





CSE/CEN 598

Hardware Security & Trust

Secure Computation Approaches: Homomorphic Encryption

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Secure Computation Approaches

Multi-Party Computation (MPC) Pros - Low compute requirements - Easy to accelerate - Provably secure - Supports multiple threat models - Easy to map existing algorithms Cons - High communication costs - High latency - Information theoretic proofs are weaker than PKE ones	Fully Homomorphic Encryption (FHE) Pros - Very low communication costs - Requires a single round of communications, i.e., "fire and forget" - Useful when one side is limited in compute / memory / storage - Provably secure – relies on strength of PKE Cons - Very high computational requirements - Harder to accelerate - Mapping existing algorithms to FHE may be difficult	Trusted Execution Environments (TEE) Pros - No communication required - Trivial to accelerate - Great support for existing software Cons - Weaker security guarantees - Cannot stop determined adversaries - Historically plagued by vulnerabilities and breaches - Long term deployment is difficult – TEE's can 'run out' of entropy / CRPs, etc.
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Outsourced Computation

- Cloud storage and computing have many advantages
 - The rise of connected and sensor-based devices have led to cloud computing being used as a commodity technology service



Google Cloud



Microsoft Azure



Alibaba Cloud



IBM Cloud



aws



SECURE?

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Outsourced Computation

- One of the key issues with cloud-based computation is data privacy
 - Sensitive data is stored and computed over the cloud, which at most times, is a shared resource
 - We currently have more than 2,500 cloud vulnerabilities
 - 150% increase just in the last five years

Number of cloud vulnerabilities tracked by IBM Security X-Force

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Outsourced Computation

2017

Verizon

Data of at least 6 million users are exposed due to "human error" from misconfigured Amazon S3 cloud server.

2019

Capital One

Former worker of AWS illegally accessed into Capital One's AWS cloud server and leaked personal data of 106 million people.

2019

Best Western

179 GB database included personal data of members of the U.S. government, military, and DHS are exposed via unsecured Elastic-search database hosted on AWS.

2020

Alibaba

Alibaba Cloud's staffs obtained client contact information and leaked it to a third-party partner without consent.

2021

Microsoft

A flaw in Microsoft's Azure Cosmos DB database left more than 3,300 Azure customers (including many Fortune 500 companies) open to complete unrestricted access by attackers.

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Homomorphic Encryption

- What is Homomorphic Encryption (HE)?
 - Encryption scheme that allow computation on encrypted data without decryption
 - Homomorphic encryption can be used along with cloud services to perform computations on encrypted data, guaranteeing data privacy

Cloud server processes the data in encrypted form and returns it to the client

Third-party cloud service provider

Homomorphic encryption-based data processing platform

Client encrypts data with their own key and sends it to cloud

Client decrypts and reads the computation results

Data owner

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Homomorphic Computation

- Homomorphic computation is a form of computation that allows computation on encrypted data
 - The result of this encrypted data processing itself is encrypted
 - But when the result is decrypted, it should match the output of the program if it runs directly on the encrypted input data

```

graph LR
    Alice --> EncX[Enc(x)]
    EncX --> Bob
    subgraph Bob
        BobText[Bob has the program P]
        EncP[Enc(P)]
    end
    EncP --> Alice
    EncX --> EncP
    EncP --> Result[P(Enc(x))]
    Result --> Alice
      
```

The diagram illustrates the homomorphic computation process. Alice sends an encrypted input $Enc(x)$ to Bob. Bob has a program P and its encrypted version $Enc(P)$. Bob runs the program P on the encrypted input $Enc(x)$ to produce an encrypted result $P(Enc(x))$, which is then sent back to Alice.

