

Somewhat Homomorphic Encryption based on Random Ideal Codes

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Outline

- 1 What is homomorphic encryption?
- 2 The difficulty with LPN-based FHE
- 3 New idea: Sample errors in the same support
- 4 Our construction

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What is Homomorphic Encryption?

Public-key version

- $\text{KeyGen}(1^\lambda) \rightarrow (\text{pk}, \text{sk})$
- $\text{Enc}(m, \text{pk}) \rightarrow \text{ct}$
- $\text{Dec}(\text{ct}, \text{sk}) \rightarrow m$
- $\text{Eval}(f, \text{ct}_1, \text{ct}_2) \rightarrow \text{ct}$

Proposition (Correctness)

$$\text{Dec}(\text{Eval}(f, \text{Enc}(m_1, \text{pk}), \text{Enc}(m_2, \text{pk})), \text{sk}) = f(m_1, m_2)$$

What is Homomorphic Encryption?

- $f \in \{+, \times\} \rightarrow$ **partial** homomorphic encryption (RSA)
- $f \in \mathbb{F}_d[X] \rightarrow$ **somewhat** homomorphic encryption [BGN05]
- $f \in \mathbb{F}[X] \rightarrow$ **fully** homomorphic encryption [Gen09]

Noisy ciphertexts

$$\text{ct}_1 = \mathbf{A}\mathbf{m}_1 + \mathbf{e}_1$$

$$\text{ct}_2 = \mathbf{A}\mathbf{m}_2 + \mathbf{e}_2$$

$$\text{Eval}(+, \text{ct}_1, \text{ct}_2) = \mathbf{A}(\mathbf{m}_1 + \mathbf{m}_2) + \underbrace{\mathbf{e}_1 + \mathbf{e}_2}_{\text{double noise}}$$

In general:

$$\text{Eval}(f, \text{Enc}(m_1, \text{pk}), \text{Enc}(m_2, \text{pk})) \neq \text{Enc}(f(m_1, m_2), \text{pk}).$$

Bootstrapping: how to reduce ciphertext noise

$$(\text{pk}_1, \text{sk}_1) = \text{KeyGen}(1^\lambda)$$

$$\text{ct}_0 \approx \text{Enc}(m, \text{pk}_1)$$

$$(\text{pk}_2, \text{sk}_2) = \text{KeyGen}(1^\lambda)$$

$$\text{ct}_1 = \text{Enc}(\text{ct}_0, \text{pk}_2) \approx \text{Enc}(\text{Enc}(m, \text{pk}_1), \text{pk}_2)$$

$$\text{ct}_2 = \text{Enc}(\text{sk}_1, \text{pk}_2)$$

$$\text{Eval}(\text{Dec}(\cdot, \cdot), \text{ct}_1, \text{ct}_2) = ?$$

Bootstrapping: how to reduce ciphertext noise

$$\text{ct}_0 \approx \text{Enc}(m, \text{pk}_1)$$

$$\text{ct}_1 = \text{Enc}(\text{ct}_0, \text{pk}_2)$$

$$\text{ct}_2 = \text{Enc}(\text{sk}_1, \text{pk}_2)$$

$$\text{Dec}(\text{Eval}(\text{Dec}(\cdot, \cdot), \text{ct}_1, \text{ct}_2), \text{sk}_2) = \text{Dec}(\text{ct}_0, \text{sk}_1) = m$$

$$(\text{Eval}(\text{Dec}(\cdot, \cdot), \text{ct}_1, \text{ct}_2) \approx \text{Enc}(\text{m}, \text{pk}_2))$$

History of fully homomorphic encryption

There has been a burst of activity in the last decade:

- 2009: Gentry's first FHE [Gen09]
- 2010-2015: Practical somewhat homomorphic encryption
- 2016: TFHE [CGGI16], bootstrapping below 100ms
- 2016-present: remarkable progress

... but most of existing constructions are based on **structured lattices**.

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What is Homomorphic Encryption?

