

Lecture 19: Fully Homomorphic Encryption (FHE)

Pt. 1

Plan

Recap: LWE

Fully Homomorphic Encryption

Introduction & history

Syntax & security

Building leveled FHE

Recap: Learning with Errors

$LWE(n, m, q, \chi_B)$

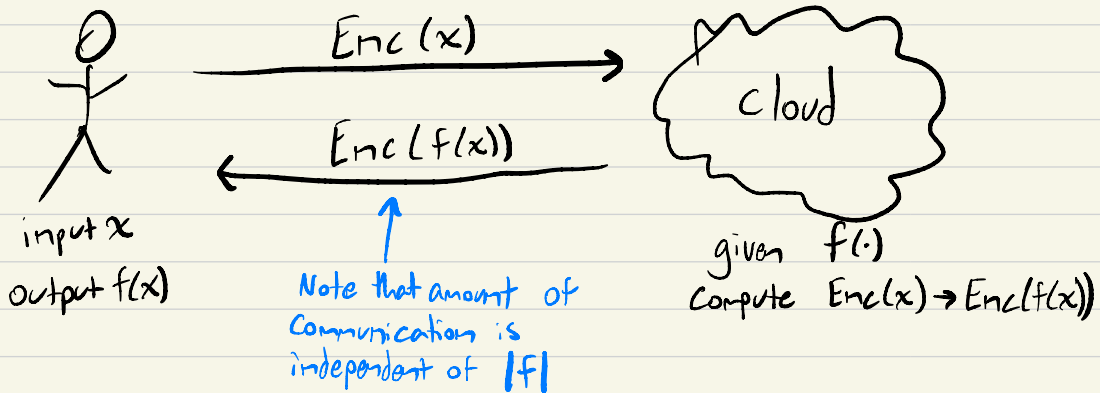
$$\left\{ (A, A\vec{s} + \vec{e}) : \begin{array}{l} A \in \mathbb{Z}_q^{m \times n} \\ \vec{s} \in \mathbb{Z}_q^n \\ \vec{e} \in \chi_B^n \end{array} \right\} \approx_c \left\{ (A, u) : \begin{array}{l} A \in \mathbb{Z}_q^{m \times n} \\ u \in \mathbb{Z}_q^n \end{array} \right\}$$

$$\begin{matrix} n \\ \boxed{A} \\ m \end{matrix} \cdot \begin{matrix} 1 \\ \boxed{s} \\ n \end{matrix} + \begin{matrix} 1 \\ \boxed{e} \\ m \end{matrix} = \begin{matrix} 1 \\ \boxed{As+e} \\ m \end{matrix}$$

$$\begin{matrix} n+1 \\ \boxed{A \parallel As+e} \\ m \end{matrix} \approx_c \begin{matrix} \text{Uniform} \\ \text{in} \\ \mathbb{Z}_q^{m \times (n+1)} \end{matrix}$$

Fully Homomorphic Encryption

Idea: outsource computation without revealing inputs!



Examples:

- PIR: input i , output $f(i) = DB[i]$
- private ML: training, inference
- whatever you want to outsource!

Brief history

- 1978 - Rivest, Adleman, Dertouzos introduced a notion of FHE

Context: Diffie-Hellman introduced PK crypto in 1976

↓ no unbroken candidates for many years

- 2009 - Craig Gentry (Stanford PhD student)
↳ CS355 TA fall '07
gives first construction!
 - from new (non-standard) assumption
 - introduces "bootstrapping" idea (see next lecture)


- 2011 - Brakerski, Vaikuntanathan

FHE based on LWE

- 2013 - Gentry, Sahai, Waters


"3rd gen" FHE ← today's topic


FHE Syntax

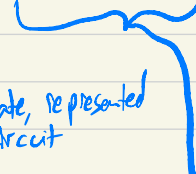
$\text{KeyGen}(1^n) \rightarrow \text{sk}$  We will look at Symmetric-Key FHE. There is actually a generic transformation to get PK FHE to

$\text{Enc}(\text{sk}, M) \rightarrow \text{ct}$

$\text{Dec}(\text{sk}, \text{ct}) \rightarrow M$

$\text{Eval}(F, \text{ct}_1, \dots, \text{ct}_\ell) \rightarrow \tilde{\text{ct}}$  encryption of circuit output

 function to evaluate, represented as boolean circuit

 encryptions of inputs

FHE Properties

1) Correctness: $\forall F: \{0,1\}^l \rightarrow \{0,1\}, \mu_1, \dots, \mu_g \in \{0,1\}$

$sk \leftarrow \text{KeyGen}(1^\lambda)$, then with probability 1:

$$\text{Dec}(sk, \text{Eval}(F, \text{Enc}(sk, \mu_1), \dots, \text{Enc}(sk, \mu_g))) = F(\mu_1, \dots, \mu_g)$$

