

T-79.514 Special Course on Cryptology

# Seminar 10: Secure Approximate Matching

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## Motivation

- A scenario: Alice wants to compare her DNA against a DNA DB with known genetic diseases  $\Rightarrow$  privacy concerns!
- Need for privacy in e.g. e-commerce, banking/health/etc. records
- In many cases exact matching is not possible
- Exact matching well-studied, approximate not so much
- High interest in *efficient* protocols (MPC too general)

## Overview of the Lecture

- Secure Database Access (SDA)
- SDA in Different Models and Metrics
- Overview of Protocols for the Models
- More In-Depth Look at one Protocol

Based on *W. Du, M.J. Atallah. Protocols for Secure Remote Database Access with Approximate Matching*, appeared in ACM CCS 2000.

# Secure Database Access (SDA)

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The SDA Problem:

Alice has a string  $q$ , and Bob has a database of strings  $T = \{t_1, \dots, t_N\}$ . Alice wants to know whether there exists a string  $t_i \in T$  that *matches*  $q$ . Give a protocol that accomplishes this without revealing to Bob neither (i)  $q$  nor (ii) the found match.

- The answer depends on whether exact or approximate PM is considered
- Depending on the model, the result can be either the closest match or the distance to the closest match

# Metrics

Let  $a = (a_1 \dots a_n)$ ,  $b = (b_1 \dots b_n)$  be two strings. Possible metrics are:

- $\sum_{i=1}^n |a_i - b_i|$  (e.g. in image processing)
- $\sum_{i=1}^n (a_i - b_i)^2$  (e.g. in image processing)
- $\sum_{i=1}^n f(a_i, b_i)$  ( $f$  a function)
- *edit distance* (e.g. in string matching)
- # of indices in which  $a$  and  $b$  differ, etc.

## Models: Overview

- Database  $T$ , possessed by Bob
  - ★ Number of entries (strings)  $N$
  - ★ Each string of length  $n$
  - ★ Each string over an alphabet of size  $m$  (might be infinite)
- Four models, differences in
  - ★ whether  $T$  is private;
  - ★ who owns  $T$ ; and
  - ★ who may query  $T$ .

## Models: PIM

### **Private Information Matching model (PIM).**

- Alice has a query string  $q$ , and wants to know  $\text{Match}(q, T)$  without revealing  $q$  nor  $\text{Match}(q, T)$  to Bob.
- Bob, the *sole* possessor of  $T$ , doesn't want to reveal any  $t_i \in T$  to Alice except what can be derived from  $\text{Match}(q, T)$ .
- Alice has to query  $T$  through Bob.

## Models: PIMPD

### **Private Information Matching from Public Database model (PIMPD).**

As PIM, but

- $T$  is public
- the privacy concerns is that Alice doesn't want to reveal  $q$  nor  $\text{Match}(q, T)$  to Bob.

## Models: SSO

### **Secure Storage Outsourcing model (SSO):**

- The owner of  $T$  is Alice, but  $T$  has been outsourced to Bob (e.g. for storage space reasons).
- Alice wants to query  $T$  without revealing  $T$  nor  $q$  to Bob.

## Models: SSCO

### **Secure Storage and Computing Outsourcing model (SSCO):**

SSO with the following extension:

- any individual may query  $T$
- Alice should be aware of any such queries.
- The individual making the query should learn the distance of the closest match from the query, while this should be kept secret from Alice.

## Overview of Results

Model	Metrics	CC	3rd ?
PIM	$\sum_{i=1}^n (a_i - b_i)^2$	$\mathcal{O}(nN)$	yes
	$\sum_{i=1}^n  a_i - b_i $	$\mathcal{O}(nWN)$	yes
	$\sum_{i=1}^n f(a_i, b_i)$	$\mathcal{O}(mnN)$	yes
SSO	$\sum_{i=1}^n (a_i - b_i)^2$	$\mathcal{O}(n)$	no
SSCO	$\sum_{i=1}^n (a_i - b_i)^2$	$\mathcal{O}(n^2)$	yes

- $W$  an accuracy parameter (in a Monte Carlo – based protocol)
- PIMPD is a special case of PIM  $\Rightarrow$  same protocols applicable
- Third party needed for computing scalar products  $x \cdot y$  of Alice's  $x$  and Bob's  $y$ .

## Protocol for SSO: Preliminaries

Idea: pick a random matrix and disguise  $T$  before outsourcing. Do the same for  $q$ .

- Let  $Q$  be an  $(n + 3) \times (n + 3)$  random invertible matrix
- Let  $R$ ,  $R_A$  and  $R_i$ ,  $i \in \{1, \dots, N\}$ , be random numbers, private to Alice
- For each string  $t_i = t_{i,1} \dots t_{i,n} \in T$ , we have a vector  $\mathbf{t}_i = (\sum_{k=1}^n t_{i,k}^2 + R - R_i, t_{i,1}, \dots, t_{i,n}, 1, R_i)$  of length  $n + 3$
- In  $T'$ , the outsourced version of  $T$ , we have the entry  $\mathbf{t}'_i = Q\mathbf{t}_i^T$

# Protocol for SSO

## 1. Alice

- generates  $R_A$ ,
- constructs  $\mathbf{q} = (1, -2q_1, \dots, -2q_n, R_A, 1)$ , and
- sends  $\mathbf{q}\mathbf{Q}^{-1}$  to Bob.

## 2. Bob

- computes  $\text{score}_i = \mathbf{q} \cdot \mathbf{t}_i^T$  for each  $\mathbf{t}'_i \in T'$ ,
- determines  $\arg \min_{i=1}^N \text{score}_i$ , and
- sends  $\mathbf{t}'_i$  to Alice.

## 3. Alice determines the closest match $\mathbf{t}_i = \mathbf{Q}^{-1}\mathbf{t}'_i$ .

## Notes on the Protocols (1/2)

For SSO and SSCO

- Quite similar solutions
- As Carl may also query, calculating  $x \cdot y$  between Alice and Carl brings  $\mathcal{O}(n)$  to communication complexity
- For SSCO the answer is only the distance to the closest match

## Notes on the Protocols (2/2)

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For PIM and PIMPD

- Not reasonable due to high communication complexity
- Similar to computing  $x \cdot y$  for  $\sum_{i=1}^n (a_i - b_i)^2$
- A bit obfuscated Monte–carlo based protocol for  $\sum_{i=1}^n |a_i - b_i|$ , answer is only the distance to the closest match ...
- ... as well as for  $f$
- For  $f$ , predefined *finite* alphabet is required

## In Addition

- No protocol given for edit distance, although it is said that one exists
- The need for a third party problematic; could this be avoided?
- It is proposed that a sublinear dependency w.r.t.  $N$  might be possible