Secret-key version

- $\text{KeyGen}(1^\lambda) \rightarrow \text{sk}$
- $\text{Enc}(m, \text{sk}) \rightarrow \text{ct}$
- $\text{Dec}(\text{ct}, \text{sk}) \rightarrow m$
- $\text{Eval}(f, \text{ct}_1, \text{ct}_2) \rightarrow \text{ct}$

Proposition (Correctness)

$$\text{Dec}(\text{Eval}(f, \text{Enc}(m_1, \text{sk}), \text{Enc}(m_2, \text{sk})), \text{sk}) = f(m_1, m_2)$$

Lattice-based HE (secret-key version)

$$\text{sk} = \textcolor{brown}{s} \in \mathbb{F}_q^k$$

$$\text{Enc}(m, \text{sk}) = (\textcolor{teal}{A} \in \mathbb{F}_q^{n \times k}, \textcolor{teal}{v} = \textcolor{teal}{A}\textcolor{brown}{s} + \textcolor{brown}{e} + \text{Encode}(\textcolor{brown}{m}))$$

$$\text{Dec}(\text{ct}, \text{sk}) = \text{Decode}(\textcolor{teal}{v} - \textcolor{teal}{A}\textcolor{brown}{s})$$

Usually: $\text{Encode}(m) = m \pmod p$ (with $p < q$)

The LWE assumption

$$\mathbf{A}, \begin{pmatrix} & \\ & \mathbf{A} \\ & \end{pmatrix} + \begin{pmatrix} \mathbf{s} \\ \mathbf{e} \end{pmatrix} \approx \mathbf{A}, \begin{pmatrix} & \\ & \mathbf{y} \\ & \end{pmatrix}$$

$$\mathbf{A} \leftarrow \mathbb{F}_q^{k \times n}$$

$$\mathbf{s} \leftarrow \mathbb{F}_q^k$$

$$\mathbf{e} \leftarrow \chi_{q,n,\sigma}$$

$$\mathbf{A} \leftarrow \mathbb{F}_q^{k \times n}$$

$$\mathbf{y} \leftarrow \mathbb{F}_q^n$$

Linear homomorphism

$$\begin{aligned} \text{ct}_1 &= (\mathbf{A}_1, \mathbf{A}_1 \mathbf{s} + \mathbf{e}_1 + \text{Encode}(\mathbf{m}_1)) \\ \text{ct}_2 &= (\mathbf{A}_2, \mathbf{A}_2 \mathbf{s} + \mathbf{e}_2 + \text{Encode}(\mathbf{m}_2)) \end{aligned}$$

$$\text{Eval}(+, \text{ct}_1, \text{ct}_2) = \text{ct}_1 + \text{ct}_2 = (\mathbf{A}_1 + \mathbf{A}_2, (\mathbf{A}_1 + \mathbf{A}_2)\mathbf{s} + \underbrace{\mathbf{e}_1 + \mathbf{e}_2}_{\sigma' \approx \sqrt{2}\sigma} + \text{Encode}(\mathbf{m}_1 + \mathbf{m}_2))$$

Ideal Lattice-based HE

$$\text{sk} = \textcolor{orange}{s} \in \mathcal{R} = \mathbb{F}_q[X]/\mathcal{I}$$

$$\text{Enc}(m, \text{sk}) = (\textcolor{teal}{a} \in \mathcal{R}, \textcolor{teal}{v} = \textcolor{teal}{a} \cdot \textcolor{orange}{s} + \textcolor{orange}{e} + \textit{Encode}(\textcolor{orange}{m}))$$

$$\text{Dec}(\text{ct}, \text{sk}) = \textit{Decode}(\textcolor{teal}{v} - \textcolor{teal}{a} \cdot \textcolor{orange}{s})$$

Multiplication

$$\begin{aligned} \text{ct}_1 &= (\textcolor{teal}{a}_1, \textcolor{teal}{v}_1) = (\textcolor{teal}{a}_1, \textcolor{teal}{v}_1 = \textcolor{teal}{a}_1 \cdot s + \textcolor{orange}{e}_1 + \text{Encode}(\textcolor{orange}{m}_1)) \\ \text{ct}_2 &= (\textcolor{blue}{a}_2, \textcolor{blue}{v}_2) = (\textcolor{blue}{a}_2, \textcolor{blue}{v}_2 = \textcolor{blue}{a}_2 \cdot s + \textcolor{orange}{e}_2 + \text{Encode}(\textcolor{orange}{m}_2)) \end{aligned}$$

