

Secure and Verifiable Outsourcing of Computation and Storage

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Outline

- ▶ Biometric Computation Outsourcing (Tissec '13)
- ▶ Storage Outsourcing (Asiaccs' 13)
- ▶ A General Framework for Private Distributed Computation (NDSS13, CCS'13)

Secure Outsourcing of Biometric Computation

▶ Biometric Data

- Iris, Fingerprint, Face

▶ Privacy

- biometric data are sensitive in nature
- used for authentication purposes (cannot be revoked)

▶ Computation

- test new algorithms for extracting biometric features.
- force computation outsourcing to a distributed environment
- preserve the integrity of outsourced computation

Secure and verifiable outsourcing of biometric computation !

Abstractions for Biometric Research

- ▶ Four abstractions can aid biometric research
 - Select(R): select a set of images based on requirements
 - gender, age, color
 - produce a image set S
 - Transform(S): convert each image in S to extract features.
 - All-Pairs(S, F): compare each pair of items in S using a function F.
 - produce a matrix M, where $M[x][y] = F(S[x], S[y])$
 - Quality(M, D): Reduce M into a metric D that represents quality.
 - gather a distance distribution in matrix M

Contribution

- ▶ Provide a secure verification mechanism for biometric computation
 - All-Pairs
 - Quality
- ▶ Apply the technique for three different distance functions F
 - Hamming Distance (for iris)
 - Euclidean Distance (for face)
 - Set Intersection Cardinality (for fingerprint)
- ▶ Provides rigorous security analysis and performance evaluation results

Problem Formulation

► All-Pairs Computation:

- given two biometric sets **S1** and **S2**, compute distance between every pair of items $x \in S1$ and $y \in S2$ and outputs a distance matrix **M**.

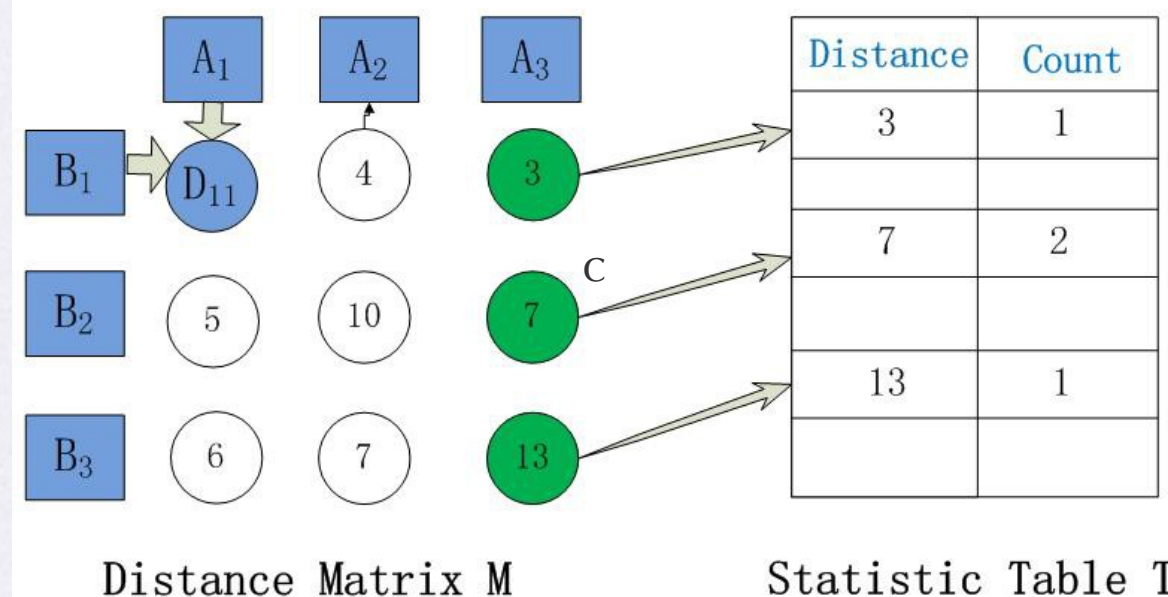
► Quality Computation

- post-process the distances in **M** to compute distribution information for all possible distances.

► For example,

- $S1 = \{A1, A2, A3\}$
- $S2 = \{B1, B2, B3\}$
- $F = \text{Hamming Distance}$

$$D_{ij} = \text{Distance Function}(A_i, B_j)$$



How do we achieve privacy?