+ usual encryption correctness

2) Semantic security

$$\forall \mu_0, \mu_1 \in \{0,1\} \quad \{\text{Enc}(sk, \mu_0)\} \approx_c \{\text{Enc}(sk, \mu_1)\}$$

3) Compactness

$$\forall F, sk \quad ct_i \leftarrow \text{Enc}(sk, \mu_i)$$

$$\text{if } \tilde{ct} \leftarrow \text{Eval}(F, ct_1, \dots, ct_g)$$

then $|\tilde{ct}| = \text{poly}(\lambda) \leftarrow \tilde{ct}$ size is independent of $|F|, \lambda$

Note: without compactness, any encryption scheme is also fully homomorphic!

$$\text{Eval}(F, \{ct_i\}) \rightarrow (F, \{ct_i\})$$

$$\text{Dec}(sk, (F, \{ct_i\})) \rightarrow F(\text{Dec}(sk, ct_1), \dots, \text{Dec}(sk, ct_g))$$

Eval just writes down F and all the inputs,
Dec evaluates F after decrypting!

Constructing FHE

Today: construct leveled FHE,

↳ can only evaluate low-depth circuits

Reason: ciphertexts have noise that grows with each gate in circuit. Eventually, the noise overwhelms the msg.

Next time: use bootstrapping to remove restriction on circuit depth

↳ Idea: refresh ciphertexts to clear accumulated noise.

Attempt 1 (insecure)

Secret Key is a vector \vec{s}

Enc(\vec{s}, M) \rightarrow matrix C s.t. $C \cdot \vec{s} = M \cdot \vec{s}$

\vec{s} is eigenvector of C w/ eigenvalue M

Dec(\vec{s}, C) \rightarrow Compute $C \cdot \vec{s} = M \cdot \vec{s}$ and find M

Homomorphism: if C_1, C_2 are encryptions of M_1, M_2

Addition: $\tilde{C} = C_1 + C_2$ $(C_1 + C_2) \vec{s} = C_1 \vec{s} + C_2 \vec{s} = M_1 \vec{s} + M_2 \vec{s} = (M_1 + M_2) \vec{s}$
Eval("+", C_1, C_2) \uparrow
by def of Enc

Multiplication: $\tilde{C} = C_1 \cdot C_2$
Eval(".", C_1, C_2)

$$(C_1 \cdot C_2) \vec{s} = C_1 \cdot (C_2 \vec{s}) = C_1 M_2 \vec{s} = M_2 \cdot (C_1 \vec{s}) = M_2 M_1 \vec{s} = M_1 M_2 \vec{s}$$

\uparrow by def of Enc \uparrow by def of Enc

Can eval $+, \cdot \rightarrow$ fully homomorphic!

Problem: Given C , it's easy to find \vec{s} using Gaussian elimination $\ddot{=}$

idea: We can make Gaussian elimination hard by adding noise!

$$C \cdot \vec{s} = M \cdot \vec{s} + \vec{e}$$

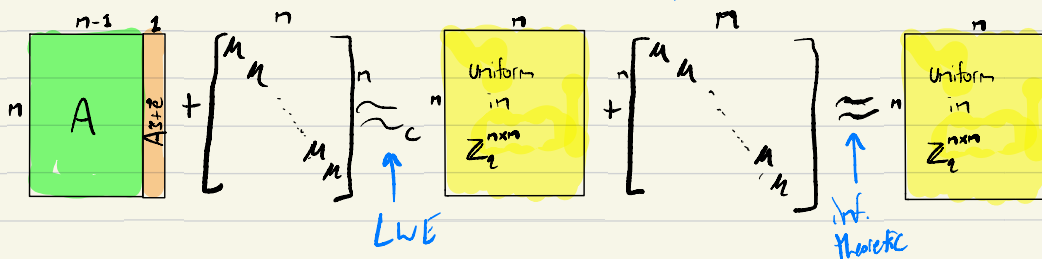
\uparrow
Small noise

Attempt 2: Secret-Key variant of Regev encryption

$$\text{KeyGen}(1^n): \tilde{s} \xleftarrow{R} \mathbb{Z}_q^{n-1}, \quad \vec{s} \leftarrow \begin{pmatrix} \tilde{s} \\ -1 \end{pmatrix} \in \mathbb{Z}_q^n$$

$$\text{Enc}(\vec{s}, \mu): A \xleftarrow{R} \mathbb{Z}_q^{n \times (n-1)} \quad \vec{e} \xleftarrow{R} \chi_B^n$$

$$\text{output } C = \underbrace{(A, A\vec{s} + \vec{e})}_{\text{Pseudorandom by LWE}} + \mu \cdot I_n \in \mathbb{Z}_q^{n \times n}$$



$$\text{Dec}(\vec{s}, C): \text{compute } C \cdot \vec{s}, \text{ output } \begin{cases} 0 & \text{if } \|C \cdot \vec{s}\|_{\infty} \text{ small} \\ 1 & \text{otherwise} \end{cases}$$

$$C \cdot \vec{s} = (A, A\vec{s} + \vec{e}) \begin{pmatrix} \tilde{s} \\ -1 \end{pmatrix} + \mu I_n \vec{s}$$

$$= \cancel{A\tilde{s}} - \cancel{A\tilde{s}} - \vec{e} + \mu \vec{s} = \mu \vec{s} + \text{noise} \rightarrow \begin{cases} \text{small: } \mu = 0 \\ \text{large: } \mu = 1 \end{cases}$$