$$\text{Eval}(\times, \text{ct}_1, \text{ct}_2) = \text{ct}_1 \cdot \text{ct}_2 = (\textcolor{teal}{a}_1 \cdot \textcolor{blue}{a}_2, \textcolor{teal}{a}_1 \cdot \textcolor{blue}{v}_2 + \textcolor{blue}{a}_2 \cdot \textcolor{teal}{v}_1, \textcolor{teal}{v}_1 \cdot \textcolor{blue}{v}_2) = (\textcolor{blue}{b}_0, \textcolor{blue}{b}_1, \textcolor{blue}{b}_2)$$

$$\textcolor{blue}{b}_0 \textcolor{orange}{s}^2 + \textcolor{blue}{b}_1 \textcolor{orange}{s} + \textcolor{blue}{b}_2 = \textcolor{orange}{e} + \text{Encode}(\textcolor{orange}{m}_1 \textcolor{orange}{m}_2)$$

Why homomorphic encryption with codes?

- An alternative to structured lattices
- Faster and simpler decryption circuit
- Support and multi-dimensional approach

The LPN assumption

$$\mathbf{A}, \begin{pmatrix} & \\ & \mathbf{A} \\ & \end{pmatrix} + \begin{pmatrix} \mathbf{s} \end{pmatrix} \approx \mathbf{A}, \begin{pmatrix} \mathbf{y} \end{pmatrix}$$

$$\mathbf{A} \leftarrow \mathbb{F}_q^{k \times n}$$

$$\mathbf{s} \leftarrow \mathbb{F}_q^k$$

$$\mathbf{e} \leftarrow \chi_{q,n,\sigma}$$

$$\mathbf{A} \leftarrow \mathbb{F}_q^{k \times n}$$

$$\mathbf{y} \leftarrow \mathbb{F}_q^n$$

The difference between LWE and LPN

LWE

$$\chi_{q,n,\sigma} = ([\mathcal{N}(0, \sigma)] \bmod q)^n$$

LPN

$$\chi_{q,n,\sigma} = (Ber_q(\sigma))^n$$

$$Ber_q(\sigma) = \begin{cases} 0 & \text{with probability } 1 - \sigma \\ i \in \mathbb{F}_q^* & \text{with probability } \sigma/(q-1) \end{cases}$$

Same linear homomorphism for LPN, but way more constrained

$$\begin{aligned} \text{ct}_1 &= (\mathbf{A}_1, \mathbf{A}_1 \mathbf{s} + \mathbf{e}_1 + \text{Encode}(\mathbf{m}_1)) \\ \text{ct}_2 &= (\mathbf{A}_2, \mathbf{A}_2 \mathbf{s} + \mathbf{e}_2 + \text{Encode}(\mathbf{m}_2)) \end{aligned}$$

$$\text{Eval}(+, \text{ct}_1, \text{ct}_2) = \text{ct}_1 + \text{ct}_2 = (\mathbf{A}_1 + \mathbf{A}_2, (\mathbf{A}_1 + \mathbf{A}_2)\mathbf{s} + \underbrace{\mathbf{e}_1 + \mathbf{e}_2}_{\sigma' \approx 2\sigma} + \text{Encode}(\mathbf{m}_1 + \mathbf{m}_2))$$

Maximum noise to decrypt:

- For LWE, $\sigma' \sqrt{n} = o(q)$
- For LPN, $\sigma' < 1/2$

Previous code-based approaches used structured codes

Examples are:

- Reed-Muller [AAPS11]
- Reed-Solomon [BL11] (broken by [GOT12])

⇒ highly **structured** codes

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Support

Gives additional information on the localization of the error.

Definition (Hamming support)

The **Hamming support** of a word $\mathbf{x} \in (\mathbb{F}_q)^n$ is the set of indexes of its non-zero coordinates:

$$Supp_h(\mathbf{x}) = \{i : x_i \neq 0\}$$

Errors with the same support form a linear subspace!