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Overview Homomorphic Encryption (HE)

- An encryption scheme is called homomorphic over an operation '*' if it supports the following
 - $\forall (m_1, m_2) \in M, Enc(m_1) * Enc(m_2) = Enc(m_1 * m_2)$
 - Where Enc is the encryption algorithm and M is the set of all possible messages
- Supporting addition and multiplication operations is sufficient to create an encryption scheme that the homomorphic evaluation of an arbitrary function
 - Any Boolean circuit can be represented using only XOR (addition) and AND (multiplication) gates

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Homomorphic Encryption

- Homomorphic Encryption
 - Is a form of encryption that allows computations to be carried out on ciphertext
 - Generates an encrypted result
 - The result when decrypted matches the result of operations performed on the plaintext
- Formally,
 - $Eval(f, c_0, c_1, \dots, c_n)$
- Example:
 - $Enc(key, 2) = \$, Enc(key, 3) = \%$
 - $Eval(+, \$, \%) = \#$
 - $Dec(key, \#) = 5$

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Simple Illustrative Example

- Function to compute
 - f is a simple addition
 - Such that $y = f(x_1, x_2, \dots, x_n)$
 - $y = x_1 + x_2 + \dots + x_n = \sum_{i=1}^n x_i$
 - n is number of terms
- We will define our encryption function as
 - $Enc(x) = (x + p) * q$ where p and q determine the key $k(p, q)$ and it is private
- We define the decryption function as
 - $Dec(Y, n) = Y/q - n * p$

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Simple Illustrative Example

- Homomorphically compute
 - $fHE(5, 9, 3)$
- Computation Process
 - Key Generation
 - Pick p and q
 - $k(7, 2)$
 - Encrypt the terms
 - $Enc(5) = (5+7) * 2 = 24$
 - $Enc(9) = (9+7) * 2 = 32$
 - $Enc(3) = (3+7) * 2 = 20$
 - Evaluation
 - $f(24, 32, 20) = 24 + 32 + 20 = 76$
 - Decrypt the result
 - $Dec(76, 3) = (76/2) + (3*7) = 17$

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Homomorphic Computation

- Fully Homomorphic Encryption (FHE)
 - It was first defined in 1978 under privacy homomorphism
 - For the purpose of searching encrypted data
 - Various approaches
 - Multiplicatively homomorphic by RSA and El Gamal Additively homomorphic by GM and Paillier
 - Quadratic formulas by BGN'05 and GHV'10a
 - Recent major advances
 - First Construction of fully homomorphic encryption by Gentry [2009]
 - Using algebraic number theory - ideal lattices

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Overview Homomorphic Encryption (HE)

- Techniques to compute on encrypted data can be classified in three (3) categories
 - Partially Homomorphic Encryption (PHE)
 - Allowing only one type of operation with an unlimited number of times
 - Somewhat Homomorphic Encryption (SHE)
 - Allowing some types of operations with a limited number of times
 - Fully Homomorphic Encryption (FHE)
 - Allowing an unlimited number of operations with unlimited number of times

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Overview Homomorphic Encryption (HE)

- Techniques to compute on encrypted data can be classified in three (3) categories
 - Partially Homomorphic Encryption (PHE)
 - Unlimited add OR multiplication
 - Somewhat Homomorphic Encryption (SHE)
 - Limited addition AND multiplication
 - Fully Homomorphic Encryption (FHE) $\Leftarrow$ SHE + Bootstrapping
 - Unlimited addition AND multiplication

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Homomorphic Encryption Approaches

- Popular Homomorphic Encryption Schemes
 - TFHE – Fast Fully Homomorphic Encryption - 2016
 - Support homomorphic evaluation on logic gates (AND, OR, NAND, NOT, MUX, etc.)
 - Best for operation on individual bits
 - BGV (Brakerski-Gentry-Vaikuntanathan - 2011) and BFV (Brakerski/Fan-Vercauteren - 2012)
 - Exact arithmetic on vectors of numbers
 - Best for vectorized operation over finite fields
 - CKKS (Cheon, Kim, Kim and Song – 2016)
 - Approximate arithmetic on vectors of numbers
 - Best for vectorized operation over real numbers

