Introduction to Secure Multi-Party Computation

Computer Systems Security ENEE 457/CMSC 498E

> Based on notes from: Vitaly Shmatikov

Motivation

- General framework for describing computation between parties who do not trust each other
- Example: elections
 - N parties, each one has a "Yes" or "No" vote
 - Goal: determine whether the majority voted "Yes", but no voter should learn how other people voted
- Example: auctions
 - Each bidder makes an offer
 - Offer should be committing! (can't change it later)
 - Goal: determine whose offer won without revealing losing offers

More Examples

Example: distributed data mining

- Two companies want to compare their datasets without revealing them
 - For example, compute the intersection of two lists of names
- Example: database privacy
 - Evaluate a query on the database without revealing the query to the database owner
 - Evaluate a statistical query on the database without revealing the values of individual entries
 - Many variations

A Couple of Observations

- In all cases, we are dealing with distributed multi-party protocols
 - A protocol describes how parties are supposed to exchange messages on the network
- All of these tasks can be easily computed by a trusted third party
 - The goal of secure multi-party computation is to achieve the same result without involving a trusted third party

How to Define Security?

Must be mathematically rigorous

- Must capture <u>all</u> realistic attacks that a malicious participant may try to stage
- Should be "abstract"
 - Based on the desired "functionality" of the protocol, not a specific protocol
 - Goal: define security for an entire class of protocols

Functionality

K mutually distrustful parties want to jointly carry out some task

Model this task as a function



 Assume that this functionality is computable in probabilistic polynomial time

Ideal Model

- Intuitively, we want the protocol to behave "as if" a trusted third party collected the parties' inputs and computed the desired functionality
 - Computation in the ideal model is secure by definition!

$$A \xrightarrow{\begin{array}{c} x_1 \\ f_1(x_1, x_2) \\ \end{array}} \xrightarrow{f_1(x_1, x_2)} F_2(x_1, x_2) \xrightarrow{f_2(x_1, x_2)} B$$

Slightly More Formally

A protocol is secure if it emulates an ideal setting where the parties hand their inputs to a "trusted party," who locally computes the desired outputs and hands them back to the parties

[Goldreich-Micali-Wigderson 1987]



Adversary Models

Some of protocol participants may be corrupt

• If all were honest, would not need secure multi-party computation

Semi-honest (aka passive; honest-but-curious)

• Follows protocol, but tries to learn more from received messages than he would learn in the ideal model

Malicious

• Deviates from the protocol in arbitrary ways, lies about his inputs, may quit at any point

For now, we will focus on two-party protocols

Correctness and Security

How do we argue that the real protocol "emulates" the ideal protocol?

Correctness

• All honest participants should receive the correct result of evaluating function f

– Because a trusted third party would compute f correctly

Security

- All corrupt participants should learn no more from the protocol than what they would learn in ideal model
- What does corrupt participant learn in ideal model?
 - His input (obviously) and the result of evaluating f

Simulation

- Corrupt participant's view of the protocol = record of messages sent and received
 - In the ideal world, view consists simply of his input and the result of evaluating f
- How to argue that real protocol does not leak more useful information than ideal-world view?
- Key idea: simulation
 - If real-world view (i.e., messages received in the real protocol) can be simulated with access only to the ideal-world view, then real-world protocol is secure
 - Simulation must be <u>indistinguishable</u> from real view

SMC Definition (First Attempt)

Protocol for computing f(*,*) betw. A and B is secure if there exist efficient simulator algorithms S_A and S_B such that

Correctness: for all input pairs, prot. output is correct.

- Intuition: outputs received by <u>honest</u> parties are indistinguishable from the correct result of evaluating f
- ◆ Security: view_A(real protocol) \approx S_A (gets to query ideal functionality) view_B(real protocol) \approx S_B (gets to query ideal functionality)
 - Intuition: a <u>corrupt</u> party's view of the protocol can be simulated from its input and output
- This definition does not work! Why?

Randomized Ideal Functionality

Consider a coin flipping functionality

f()=(-,b) where b is random bit

- f() flips a coin and tells B the result; A gets no output
- The following protocol "implements" f()
 - 1. A chooses bit b randomly
 - 2. A sends b to B
 - 3. B outputs b
- It is obviously insecure (why?)

Yet it is correct and simulatable according to our attempted definition (why?)

SMC Definition

Protocol for computing f(*,*) betw. A and B is secure if there exist efficient simulator algorithms S_A and S_B such that:

Correctness: for all input pairs, prot. output is correct.
Security:

(view_A(real prot), output_B(real prot)) \approx (S_A, y_B) (view_B(real prot), output_A(real prot)) \approx (S_B, y_A)

- Intuition: if a corrupt party's view of the protocol is correlated with the honest party's output, the simulator must be able to capture this correlation
- Does this fix the problem with coin-flipping f?

Oblivious Transfer (OT)

[Rabin]

1981]

- A inputs two bits, B inputs the index of one of A's bits
- B learns his chosen bit, A learns nothing
 - A does not learn <u>which</u> bit B has chosen; B does not learn the value of the bit that he did <u>not</u> choose
- Generalizes to bitstrings, M instead of 2, etc.

Yao's Protocol

Yao's Protocol

Compute any function securely

• ... in the semi-honest model

First, convert the function into a boolean circuit



1: Pick Random Keys For Each Wire

- Next, evaluate <u>one gate</u> securely
 - Later, generalize to the entire circuit
- Alice picks two random keys for each wire
 - One key corresponds to "0", the other to "1"
 - 6 keys in total for a gate with 2 input wires



2: Encrypt Truth Table

Alice encrypts each row of the truth table by encrypting the output-wire key with the corresponding pair of input-wire keys



3: Send Garbled Truth Table

Alice randomly permutes ("garbles") encrypted truth table and sends it to Bob



4: Send Keys For Alice's Inputs

Alice sends the key corresponding to her input bit

• Keys are random, so Bob does not learn what this bit is



5: Use OT on Keys for Bob's Input

Alice and Bob run oblivious transfer protocol

- Alice's input is the two keys corresponding to Bob's wire
- Bob's input into OT is simply his 1-bit input on that wire



6: Evaluate Garbled Gate

Using the two keys that he learned, Bob decrypts exactly one of the output-wire keys

- Bob does not learn if this key corresponds to 0 or 1
 - Why is this important?



7: Evaluate Entire Circuit

In this way, Bob evaluates entire garbled circuit

- For each wire in the circuit, Bob learns only one key
- It corresponds to 0 or 1 (Bob does not know which)
 - Therefore, Bob does not learn intermediate values (why?)



Bob tells Alice the key for the final output wire and she tells him if it corresponds to 0 or 1

• Bob does not tell her intermediate wire keys (why?)

Brief Discussion of Yao's Protocol

Function must be converted into a circuit

- For many functions, circuit will be huge
- If m gates in the circuit and n inputs, then need
 4m encryptions and n oblivious transfers
 - Oblivious transfers for all inputs can be done in parallel
- Yao's construction gives a <u>constant-round</u> protocol for secure computation of <u>any</u> function in the semi-honest model
 - Number of rounds does not depend on the number of inputs or the size of the circuit!