Reusing the same support

$$\begin{aligned} \text{ct}_1 &= (\mathbf{A}_1, \mathbf{A}_1 \mathbf{s} + \underbrace{\mathbf{e}_1}_{\text{same support}} + \text{Encode}(\mathbf{m}_1)) \\ \text{ct}_2 &= (\mathbf{A}_2, \mathbf{A}_2 \mathbf{s} + \underbrace{\mathbf{e}_2}_{\text{same support}} + \text{Encode}(\mathbf{m}_2)) \end{aligned}$$

$$\text{Eval}(+, \text{ct}_1, \text{ct}_2) = \text{ct}_1 + \text{ct}_2 = (\mathbf{A}_1 + \mathbf{A}_2, (\mathbf{A}_1 + \mathbf{A}_2) \mathbf{s} + \underbrace{\mathbf{e}_1 + \mathbf{e}_2}_{\sigma' \leq \sigma} + \text{Encode}(\mathbf{m}_1 + \mathbf{m}_2))$$

Security reduction (single ciphertext)

$$\mathbf{v} = \begin{pmatrix} \mathbf{A} \end{pmatrix} + \begin{pmatrix} \mathbf{s} \end{pmatrix} + \begin{pmatrix} \text{Encode}(\mathbf{m}) \end{pmatrix}$$

Security reduction (two ciphertexts)

$$\begin{pmatrix} s \\ \mathbf{v}_1 \\ \mathbf{v}_2 \end{pmatrix} = \begin{pmatrix} \mathbf{A}_1 \\ \mathbf{A}_2 \end{pmatrix} + \begin{pmatrix} \mathbf{e}_1 \\ \mathbf{e}_2 \end{pmatrix} + \begin{pmatrix} \text{Encode}(\mathbf{m}_1) \\ \text{Encode}(\mathbf{m}_2) \end{pmatrix}$$

Non standard error distribution!

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Changing to rank metric

\mathbb{F}_q is now an extension field : $\mathbb{F}_q = \mathbb{F}_{p^m}$

$$\text{sk} = \mathbf{s} \in \mathbb{F}_q^k$$

noise = $\mathbf{e} \in E^n$, with E a \mathbb{F}_p -subspace of \mathbb{F}_q of dimension σm

$$\text{Enc}(\mu, \text{sk}) = (\mathbf{A} \in \mathbb{F}_q^{n \times k}, \mathbf{v} = \mathbf{A}\mathbf{s} + \mathbf{e} + \text{Encode}(\mu))$$

$$\text{Dec}(\text{ct}, \text{sk}) = \text{Decode}(\mathbf{v} - \mathbf{A}\mathbf{s})$$

The Rank Decoding Problem (RDP) assumption

$$\mathbf{A}, \begin{pmatrix} & \\ & \mathbf{A} \\ & \end{pmatrix} + \begin{pmatrix} \mathbf{s} \\ \mathbf{e} \end{pmatrix} \approx \mathbf{A}, \begin{pmatrix} & \\ & \mathbf{y} \\ & \end{pmatrix}$$

$$\mathbf{A} \leftarrow \mathbb{F}_q^{k \times n}$$

$$\mathbf{s} \leftarrow \mathbb{F}_q^k$$

$$\mathbf{E} \leftarrow \chi_{p^m, \sigma}$$

$$\mathbf{e} \leftarrow \mathbf{E}^n$$

$$\mathbf{A} \leftarrow \mathbb{F}_q^{k \times n}$$

$$\mathbf{y} \leftarrow \mathbb{F}_q^n$$

A well studied assumption

- Enables efficient public-key encryption (NIST Round 2 candidates: ROLLO, RQC)
- Connections with the well-known MinRank problem
- Many cryptanalysis efforts

Reusing the same support

$$\begin{aligned} \text{ct}_1 &= (\mathbf{A}_1, \mathbf{A}_1 \mathbf{s} + \underbrace{\mathbf{e}_1}_{\substack{\mathbf{e}_2 \\ \text{same support } E}} + \text{Encode}(\boldsymbol{\mu}_1)) \\ \text{ct}_2 &= (\mathbf{A}_2, \mathbf{A}_2 \mathbf{s} + \underbrace{\mathbf{e}_2}_{\substack{\mathbf{e}_1 \\ \text{same support } E}} + \text{Encode}(\boldsymbol{\mu}_2)) \end{aligned}$$

$$\text{Eval}(+, \text{ct}_1, \text{ct}_2) = \text{ct}_1 + \text{ct}_2 = (\mathbf{A}_1 + \mathbf{A}_2, (\mathbf{A}_1 + \mathbf{A}_2) \mathbf{s} + \underbrace{\mathbf{e}_1 + \mathbf{e}_2}_{\sigma' \leq \sigma} + \text{Encode}(\boldsymbol{\mu}_1 + \boldsymbol{\mu}_2))$$

