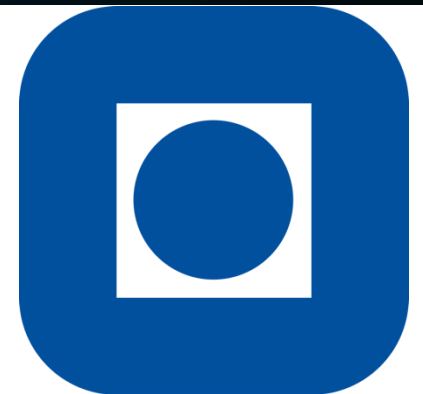
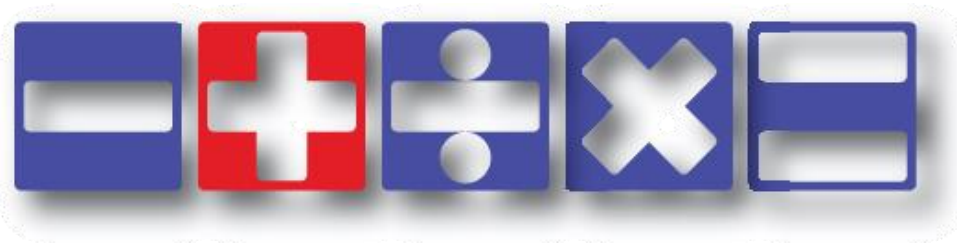


# Arithmetic questions in Fully Homomorphic Encryption

**ARITH 2025**

Anamaria Costache



NTNU

# Introduction

# Fully Homomorphic Encryption

- Allows to compute on encrypted data in a way that matches the computation on the underlying plaintext

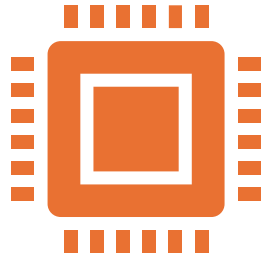
$$f(Enc(m)) = Enc(f(m))$$

- Proposed as a concept in '78, first construction in '09

**On data banks and privacy homomorphisms.** RL Rivest, L Adleman, ML Dertouzos. Foundations of Secure Computation 1978.

**Fully homomorphic encryption using ideal lattices.** Craig Gentry. ACM symposium on Theory of computing, 2009

# Applications



**Countless applications, most easily illustrated with *outsourcing computation***



**Can use FHE to construct many complex privacy-preserving protocols**

Secure voting

Private Set Intersection

Private Information Retrieval

Secure medical computations/ analytics

Secure biometric identification/ verification

...

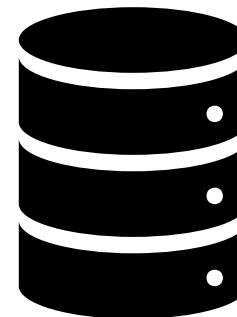
$$(\textcolor{red}{sk}, \textcolor{green}{pk}) \leftarrow \text{KeyGen}(1^\lambda)$$
$$\textcolor{green}{ct} \leftarrow \text{Enc}(\textcolor{red}{m}, \textcolor{green}{pk})$$

$$\textcolor{green}{ct}' \leftarrow f(\textcolor{green}{ct}, \textcolor{green}{pk}) = \text{Enc}(\textcolor{red}{f(m)}, \textcolor{green}{pk})$$



$(\textcolor{green}{ct}, f, \textcolor{green}{pk})$

$\textcolor{green}{ct}'$



$$\textcolor{red}{f(m)} \leftarrow \text{Dec}(\textcolor{green}{ct}', \textcolor{red}{sk})$$

# Privacy-preserving Machine Learning



FHE is particularly well-suited to Privacy-Preserving Machine Learning applications



Enables a client (making the query) to keep their data private



Can also enable the server to **hide their model**



Many challenges here: more on this later!