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Overview Homomorphic Encryption (HE)

- A Homomorphic Encryption algorithm has four primary operations
 - KeyGen, Enc, Dec, and Eval
- KeyGen, Enc and Dec are essentially not different from their classical tasks in conventional encryption algorithms
 - KeyGen operation generates a secret and public key pair for an asymmetric encryption scheme and a single key for the symmetric encryption scheme
- Eval operation is the true homomorphic encryption specific operation, it takes ciphertexts as input and outputs evaluated ciphertexts
 - Eval performs the function $f()$ over the ciphertexts (c_1, c_2) without seeing the messages (m_1, m_2)
 - The format of the ciphertexts must be preserved after an evaluation process to be decrypted correctly
 - The size of the ciphertext should also be constant to support unlimited number of operations
 - Increase in the ciphertext size will require more resources and will limit the number of operations

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RLWE-Based Homomorphic Encryption

- What LWE – Learning with Error and Ring LWE?
 - Learning with Error is a computation problem that given a set of linear equations, we solve for the secret

$14s_1 + 5s_2 + 15s_3 + 7s_4 \approx 8 + 1 \pmod{17}$
 $3s_1 + 7s_2 + 4s_3 + 12s_4 \approx 16 + 3 \pmod{17}$

$8s_1 + 10s_2 + 11s_3 + 3s_4 \approx 7 + 2 \pmod{17}$
 $6s_1 + 7s_2 + 16s_3 + 2s_4 \approx 3 + 4 \pmod{17}$

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HE Computational Challenges

- Brakerski-Fan-Vercauteren (BFV) scheme as illustrative example
- FV.SH.SecretKeyGen():
 - sample s from a Gaussian distribution, and
 - output
 - $sk = s$
- FV.SH.PublicKeyGen(sk):
 - set $s = sk$,
 - sample a from R_q and e from Gaussian distribution,
 - output
 - $pk[0] = b = \lfloor -(a \cdot s + e) \rfloor$
 - $pk[1] = a$

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HE Computational Challenges

- Brakerski-Fan-Vercauteren (BFV) scheme as illustrative example
- FV.SH.Encrypt(pk, m):
 - encrypt a message $m \in \mathbb{R}_t$
 - compute $t = q/T$
 - sample r_0 from R_2
 - sample r_1 and r_2 from the Gaussian distribution, and
 - return
 - $ct[0] = [b.r_0 + r_2 + t.m]_q$ and
 - $ct[1] = [a.r_0 + r_1]_q$

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HE Computational Challenges

- Brakerski-Fan-Vercauteren (BFV) scheme as illustrative example
- FV.SH.Add(ct₁, ct₂):
 - Compute $ct_1 + ct_2$
 - return
 - $c_0 = [ct_1[0] + ct_2[0]]_q$,
 - $c_1 = [ct_1[1] + ct_2[1]]_q$
- FV.SH.Decrypt(sk, ct):
 - set $s = sk$, and ciphertexts = c_0, c_1
 - compute $\left\lfloor \frac{[c_0 + c_1 s]}{t} \right\rfloor_q$
 - compute $\tilde{m} \pmod{T}$

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HE Computational Challenges

- Brakerski-Fan-Vercauteren (BFV) scheme as illustrative example
- FV.SH.MUL(ct₁, ct₂):
 - Compute $ct_1 * ct_2$
 - return
 - $c_0 = \left\lfloor \frac{[c_{t1}[0] \cdot c_{t2}[0]]}{t} \right\rfloor_q$
 - $c_1 = \left\lfloor \frac{[c_{t1}[0] \cdot c_{t2}[1] + c_{t1}[1] \cdot c_{t2}[0]]}{t} \right\rfloor_q$
 - $c_2 = \left\lfloor \frac{[c_{t1}[1] \cdot c_{t2}[1]]}{t} \right\rfloor_q$
- FV.SH.Decrypt(sk, ct):
 - set $s = sk$,
 - ciphertexts c_0, c_1 and c_2
 - compute $\tilde{m} = (c_0 s^0 + c_1 s^1 + c_2 s^2)_q$
 - compute $\tilde{m} \pmod{T}$