- ▶ **Step 1:** client sends each biometric item to the server in a secret form
 - secret shared
- ▶ **Step 2:** client sends each distance in the statistics table to server.
- ▶ **Step 3:** server performs All-Pairs computation securely
- ▶ **Step 4:** server obviously updates the statistics table.
- ▶ **Step 5:** server sends the statistics results to client in a secret format
- ▶ **Step 6:** client reconstructs the results, and verifies its correctness.

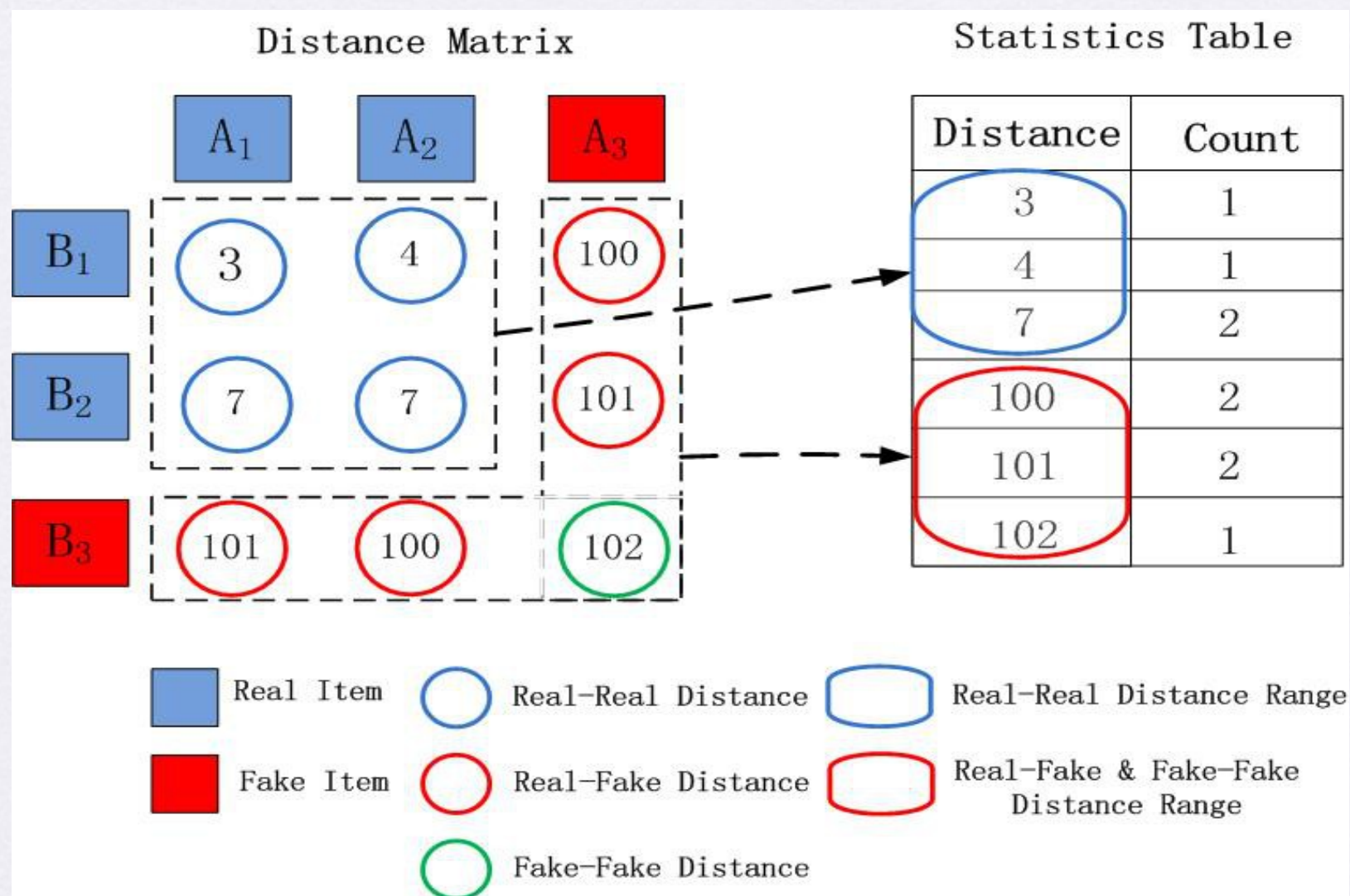
Distance	Count
d_0	c_0
d_1	c_1
.	
.	
d_{L-2}	c_{L-2}
d_{L-1}	c_{L-1}

```
[d] := dist([x], [y])
for i = 0, ..., (L-1)
  [b] := ([d]  $\stackrel{?}{=}$  [di])
  [ci] := [ci] + [b]
```

How to verify the distribution information?