# Terminology

**Partial**  
homomorphic  
encryption

Can support **one**  
**of** additions or  
multiplications

**Somewhat**  
homomorphic  
encryption

Can support  
both, but a  
**limited** number

**Fully**  
homomorphic  
encryption

**Unlimited**  
number of  
operations

# Technical Introduction



# Learning with Errors

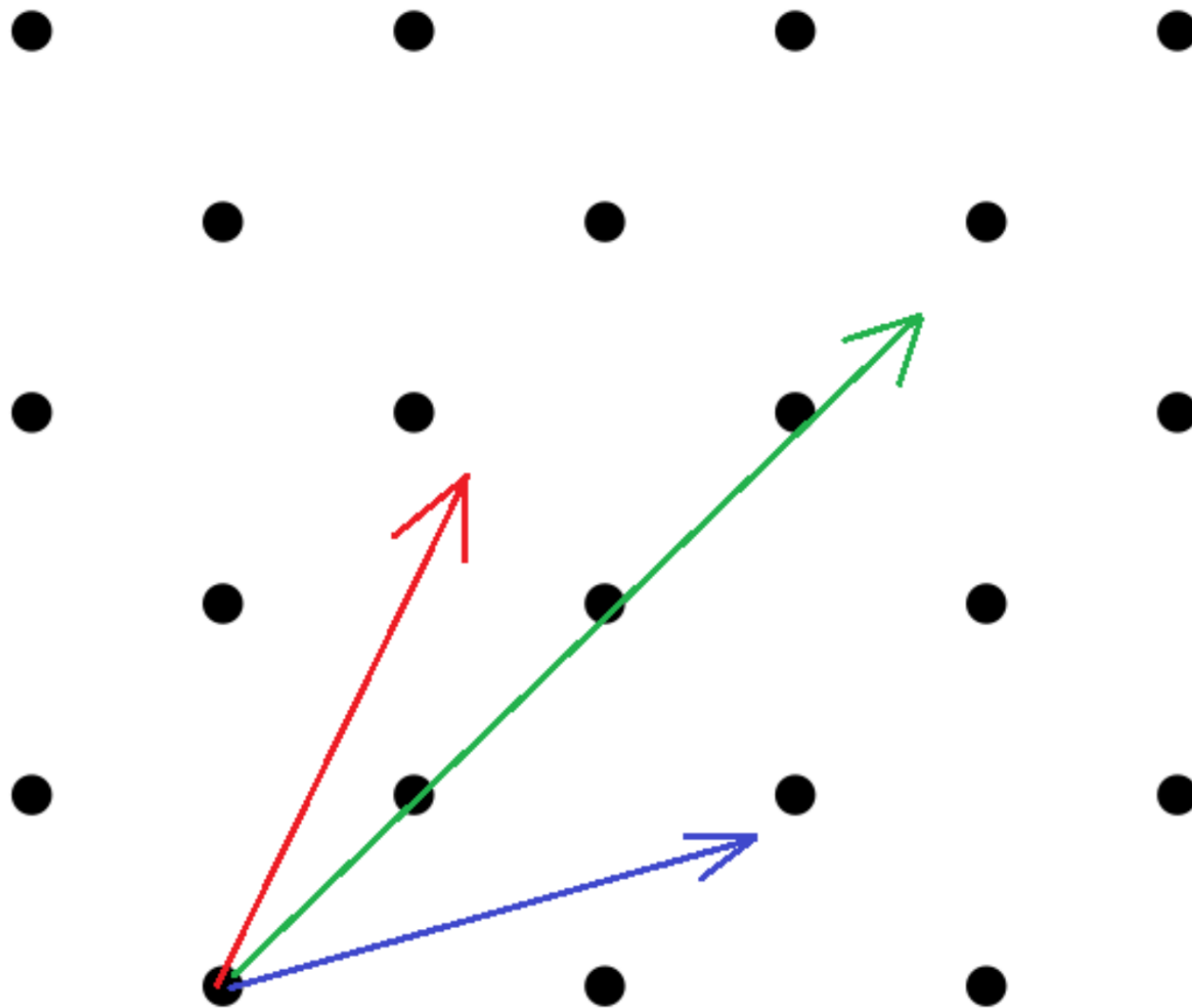
**Definition (Learning with Errors):** Let  $q$  be a positive integer, let  $n$  be a dimension and an error distribution  $\chi$ . For a secret  $s \in \mathbb{Z}_q^n$ , the LWE distribution outputs pairs of the form

$$(a_i, b_i) = (a_i, b_i = \langle a_i, s \rangle + e_i) \in \mathbb{Z}_q^n \times \mathbb{Z}_q,$$

where  $e_i$  are sampled according to  $\chi$  and  $a_i \in \mathbb{Z}_q^n$  are sampled uniformly at random,.