\vec{s} is "approximate eigenvector" of C
with approx. eigenvalue μ .

Homomorphism:

Addition: $\tilde{C} \leftarrow C_1 + C_2$

$$(C_1 + C_2)\vec{s} = C_1\vec{s} + C_2\vec{s} = \mu_1\vec{s} + \vec{e}_1 + \mu_2\vec{s} + \vec{e}_2 = (\mu_1 + \mu_2)\vec{s} + (\vec{e}_1 + \vec{e}_2)$$

Noise doesn't grow too much, so we have additive homomorphism

noise still reasonably small

(Would need to adjust Dec for $\mu \notin \{0, 1\}$)

Multiplication: Can we do $\tilde{C} \leftarrow C_1 \cdot C_2$?

$$(C_1 \cdot C_2)\vec{s} = C_1(\mu_2\vec{s} + \vec{e}_2) = \mu_2 C_1\vec{s} + C_1\vec{e}_2$$

$$\begin{aligned} &= \mu_2(\mu_1\vec{s} + \vec{e}_1) + C_1\vec{e}_2 \\ &= \mu_1 \cdot \mu_2 \cdot \vec{s} + \underbrace{\mu_2 \cdot \vec{e}_1}_{\text{still reasonably small for } \mu_2 \in \{0, 1\}} + \underbrace{C_1 \cdot \vec{e}_2}_{\text{can be large } \ddot{}} \end{aligned}$$

still reasonably small
for $\mu_2 \in \{0, 1\}$

can be large $\ddot{}$

So we're still unable to multiply b/c noise grows with $\|C_1\|_\infty$, which can be large.

Need a way to make ciphertext matrices have small norm.

Idea: represent number $x \in \mathbb{Z}_q$ as a small-norm vector via

binary decomposition!

Binary Decomposition

For $x \in \mathbb{Z}_q$,

Define $\hat{x} = (x_0, x_1, \dots, x_{\log_2 q - 1})$ s.t. $x = \sum_{i=0}^{\log_2 q - 1} x_i \cdot 2^i$

inverse operation $\hat{x} \rightarrow x$
is linear!

let \vec{G} be the vector that recovers x from \hat{x} : $\hat{x} \cdot \vec{G} = x$

$$\hat{x} \cdot \vec{G} = x$$

$$\underbrace{(x_0, \dots, x_{\log_2 q - 1})}_{\log_2 q} \cdot \underbrace{\begin{pmatrix} 2^0 \\ 2^1 \\ \vdots \\ 2^{\log_2 q - 1} \end{pmatrix}}_1 \Bigg\}^{\log_2 q} = x$$

Note: we call it \vec{G} for "gadget"

can extend ($\hat{\cdot}$) operation to vectors

$$\vec{x} \in \mathbb{Z}_q \rightarrow \hat{x} = (x_{0,0}, x_{0,1}, \dots, x_{0,\log q-1}, \dots, x_{n,0}, x_{n,1}, \dots, x_{n,\log q-1})$$

$\in \{0,1\}^{n \log q}$

$$\vec{x} = \hat{x} \cdot G \text{ is a linear transformation}$$

$\underbrace{\quad}_{1 \times n \log q} \quad \underbrace{\quad}_{n \log q \times n}$

where G is the matrix that recovers \vec{x} from \hat{x} : $\hat{x} \cdot G = \vec{x}$

$$G = \begin{bmatrix} 1 & & & & \\ 2 & & & & \\ \vdots & & & & \\ \log q - 1 & & & & \\ & 1 & & & \\ & 2 & & & \\ & \vdots & & & \\ & \log q - 1 & & & \\ & & \ddots & & \\ & & & 1 & \\ & & & 2 & \\ & & & \vdots & \\ & & & \log q - 1 & \end{bmatrix}$$

n

$\in \mathbb{Z}_q^{n \log q \times n}$

finally, can also extend $(\hat{\cdot})$ to matrices

$$C = \begin{pmatrix} \vec{c}_1 \\ \vdots \\ \vec{c}_m \end{pmatrix} \rightarrow \hat{C} = \begin{pmatrix} \hat{c}_1 \\ \vdots \\ \hat{c}_m \end{pmatrix}$$

$m \times n$ $m \times n \log q$

And $C = \hat{C} \cdot G$ is still a linear transformation

(With gadget matrix G same as above)

Note: Some sources refer to G as G^{-1} because it inverts bit decomposition

Now, let's get back to FHE!