► Informal answer

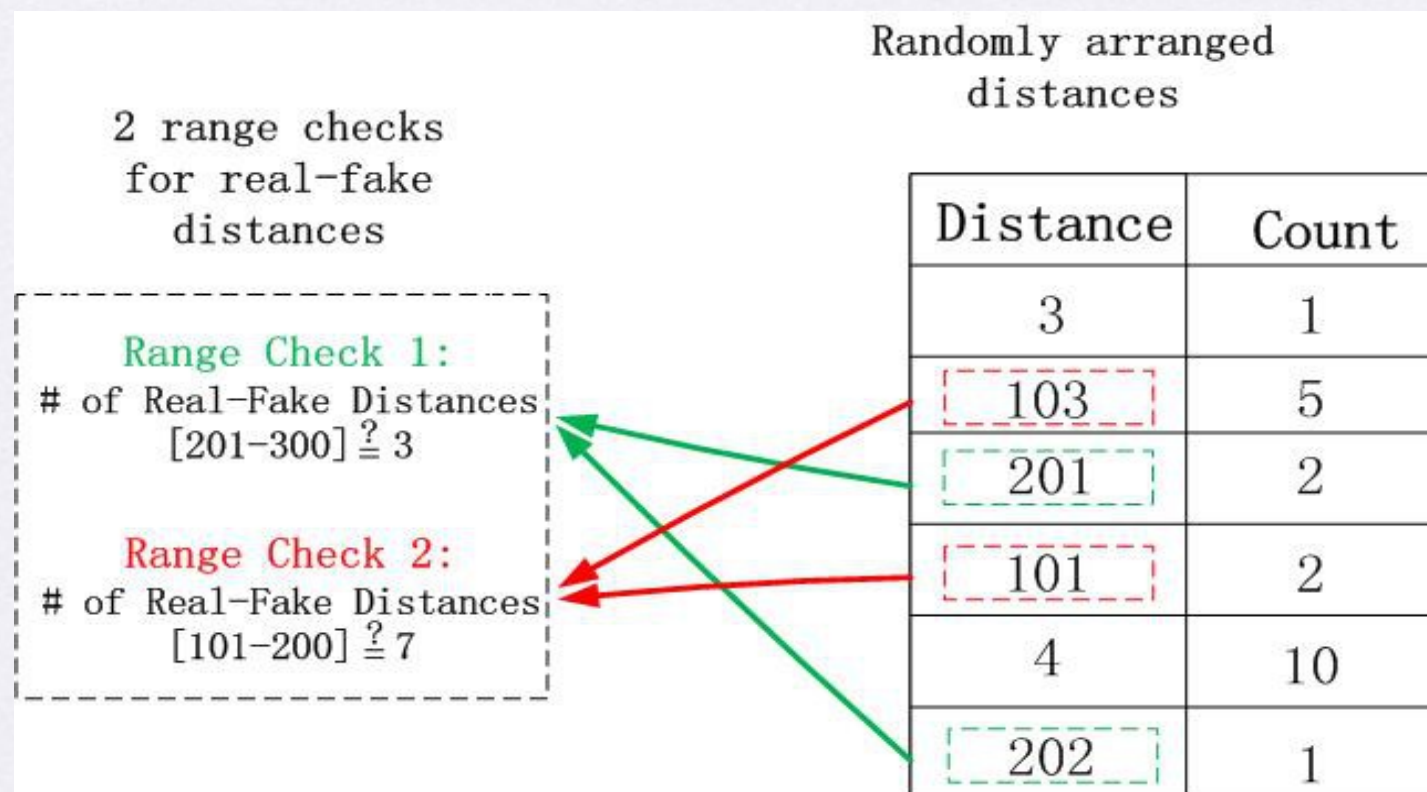
- insert fake biometric items into each data set
- separate the distance between real and real items from the ones between real and fake items.