- The LWE **decision** problem is: given pairs of the form  $(a_i, b_i)$ , decide if they were sampled according to the LWE distribution or uniformly at random.
- The LWE **search** problem is: given pairs  $(a_i, b_i)$ , recover the secret  $s$ .
- We will use the *Ring-Learning with Errors* version of this problem, where  $\mathbb{Z}_q^n$  is replaced by a polynomial ring

# Noise in FHE



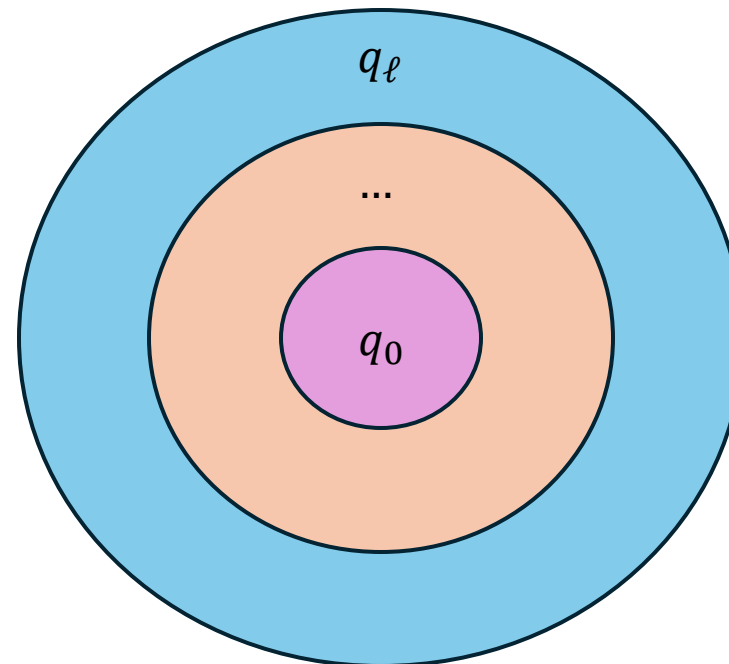
# Two paradigms for handling noise

## Bootstrapping

- Decrypting removes **all** noise
- But that would reveal the message
- Perform decryption function homomorphically, reduce the noise

## Levelled schemes

- “Layer” the ciphertext space into levels
- After each multiplication, divide and round
- Controls the noise growth



# Four generations of schemes

## First generation

- Gentry's original scheme (2009)

## Second generation

- BGV, BFV, (NTRU)
- Good packing capabilities, slow *bootstrapping*, exact computations, additions and multiplications

## Third generation

- GSW, FHEW, TFHE
- Very fast *bootstrapping*, non-linear functions, low packing capabilities

## Fourth generation

- CKKS
- (very) Good packing capabilities, slow *bootstrapping*, additions and multiplications

# Parameters (for this talk)

- We often operate on the space  $\mathcal{R}_q = \mathbb{Z}[x]/(\phi_n, q)$  where  $\phi_n$  is the  $2n$ -th cyclotomic polynomial

$$\phi_n(x) = x^n + 1,$$

for  $n = 2^k$ , for some  $k$ , and  $q$  is the (Ring-)LWE modulus.

- Ciphertexts are of the form  $(c_0, c_1) \in \mathcal{R}_q^2$ .
- Typically,  $q$  is chosen to be a product of  $\ell$  primes  $p_0, \dots, p_{\ell-1}$
- We write  $q_k = \prod_{i=0}^k p_i$
- Each prime  $p_i, i \in \{0, \dots, \ell - 1\}$  is roughly of the same size, and is chosen such that  $p_i \equiv 1 \pmod{2n}$
- In particular, we can use the Chinese Remainder Theorem (CRT) to write the following isomorphism

$$\mathcal{R}_{q_{\ell-1}} \cong \frac{\mathbb{Z}[x]}{\phi_n, p_0} \times \dots \times \frac{\mathbb{Z}[x]}{\phi_n, p_{\ell-1}}.$$

- We are particularly interested in using the **Residue Number System (RNS)**

# Encodings

Mod  $q$  and mod  $t$  operations **do not commute!** The noise typically needs to remain **smaller than  $q$** !

