{1-b} \xleftarrow{\$} \{0, 1\}^n$
- 3 Compute  $z_b = h(x) \oplus a_b$  and  $z_{1-b} \xleftarrow{\$} \{0, 1\}$
- 4 Output  $(z_0, z_1)$  as third message and stop

**Claim:** The following two distributions are indistinguishable:

$$\left\{ \mathcal{S}_B(b, a_b), \perp \right\} \text{ and } \left\{ e \leftarrow [A(a_0, a_1) \leftrightarrow B(b)] : \text{View}_B(e), \text{Out}_A(e) \right\}$$

**Proof:** The only difference is in step 3, where  $\mathcal{S}_B$  computes  $z_{1-b}$  as a random bit. However, since  $h(f_i^{-1}(y_{1-b}))$  is indistinguishable from random (even given  $y_{1-b}$ ), this change is indistinguishable

## 1-out-of- $k$ OT:

- The previous protocol can be easily generalized to construct 1-out-of- $k$  OT for  $k > 2$

## Semi-honest vs Malicious:

- In reality, adversary may be malicious and not semi-honest
- Goldreich-Micali-Wigderson [GMW] gave a compiler to transform *any* protocol secure against semi-honest adversary into one secure against malicious adversary
- The transformation uses coin-flipping (to make sure that adversary's random tape is truly random) and zero-knowledge proofs (to make sure that adversary is following the protocol instructions)
- Details outside the scope of this class