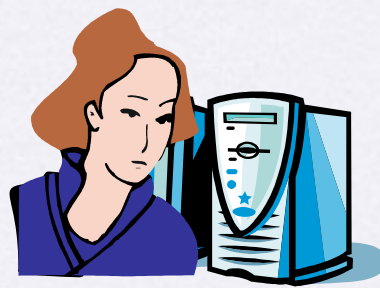
Minimize the server's success rate

► Strategy

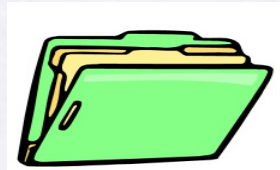
- randomize the distances in the statistics table
- check a large number of ranges for real-fake distances



Storage Outsourcing



Complete?



Outsource.

Read

Write



**SafeFile Inc.
Seattle, USA**

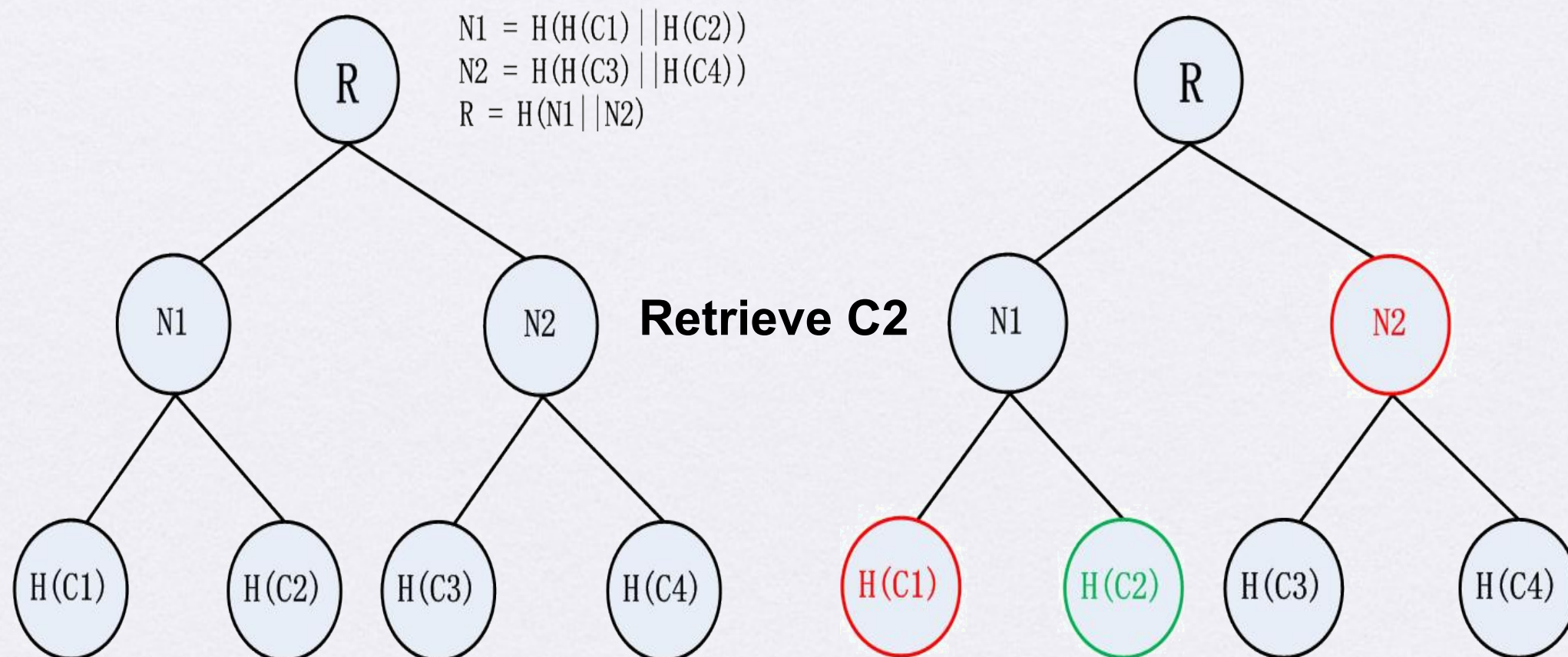
Previous Works

- ▶ POR (Proofs of Retrievability) and PDP (Proofs of Data Possession)
 - partition a collection of data into data blocks
 - store the blocks together with metadata
 - issue integrity verification queries periodically
 - use the metadata to ensure data integrity
- ▶ DPDP (our focus)
 - allow to issue dynamic operations