A ciphertext is a pair  $(c_0, c_1)$ , under some secret key  $s$ .

	Lower bits	Upper bits	Approximate
Message encoding	$m + t \cdot e$	$\Delta \cdot m + e$	$m + e$
Decryption	$\left[ [c_0 + sc_1]_q \right]_t$	$\left[ \frac{t}{q} \left[ [c_0 + sc_1]_q \right] \right]_t$	$[c_0 + sc_1]_q$

$\Delta$  is a scaling factor,  $t$  is the plaintext modulus,  $q$  is the ciphertext modulus

$[\cdot]_q$  denotes modular reduction (centered around 0) mod  $q$ ,  $|\cdot|_q$ , “regular” modular reduction

# BFV operations - deep dive

# BFV Encryption

- Let  $t$  be a plaintext modulus, let  $q_\ell$  be as before. Let  $e, e_0, e_1 \leftarrow \chi$  be noise terms sampled from the noise distribution  $\chi$ , and let  $s \leftarrow \mathcal{S}$  be a secret key
- Then, the public key is

$$(p_0, p_1) = pk = [-a \cdot s + e, a]_q$$

- We encrypt a message  $m \in \mathcal{R}_t$  as follows. Let  $v \leftarrow \mathcal{S}$  be an ephemeral key

$$Enc(m, pk) = [\Delta m + vp_0 + e_0, vp_1 + e_1]_q = (c_0, c_1) \in \mathcal{R}_q^2$$

- **Addition** is simple (linear)
  - Coordinate-wise
  - Noise growth is very small

$[\cdot]_q$  denotes modular reduction (centered around 0) mod  $q$ ,  $|\cdot|_q$ , “regular” modular reduction



# Multiplication – the challenges

Multiplication is a more complex operation

A “naïve” approach (simply multiplying the components) leads to unmanageable noise growth, leading to losing the message

Use **base decomposition** instead

# Multiplication I

- **Multiplication** is a more complex operation, and consists of **two** steps. Multiplying the “naïve” way would result in unmanageable noise growth.
- Consider two ciphertexts, encrypted under the same key

$$\begin{aligned} ct &= Enc(m_0, pk) = [\Delta m_0 + v p_0 + e_0, v p_1 + e_1]_q = (c_0, c_1) \in \mathcal{R}_q^2 \\ ct' &= Enc(m_1, pk) = [\Delta m_1 + v' p_0 + e'_0, v' p_1 + e'_1]_q = (c'_0, c'_1) \in \mathcal{R}_q^2 \end{aligned}$$

- **First step:**

$$(d_0, d_1, d_2) = (c_0 \cdot c'_0, c_0 \cdot c'_1 + c'_0 \cdot c_1, c_1 \cdot c'_1)$$

$$\left( \left\lfloor \frac{t \cdot d_0}{q} \right\rfloor, \left\lfloor \frac{t \cdot d_1}{q} \right\rfloor, \left\lfloor \frac{t \cdot d_2}{q} \right\rfloor \right)$$

- This results in a 3-element ciphertext that decrypts under  $s^2$ , rather than  $s$

$$d_0 + s \cdot d_1 + s^2 \cdot d_2 = \Delta m + E,$$

for some noise  $E$ .

# Multiplication II

- We have a 3-element ciphertext that decrypts under  $s^2$ , rather than  $s$   
$$d_0 + s \cdot d_1 + s^2 \cdot d_2 = \Delta m + E,$$
for some noise  $E$ .

