Fully Homomorphic Encryption Lab

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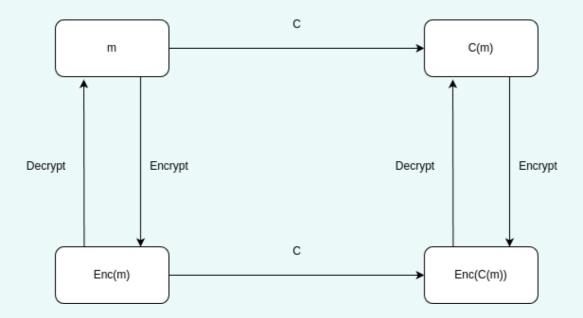
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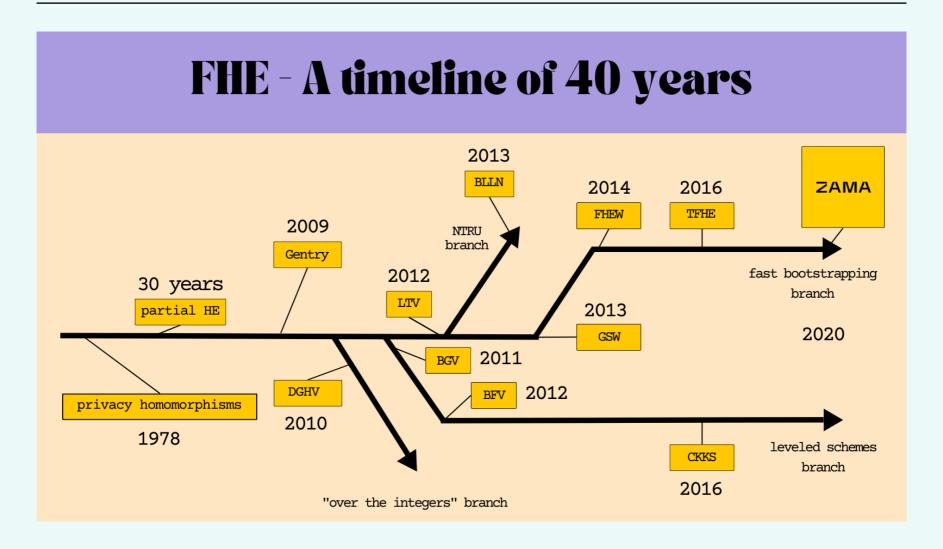
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These include:

- Brakerski-Fan/Vercauteren (BFV) [2]
- · Brakerski-Gentry-Vaikuntanathan (BGV) [3]
- · Cheon-Kim-Kim-Song (CKKS), which supports approximate arithmetic [4]
- Gentry-Sahai-Waters (GSW) [5]
- Homomorphic Encryption over the Torus (TFHE) [6]
- · FHEW due to Ducas and Micciancio [7]



Introduction to OpenFHE

OpenFHE is a comprehensive library for employing FHE in code that provides APIs in C/C++, Python and Rust, supporting schemes like BGV, CKKS, and BFV. Here, we demonstrate usage using the Python bindings for OpenFHE's BFV scheme implementation.

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Documentation:

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Some common methods:

- MakePackedPlaintext(pt) -> pt1 Encode the plaintext vector into packed form
- EvalAdd(ct1, ct2) -> ct3 Perform homomorphic addition
- EvalSub(ct1, ct2) -> ct3 Perform homomorphic subtraction
- EvalMult(ct1, ct2) -> ct3 Perform homomorphic multiplication
- EvalMultKeyGen(sk) Generate the relinearization key used for EvalMult
- EvalSum(ct, batchSize) -> ct1 Evaluate the sum of all components in a vector
- EvalSquare(ct) -> ct1 Compute the square of a ciphertext

OpenFHE: Basic usage

This is a sample code template that demonstrates usage of the BFV-RNS (residue number system) scheme for performing FHE (add and mult) on two plaintext vectors:

```
from openfhe import *
# Set CryptoContext
parameters = CCParamsBFVRNS() # Create instance of the BFV-RNS scheme
parameters.SetPlaintextModulus(65537) # Define plaintext space
parameters. SetMultiplicativeDepth(4) # Max no. of mults w/o bootstrapping
crypto context = GenCryptoContext(parameters)
crypto context.Enable(PKESchemeFeature.PKE) # Allow public-key encryption
crypto context.Enable(PKESchemeFeature.LEVELEDSHE) # Enable leveled FHE w/o
bootstrapping
crypto context.Enable(PKESchemeFeature.KEYSWITCH) # Enable key switching /
relinearization
# Generate (pk, sk)
key pair = crypto context.KeyGen()
```

```
# Generate the relinearization key
crypto context.EvalMultKeyGen(key pair.secretKey)
# Encode first plaintext vector
vec1 = [1, 2, 3, 4, 5, 6, 7, 8, 9, 10]
pt1 = crypto context.MakePackedPlaintext(vec1)
# Encode second plaintext vector
vec2 = [11, 12, 13, 14, 15, 16, 17, 18, 19, 20]
pt2 = crypto context.MakePackedPlaintext(vec2)
# Encrypt the two vectors using the same public key
ct1 = crypto context.Encrypt(key pair.publicKey, pt1)
ct2 = crypto context.Encrypt(key pair.publicKey, pt2)
# Homomorphic addition
ct add = crypto context.EvalAdd(ct1, ct2)
# Homomorphic multiplication
ct mult = crypto context.EvalMult(ct1, ct2)
# Decrypt the result of the addition
```

```
pt add = crypto context.Decrypt(ct add, key pair.secretKey)
# Decrypt the result of the multiplication
pt mult = crypto context.Decrypt(ct mult ,key pair.secretKey)
print("Plaintext #1: " + str(pt1))
print("Plaintext #2: " + str(pt2))
# Output results
print("#1 + #2 = " + str(pt add))
print("#1 * #2 = " + str(pt mult))
Output:
Plaintext #1: ( 1 2 3 4 5 6 7 8 9 10 ...)
Plaintext #2: ( 11 12 13 14 15 16 17 18 19 20 ... )
#1 + #2 = (12 14 16 18 20 22 24 26 28 30 ...)
#1 * #2 = ( 11 24 39 56 75 96 119 144 171 200 ... )
```

Refer to the OpenFHE GitHub repository for more detailed examples that also demonstrate bootstrapping and Threshold-FHE, both of which are beyond the scope of this lab.

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 - Encrypt two arbitrary integers a=6,b=8
 - Compare them homomorphically
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$$(a^2 - b^2) = (a+b)(a-b)$$

$$(a+b)^3 = a^3 + 3a^2b + 3ab^3 + b^3$$

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$$(a+b+c)^2 = a^2 + b^2 + c^2 + 2(ab+bc+ca)$$

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 - Compare them homomorphically
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- Polynomial evaluation:
 - Take a multivariate polynomial, say, $P(x,y) = 2x^2 + 3xy + 4y^2 + 5x + 6y + 7$
 - Evaluate the polynomial homomorphically on, say, x=3,y=4
 - Decrypt and verify the result P(3,4) = 164

- Matrix multiplication:
 - Encrypt two matrices $A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}, B = \begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix}$
 - Perform homomorphic matrix multiplication
 - Decrypt and verify the result matches the product $C = A \cdot B = \begin{pmatrix} 19 & 22 \\ 43 & 50 \end{pmatrix}$

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- · Arithmetic mean:
 - Encrypt a dataset A = [412, 8423, 66, 891, 277, 84, 5, 9]
 - ullet Homomorphically compute the arithmetic mean of A
 - Decrypt and verify the result

Bibliography

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