Previous Works

► Authenticated Data Structures

- Merkle Hash Tree [Oprea et al. 07, Wang et al. 09, Popa et al. 11]
- Skip List [Goodrich et al. 08, Heitzmann et al. 08, Erway et al. 09]



Contribution

►Pros:

- efficiency in communication and computation
- support for revision control and range operations
- size of data structure independent of outsourced file

►Cons:

- client needs to maintain the data structure by himself

Balanced Update Tree

- ▶ A node represents a dynamic operation
 - Insert [50, 60]
 - except for the retrieval operation
- ▶ Each node contains the following attributes
 - **[L, U]**: lower and upper bound of block range
 - **Op**: operation type (e.g., insert, delete, modify)
 - **GID**: identification of different insert or delete ops
 - **Version**: identification of different versions of same block

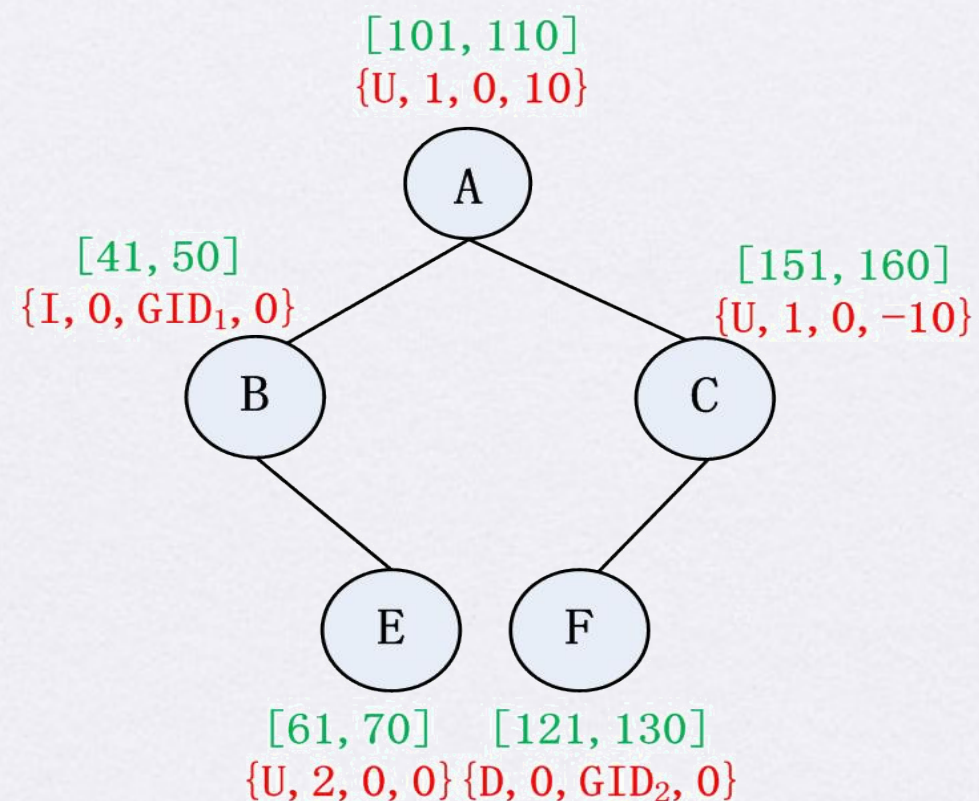
Balanced Update Tree

- ▶ How to organize the nodes within the structure?
 - the nodes are ordered based on their ranges $[L, U]$
 - all the nodes within a left subtree of a node have ranges fall into $[0, L-1]$
 - all the nodes within a right subtree of a node have ranges fall into $[U+1, +\infty]$.

▶ Balance the tree if necessary

▶ For example,

- $[L, U]$
- $\{Op, V, ID, R\}$



How to verify different operations?