3rd (and final) attempt: the GSW scheme

$$\text{KeyGen}(1^n): \tilde{s} \leftarrow \mathbb{Z}_q^{n-1} \quad \tilde{S} \leftarrow \begin{pmatrix} \tilde{s} \\ -1 \end{pmatrix} \in \mathbb{Z}_q^n$$

$$\text{Enc}(\tilde{s}, \mu): A \leftarrow \mathbb{Z}_q^{m \times (n-1)} \quad \text{for } m = n \log q$$

$$\vec{e} \leftarrow \mathcal{X}_B^m$$

$$C = \underbrace{(A, A\tilde{s} + \vec{e})}_{m \times n} + \mu G$$

$$\text{output } \hat{C} \\ \underbrace{m \times m}_{m \times m = m \times n \log q}$$

Observe that $ct = \hat{C}$ has low norm since it is a $\{0,1\}$ -matrix!

$$\begin{aligned} \text{Dec}(\tilde{S}, \hat{C}): \text{ compute } \underbrace{\hat{C} \cdot G}_{C} \cdot \tilde{S} &= C \cdot \tilde{S} \\ &= (A, A\tilde{s} + \vec{e}) \begin{pmatrix} \tilde{s} \\ -1 \end{pmatrix} + \mu \cdot G \cdot \tilde{S} \\ &= \mu \cdot G \cdot \tilde{S} - \vec{e} \end{aligned}$$

if first element is small, output $\mu=0$. Else, output $\mu=1$.

Why first element?

$$(\mu \cdot G \cdot \tilde{S})_1 = \mu (1 \ 0 \ \dots \ 0) \begin{pmatrix} s_1 \\ s_2 \\ \vdots \\ s_{n-1} \\ -1 \end{pmatrix}$$

$$= \mu \cdot |s_1| \text{ which is } \Theta(q) \text{ whp. iff } \mu=1$$

Let's see how this solves our multiplication problem:

Eval(\cdot , \hat{C}_1 , \hat{C}_2): output $\hat{C} = \hat{C}_1 \cdot \hat{C}_2$ \leftarrow all $m \times m$ matrices

Need to check $\text{Dec}(\vec{s}, \hat{C}) \stackrel{?}{=} \mu_1 \cdot \mu_2$

Proof

$$\begin{aligned}\hat{C}_1 \cdot \underbrace{\hat{C}_2 \cdot G \cdot \vec{s}}_{C_2} &= \hat{C}_1 \cdot (C_2 \cdot \vec{s}) \\ &= \hat{C}_1 \cdot (\mu_2 \cdot G \cdot \vec{s} + \vec{e}_2) \\ &= \mu_2 \cdot \underbrace{\hat{C}_1 \cdot G \cdot \vec{s}}_{C_1} + \hat{C}_1 \cdot \vec{e}_2 \\ &= \mu_2 (\mu_1 \cdot G \cdot \vec{s} + \vec{e}_1) + \hat{C}_1 \cdot \vec{e}_2 \\ &= \mu_1 \mu_2 G \vec{s} + \mu_2 \cdot \vec{e}_1 + \hat{C}_1 \cdot \vec{e}_2\end{aligned}$$

Small if
 $\mu_2 \in \{0,1\}$

Small since
 $\|\hat{C}_1\|_\infty = \text{Small}$

What about addition? Could that make noise bigger?

Turns out it's sufficient to support the universal NAND gate:

$$\text{NAND}(a,b) = \text{NOT}(\text{AND}(a,b))$$

Using NAND for other gates:

$$\text{NOT}(a) = \text{NAND}(a,a)$$

$$\text{AND}(a,b) = \text{NOT}(\text{NAND}(a,b))$$

$$\text{OR}(a,b) = \text{NAND}(\text{NOT}(a), \text{NOT}(b))$$

So how to build NAND?

$$\text{Eval}(\text{NAND}, \hat{C}_1, \hat{C}_2): \underbrace{I_{n \times n}}_{\substack{\text{Ct has } \mu \text{ added} \\ \text{along its diagonal,} \\ \text{so } 1-\mu \text{ is NOT}}} - \underbrace{\hat{C}_1 \cdot \hat{C}_2}_{\substack{\text{mult over } \{0,1\} \\ \text{is AND}}}$$

Yay! We have constructed an encryption scheme that can compute a universal gate over ciphertexts!

But this is a leveled FHE, so we're not done yet.

Next time we'll see why this is not quite an FHE yet (noise growth), as well as a technique called **Bootstrapping** that will allow us to get a full FHE scheme.