- We want to recover a 2-element ciphertext, that decrypts under  $s$ . A simple solution:

$$(d_0 + s^2 d_2, d_1) \text{ decrypts as } (d_0 + s^2 \cdot d_2) + s \cdot d_1 = \Delta m + E.$$

- But:**  $s^2$  is secret! So, we encrypt it as  $(a'' \cdot s + e' + s^2, -a'')$ .
- Now, we can form the ciphertext

$$(d_0 + d_2 \cdot (a'' \cdot s + e'' + s^2), d_1 - d_2 \cdot a'')$$

- Can decrypt as

$$\begin{aligned} & d_0 + d_2 \cdot a'' \cdot s + d_2 \cdot e'' + d_2 \cdot s^2 + s \cdot d_1 - s \cdot d_2 \cdot a'' \\ &= d_0 + s \cdot d_1 + s^2 \cdot d_2 + d_2 \cdot e'' = \Delta m + E + d_2 \cdot e'' \end{aligned}$$

Remember  
decryption  
is  $c_0 + s \cdot c_1$ !

Remember the  
server  
computes this  
and  $s^2$  is secret!

# Multiplication III

- Introduce the two following functions, which will help **manage the noise growth**

$$\forall a \in \mathcal{R} = \begin{cases} \mathcal{D}_{\omega,q}(a) &= ([a]_{\omega}, [\lfloor a\omega^{-1} \rfloor]_{\omega}, \dots, [\lfloor a\omega^{-\kappa} \rfloor]_{\omega}) \in \mathcal{R}_{\omega}^{\kappa} \\ \mathcal{P}_{\omega,q}(a) &= ([a]_q, [a\omega]_q, \dots, [a\omega^{\kappa}]_q) \in \mathcal{R}_q^{\kappa} \end{cases}$$

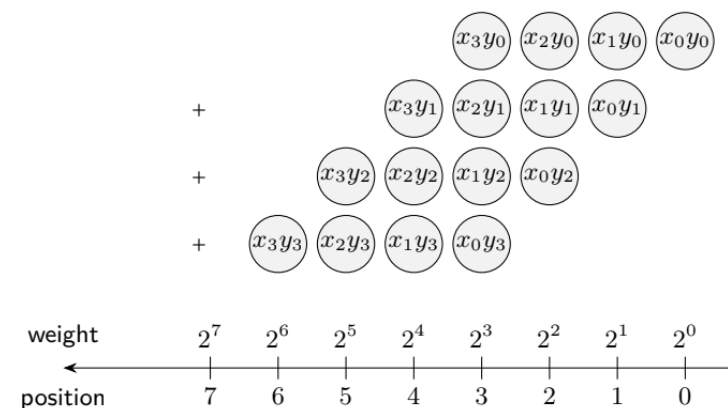
$$\forall a, b \in \mathcal{R}, \langle \mathcal{D}_{\omega,q}(a), \mathcal{P}_{\omega,q}(b) \rangle \equiv ab \pmod{q}$$

- Also need a **relinearization key** as follows (i.e. an encryption of  $s^2$ )

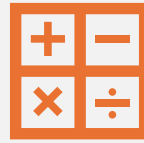
$$rlk = ([\mathcal{P}_{\omega,q}(s^2) - (e + sa)]_q, a)$$

- Finally, we compute the output ciphertext as

$$ct_{\text{out}} = ([d_0 + \langle \mathcal{D}_{\omega,q}(d_2), rlk_0 \rangle]_q, [d_1 + \langle \mathcal{D}_{\omega,q}(d_2), rlk_1 \rangle]_q)$$



# Challenges for RNS



The product in the first step is very **large** – would normally require a lift to  $\mathbb{Z}$



Dividing and rounding is incompatible with RNS (only exact division)



Decomposition in base  $\omega$  requires a positional system

# (A) Solution for Decryption

- Recall that division is performed as

$$\lfloor \frac{t}{q} [c_0 + s \cdot c_1]_q \rfloor$$

- Exact divisions are possible in RNS!

$$\lfloor \frac{t}{q} [c_0 + s \cdot c_1]_q \rfloor = \frac{t [c_0 + s \cdot c_1]_q - [t \cdot c_0 + s \cdot c_1]_q}{q}$$

- This results in an **approximate** decryption
- Then, correct the approximation

# (A) Solution for Multiplication

- Introduce an auxiliary base to contain the product in step 1 (which is large!) Replaces the lift to  $\mathbb{Z}$

$$FastBaseConv = \left( \sum_{i=1}^k \left| x_i \frac{q_i}{q} \right|_{q_i} \times \frac{q}{q_i} \pmod{m} \right)_{m \in \mathcal{B}}$$