► Update (no verification)

- modify the update tree and obtain the correct attributes
- compute a tag $t_i = \text{Mac}(d_i || \text{Op} || \text{GID} || \text{index} || V)$ for each updated block d_i .
- outsource tags together with data blocks

► Retrieve (do verification)

- the server returns (d_i, t_i) for each retrieved block
- the client obtains the values of Op' , ID' , Index' , and V' for each block from the update tree.
- the client verifies $t_i = \text{Mac}(d_i || \text{Op}' || \text{GID}' || \text{index}' || V')$

Performance

► Computation

Volume	Max Offset	Operations (x10 ⁶)	MHT (sec)	SL (sec)	Utree (sec)
Project -0	170GB	4.2	6400	21000	8.4
Project -1	880GB	24	29000	95000	1200
Project -2	880GB	29	48000	120000	2200
Project -3	240GB	2.2	670	2300	3
Project-4	240GB	6.5	3400	12000	150

Private Distributed Computation

► Goal:

- build a computational framework that privacy preserving execution on any functionality or program.

► Features:

- support as wide range of functionalities as possible
 - data Structures
 - algorithms
- Support for all major types of computation
- achieves both theoretical security and **efficiency**
 - reduce the number of interactive ops or rounds
 - explore the use of parallelism

Framework

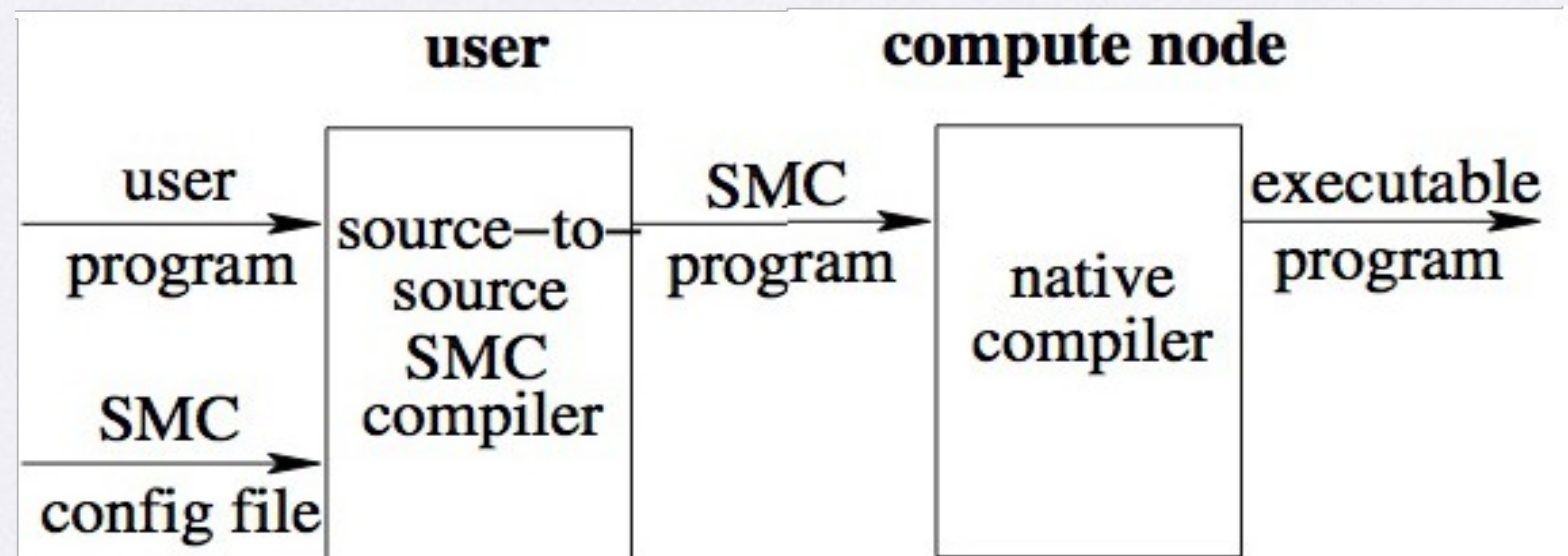
► How it works?