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HE Computational Challenges

- Brakerski-Fan-Vercauteren (BFV) scheme as illustrative example
- Relinearisation challenge
 - Relinearisation is a procedure that takes a degree 2 ciphertext and reduces it again to a degree 1 ciphertext
 - Let $ct = [c_0, c_1, c_2]$ denote a degree 2 ciphertext, then we need to find $ct' = [c'_0, c'_1]$ such that
 - $[c_0.s^0 + c_1.s^1 + c_2.s^2]_q = [c'_0.s^0 + c'_1.s^1]_q$
 - To eliminate $c_2.s^2$ term we need to mask it
 - Masking is done using relinearisation keys/ homomorphism keys/ evaluation keys

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HE Computational Challenges

- Brakerski-Fan-Vercauteren (BFV) scheme as illustrative example
- Noise growth challenge
 - Noise in freshly encrypted ciphertext is given by
 - ct_1 has B amount of noise
 - ct_2 has B amount of noise
 - $SH.Add(ct_1, ct_2)$:
 - Noise growth: $B + B = 2B$
 - $SH.Mul(ct_1, ct_2)$:
 - Noise growth: $B * B = B^2$
 - With L levels of multiplication
 - Noise growth: B^{2^L}

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HE Computational Challenges

- Brakerski-Fan-Vercauteren (BFV) scheme as illustrative example
- Noise growth challenge
 - Decryption will be correct, if
 - Noise $\leq q/4$
 - To perform L levels of multiplication,
 - $B^{2^L} \leq q/4$ which means $q \geq 4B^{2^L}$

For $B = 10$,

L	q	$\log_2 q$	n
1	400	9	1024
2	40000	16	1024
3	400000000	29	2048
4	4000000000000000000	56	2048

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HE Computational Challenges

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- Noise growth challenge
 - Noise in freshly encrypted ciphertext is given by
 - ct1 has B amount of noise
 - ct2 has B amount of noise
 - SH.Add(ct1, ct2):
 - Noise growth: $B + B = 2B$
 - SH.Mul(ct1, ct2):
 - Noise growth: $B * B = B^2$
 - With L levels of multiplication
 - Noise growth: B^{2^L}

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HE Computational Challenges

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For B = 10,

L	q	$\log_2 q$	n
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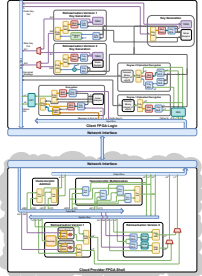
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Arithmetic Hardware Library for Accelerated HE

- A library consisting of modules involved in these HE algorithms
 - Residue Number System (RNS)
 - Chinese Remainder Theorem (CRT)
 - NTT-based polynomial multiplication
 - Modulo Inverse
 - Modulo Reduction
 - Polynomial and scalar operations
 - And others ...



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Arithmetic Hardware Library for Accelerated HE

- Gaussian Noise Sampler

A Post-Quantum Secure Discrete Gaussian Noise Sampler

Sampling Algorithms

Box-Muller Sampling

Ziggurat Sampling

Rejection Sampling

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Homomorphic Encryption Libraries

- Some of the popular homomorphic encryption libraries
 - PALISADE**
 - <https://palisade-crypto.org/>
 - SEAL**
 - <https://www.microsoft.com/en-us/research/project/microsoft-seal/>
 - HElib**
 - <https://homenc.github.io/HElib/>
 - HEAAN**
 - <https://heaan.it/>
 - TFHE**
 - <https://tfhe.github.io/tfhe/>

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Upcoming Lectures

- Other Hardware Security Topics
 - Related Topics
 - Reviews

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