- Use RNS instead – if  $\omega$  has the same size as the moduli in  $q$ , this is a good approximation

$$\xi_q(\bar{c}_2) = \left( \left| \bar{c}_2 \frac{q_1}{q} \right|_{q_1}, \dots, \left| \bar{c}_2 \frac{q_k}{q} \right|_{q_k} \right)$$

$$\mathcal{P}_{RNS,q} = \left( \left| s^2 \frac{q}{q_1} \right|_{q_1}, \dots, \left| s^2 \frac{q}{q_k} \right|_{q_k} \right)$$

# Decomposition base $2^k$

- Elements in the ciphertext space are polynomials  $a \in \mathcal{R}_q$
- Write  $a = \sum_{i=0}^{N-1} a_i X^i$
- Introduce a second variable  $Y$
- Write further  $a_i = \sum_{j=0}^{L-1} a_{i,j} 2^{jk} = \sum_{j=0}^{L-1} a_{i,j} Y^j$ , where  $L = D/k$ , for some chosen precision  $D$
- This can replace the RNS representation

**Revisiting Key Decomposition Techniques in FHE: Simpler, Fast and More Generic.** Mariya Georgieva Belorgey, Sergiu Carpov, Nicolas Gama, Sandra Guasch, Dimitar Jetchev. AsiaCrypt 2024

**Accelerating HE Operations from Key Decomposition Technique.** Miran Kim, Dongwon Lee, Jinyeong Seo, Yongsoo Song. Crypto 2023

**An HPR variant of the FV scheme: Computationally Cheaper, Asymptotically Faster.** Jean-Claude Bajard, Julien Eynard, Paulo Martins, Leonel Sousa, Vincent Zucca. Transactions on Computers 2019.



# Plaintext types

# Plaintext types

## Plaintext types for the different schemes

- (small) Integers for TFHE
- Polynomials in  $\mathcal{R}_t = \frac{\mathbb{Z}[x]}{x^{N+1,t}}$  for BGV, BFV
- (fixed precision) vectors in  $\mathbb{C}^{N/2}$  for CKKS ( $N$  is the degree of the ring)

## One interesting optimisation question

- BGV, BFV, CKKS can only evaluate **additions, multiplications, rotations**
- This means that non-linear functions cannot be computed **exactly** (ReLU, SoftMax etc)
- How to best approximate these?
  - Remember, multiplications are expensive!

## This is different for TFHE!

- Can evaluate lookup tables efficiently! So can evaluate most nonlinear functions **exactly** and **efficiently**
- However, it performs best on messages of 5 bits of precision
- Cannot pack as many values as the other schemes