- the compiler converts a C extension program into its secure implementation
 - based on linear secret sharing
- the resulting program will be compiled by the native compiler and run by a number of distributed computational nodes
- the compiler includes several aid programs
 - pre-process private inputs
 - recover the final outputs

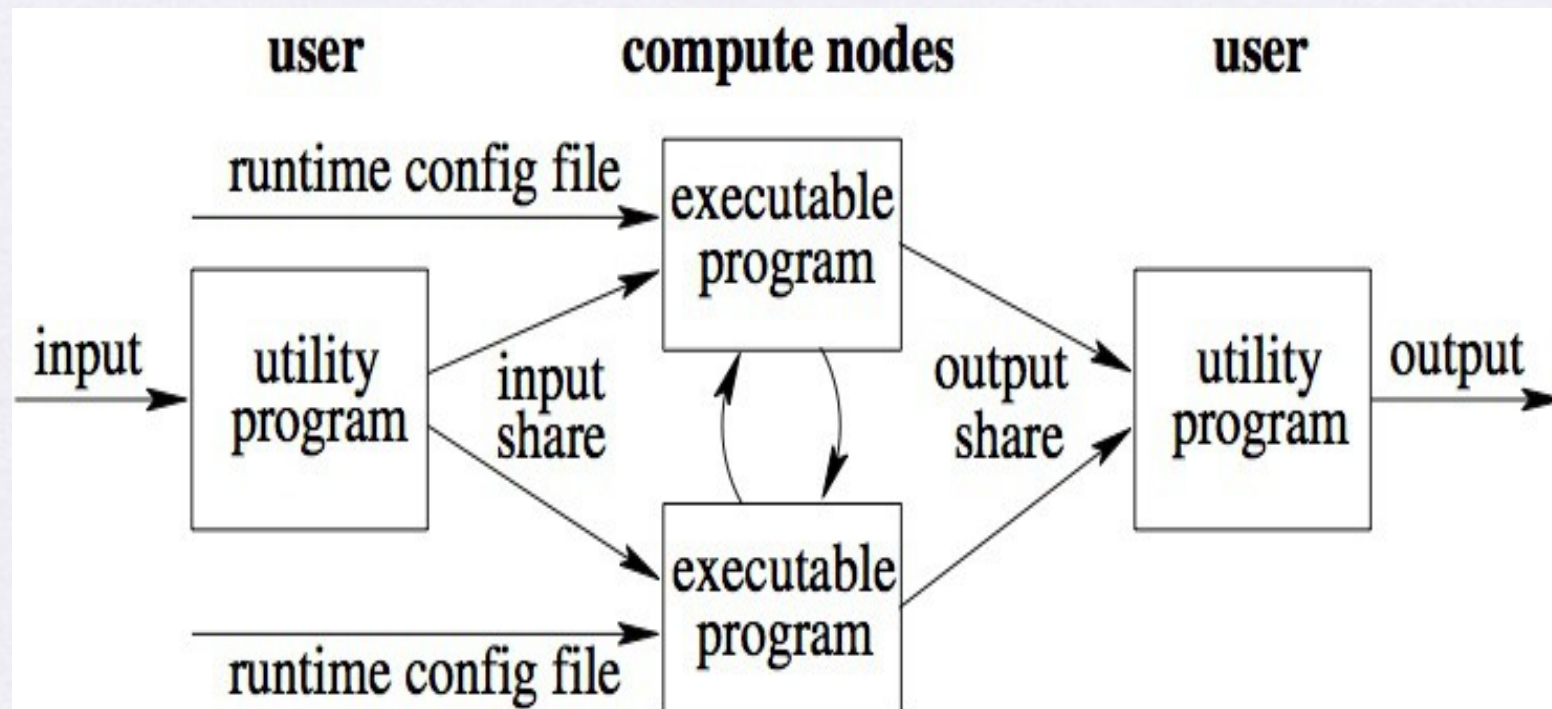
Framework

► How it works?

Compile:



Execution:



Sample Program

Source Program

```
private int t;  
  
if (t>0)  
  for (i=0; i<n; i+=5)  
    a[i]=a[i]+1;
```

Compiled Program

```
mpz_t t;  
  
mpz_t cond1;  
  
mpz_t tmp1;  
  
smc_gt(t,0,cond1);  
  
for (i=0; i<n; i+=5) {  
  tmp1=a[i];  
  a[i]=a[i]+1;  
  a[i]=cond1*a[i]+(1-cond1)*tmp1;  
}
```

Performance

Experiment (ms)	PICCO (LAN)	PICCO (WAN)	Sharemind (LAN)	Sharemind (WAN)
Matrix Multiplication (8 x 8)	0.45	32.1	168	376
Merge-sort (64 elements)	649.6	12080	15145	47636
AES (128 bit)	35.1	3179	652	N/A
FingerPrint (20 minutiae)	830	74704	24273	75820

Thank you !
