

UNIVERSITY OF BRISTOL



BACKGROUND CHAPTER

---

# Secure two party computation

---

---

AUTHOR: NICHOLAS TUTTE (NT1124)

SUPERVISOR: PROF. NIGEL SMART

# 1 Background Chapter

## 1.1 Problem definition

Secure multi-party computation(SMC) is a fundamental problem in Cryptography. We have a set of parties who wish to cooperate to compute some function on inputs distributed across the parties. However, these parties distrust one another and do not wish their inputs to be known to the other parties. We shall be focusing on the case where there are only two parties(S2PC), but most two party approaches can be generalised to the multi-party case.

A commonly used example is the Millionaires problem. A group of rich persons wish to find out who among them is the richest, but do not wish to tell each other how much they are worth. Here the parties are the rich (and somewhat vain) individuals. Their inputs are their net worths and the function will return the identifier of the individual with the highest input. Finally no party should be able to divine anything about another's inputs, apart from what the parties can infer from their own input and the output.

Whilst this is not exactly an inspiring application, it does explain convey the problem concisely. We shall cover further applications later in 1.4.

Throughout we will assume that we can establish a secure and authenticated channel of communication to all other parties in the computation. That is we assume communications between two parties cannot be eavesdropped upon or altered, and that we can detect attempts to impersonate another party.

## 1.2 Formal ideal model

There are three main properties that we wish to achieve with any SMP protocol,

- Privacy, the only knowledge parties gain from participating is the output.
- Correctness, the output is indeed that of the intended function.
- Independence of inputs, no party can choose its inputs as the function of other parties inputs.

In this sense we define the goal of an adversary to compromise any one of these properties.

We compare any protocol to the *ideal* execution, in which the parties submit their inputs to a universally trusted and incorruptible external party who then computes the value of the function and returns the output to the parties. We say that the protocol is secure if no adversary can attack the protocol with more success than they can achieve than the ideal model.

### 1.3 Security levels

Having established what goals the adversary wishes to achieve and how we can measure if said adversary has a valid attack, we next deal with the question of the capabilities of the adversary. We use three main models to describe the threat level of the adversary.

#### 1.3.1 Semi-honest Adversary

The Semi Honest(SH) adversary is the weakest adversary, and requires the strongest limitations on their capabilities. The SH adversary has also been referred to as “honest but curious”, because in this case the adversary is not allowed to deviate from the established protocol in any way (i.e. they play fair/are honest), but at the same time they will do their best to compromise one of the aforementioned security properties by examining the data they have legitimate access to. This is in some ways analogous to the classic “passive” adversary.

#### 1.3.2 Malicious Model

The Malicious adversary is allowed to employ any polynomial time strategy and is not bounded by the protocol (they can run arbitrary code instead), but we assume that for the adversary to be successful they must be able to guarantee they will not be detected. This is in some ways analogous to the classic “active” adversary.

#### 1.3.3 Covert Model

The Covert adversary model is very similar to the Malicious model, again bounded by polynomial time with freedom to ignore the protocol, however in this case the adversary is willing to take a risk and is happy with avoiding detection a certain percentage of the time. Effectively in this case the adversary is playing a cost-benefit trade off and to achieve security we need ensure the adversary’s attack is detected with a high enough probability that the adversary will not consider it worth the risk.

We call the probability that such an adversary will be caught the “deterrent probability”, usually denoted using  $\epsilon$ .

### 1.4 Applications

At first it might appear that SMC lacks applications beyond those like the trivial example provided in 1.1. In fact as this is often the first example given many dismiss SMC as of limited usefulness or as a pointless toy. Here we shall provide a number of other applications either already in use or in development.

#### 1.4.1 Secret Auctions - Danish Beets

In Denmark a significant number of farmers are contracted to grow sugar beets for Danisco (a Danish bio-products company). Farmers can trade contracts amongst themselves (effectively sub-contracting the production of the beets), bidding for these sub-contracts is done via a “double auction”. Farmers do not wish to expose their bids as this gives information about their financial state to Danisco and so refused to accept Danisco as a trusted auctioneer, similarly all other parties (e.g. Farmer union) already involved are in some way disqualified. Rather than rely on a completely uninvolved party like an external auction house the farmers use an SMC-based approach described in

[1]. Since 2008 this auction has been run multiple times

As far as team behind this auction are aware this is the first large scale application of SMC to a real world problem, this application example in particular is important as it is a concrete practical example of SMC being used to solve a problem demonstrating this is not just a Cryptological gimmick.

#### 1.4.2 Database queries

Imagine the case where one party holds a database, and the other wishes to perform a query upon this database. However, for whatever reason the querying party does not wish to reveal what their query is, nor does the holder of the database wish to give the database to the querier. This particular problem and related problems have attracted significant attention in the academic community.

#### 1.4.3 Distributed secrets

Consider the growing use of physical tokens in user authentication, e.g. the RSA SecurID. When each SecurID token is activated the seed generated for that token is loaded to the relevant server (RSA Authentication Manager), then when authentication is needed both the server and the token compute ‘something’ using the aforementioned seed. However, this means that in the event of the server being breached and the seed being compromised the physical tokens will need to be replaced. Clearly this is undesirable, being both expensive both in terms of up front costs and reputation.

In the above scenario we clearly need to store the secret(the seed) somewhere, but if we can split the seed across multiple servers and then get the servers to perform the computation as a SMC problem (where each server’s input is their share of the secret, the output the value to compare to the token’s input) then we can increase the cost to an attacker, as they will now have to compromise multiple servers. Such a service is in development by Dyadic Security (full disclosure, my supervisor Nigel Smart is a co-founder of Dyadic).

#### 1.4.4 AES Encryption

A classic test/benchmark for any general S2PC protocol is to perform an AES encryption where the message and key are held by two parties, so  $P_A$  can input a message to be encrypted,  $P_B$  without knowing the message will encrypt it using it’s key, returning the encrypted ciphertext to  $P_A$  and at no point did  $P_A$  know what key was used, nor  $P_B$  what the plaintext was.

### 1.5 Yao Garbled Circuits

Yao garbled circuits are one of the primary avenues of research into Secure multi-party computation. The concept being that we translate the function to be

### 1.6 Oblivious Transfer

## References

- [1] Multiparty Computation Goes Live,  
Peter Bogetoft et al. (2008)