# Data types



Integers (polynomials) are the most natural to deal with



Fixed point requires some care



Floating point is more delicate

**An Efficient Encrypted Floating-Point Representation Using HEAAN and TFHE**, Subin Moon, Younho Lee. SCN 2020.

**Simple Encrypted Arithmetic Library**, Kim Laine. 2017

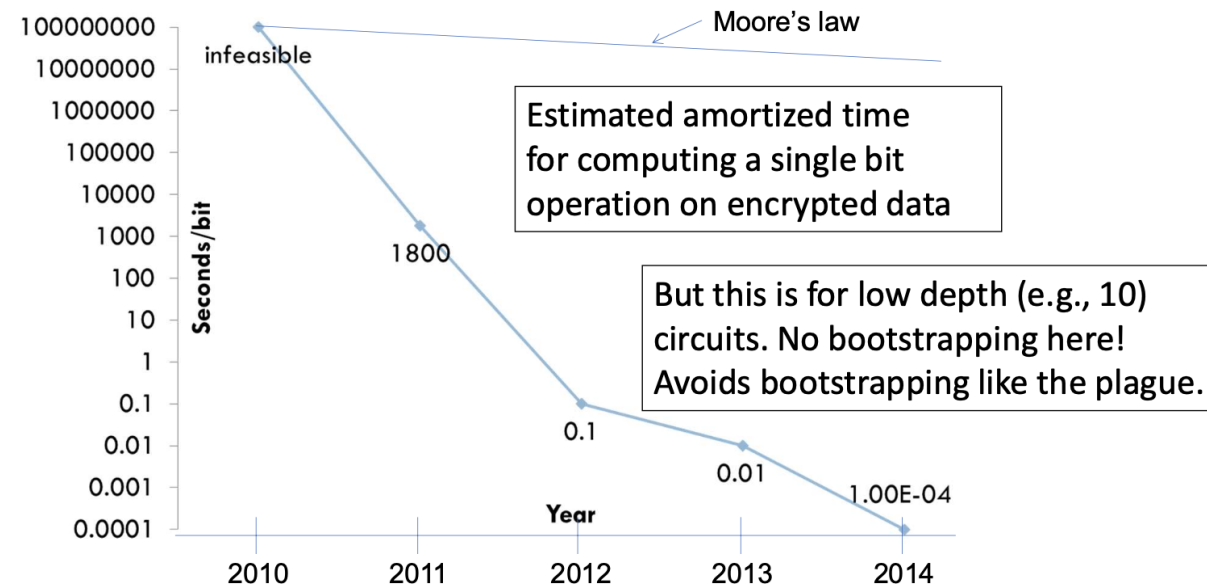
**Fixed-point Arithmetic in SHE schemes**. *Anamaria Costache*, Nigel P. Smart, Srinivas Vivek and Alexander Waller. SAC 2016.

**TFHE gets real: an Efficient and Flexible Homomorphic Floating-Point Arithmetic**. Loris Bergerat, Ilaria Chillotti, Damien Ligier, Jean-Baptiste Orfila and Samuel Tap. CHES 2025.

# Conclusion and open questions

# Evolution of speed of computing

## Speed of Computing on Encrypted Data on IBM's HElib Platform (1<sup>st</sup> to 2<sup>nd</sup> Gens)



# Cryptographic-specific challenges



## Verifiable computation:

FHE does not, by itself, offer any guarantees about the *correctness* of the function evaluated

Must operate in an *honest-but-curious* setting, where we trust that the server runs the operation intended, correctly

Verifiable Computation (VC) addresses this question exactly – but many efficiency issues

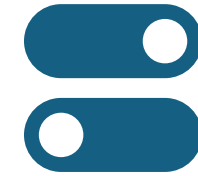


## Circuit privacy:

FHE does not, by itself, hide the circuit to be evaluated!

Particularly relevant in Privacy-Preserving Machine Learning (PPML) settings, where server will not want to release the model

Some recent work, but much remains to be done



## Scheme switching:

Some schemes have good packing capabilities, some have very fast operations and can handle nonlinear operations

Would be good to have an efficient way of efficiently switching between different types of schemes



# Thank you for your attention!

Special thanks to Laurent Imbert for helpful advice in preparing this talk!

- Gentry 2009: **Fully Homomorphic Encryption from Ideal Lattices**. Craig Gentry. ACM Symposium on Theory of Computing 2009.
- BGV: **(Levelled) Fully Homomorphic Encryption without Bootstrapping**. Zvika Brakerski, Craig Gentry, Vinod Vaikuntanathan. ACM Transactions on Computation Theory 2012.
- BFV:
  - **Fully Homomorphic Encryption without Modulus Switching from GapSVP**. Zvika Brakerski. Crypto 2012.
  - **Somewhat Practical Fully Homomorphic Encryption**. Junfeng Fan, Frederik Vercauteren. ePrint archive 2012.
- GSW: **Homomorphic Encryption from Learning with Errors: Conceptually Simpler, Asymptotically-Faster, Attribute-Based**. Craig Gentry, Amit Sahai, Brent Waters. Crypto 2013.
- FHEW: **FHEW: Bootstrapping Homomorphic Encryption in Less than a Second**. Léo Ducas, Daniele Micciancio. Eurocrypt 2014.
- TFHE: **Fast Fully Homomorphic Encryption over the Torus**. Ilaria Chillotti, Nicolas Gama, Mariya Georgieva, Malika Izabachène. Journal of Cryptography 2020.
- CKKS: **Homomorphic Encryption for Arithmetic of Approximate Numbers**. Jung Hee Cheon, Andrey Kim, Miran Kim, Yonsoo Song. Asiacrypt 2017.