Security reduction (two ciphertexts)

$$\begin{pmatrix} s \end{pmatrix} = \begin{pmatrix} A_1 \\ A_2 \end{pmatrix} + \begin{pmatrix} e_1 \\ e_2 \end{pmatrix} + \begin{pmatrix} \text{Encode}(\mu_1) \\ \text{Encode}(\mu_2) \end{pmatrix}$$
$$\begin{pmatrix} v_1 \\ v_2 \end{pmatrix}$$

In rank metric, **totally standard** error distribution!

Rank Somewhat Homomorphic Encryption (RankSHE)

$$\text{sk} = \mathbf{s} \in \mathbb{F}_q^n, \text{Supp}(\mathbf{e}) = E, \mathbf{e}^\perp \notin E$$

$$\text{Enc}(\boldsymbol{\mu} \in \mathbb{F}_p^n, \text{sk}) = (\mathbf{u} \in \mathbb{F}_q^n, \mathbf{v} = \mathbf{u} \cdot \mathbf{s} + \underbrace{\mathbf{e}}_{\mathbf{e} \in E} + \mathbf{e}^\perp \cdot \boldsymbol{\mu})$$

$$\text{Dec}((\mathbf{u}, \mathbf{v}), \text{sk}) = \varphi_{\mathbf{e}^\perp}(\mathbf{v} - \mathbf{u} \cdot \mathbf{s})$$

In this work: $\text{Encode}(\boldsymbol{\mu}) = \mathbf{e}^\perp \cdot \boldsymbol{\mu}$ with $\mathbf{e}^\perp \notin E$ and $\boldsymbol{\mu} \in \mathbb{F}_p^n$

Proposition

The security of **RankSHE** with ℓ independent ciphertexts is reduced to the ℓ -IRDP problem (decoding in an ideal $[\ell n, n]_{p^m}$ code)

Linear forms

Definition

φ_{e^\perp} is an \mathbb{F}_p -linear form defined on \mathbb{F}_q such that:

$$\varphi_{e^\perp}(e^\perp) = 1$$

$$\varphi_{e^\perp}(e) = 0 \quad (e \in E)$$

Informally,

$$\varphi_{e^\perp}(x) = \frac{\langle e^\perp, x \rangle}{\langle e^\perp, e^\perp \rangle}$$

Definition

$$\varphi_{e^\perp}(\mathbf{x}) = (\varphi_{e^\perp}(x_i))_i$$

Linear forms

Lemma

For $\mathbf{v} \in \mathbb{F}_q^n$ and $\mu \in \mathbb{F}_p^n$,

$$\varphi_{e^\perp}(\mu \cdot \mathbf{v}) = \mu \cdot \varphi_{e^\perp}(\mathbf{v}).$$

Addition

$$\text{Enc}(\mu_1, \text{sk}) + \text{Enc}(\mu_2, \text{sk}) = \text{Enc}(\mu_1 + \mu_2, \text{sk})$$

$$\mathbf{v}_1 = \mathbf{u}_1 \cdot \mathbf{s} + \mathbf{e}_1 + \mathbf{e}^\perp \cdot \mu_1$$

$$\mathbf{v}_2 = \mathbf{u}_2 \cdot \mathbf{s} + \mathbf{e}_2 + \mathbf{e}^\perp \cdot \mu_2$$

$$\mathbf{v}_1 + \mathbf{v}_2 = (\mathbf{u}_1 + \mathbf{u}_2) \cdot \mathbf{s} + \mathbf{e}_1 + \mathbf{e}_2 + \mathbf{e}^\perp \cdot (\mu_1 + \mu_2)$$

Plaintext absorption

$$\mu_1 \cdot \text{Enc}(\mu_2, \text{sk}) = \text{Enc}(\mu_1 \cdot \mu_2, \text{sk})$$

$$v_2 = u_2 \cdot s + e_2 + e^\perp \cdot \mu_2$$

$$\mu_1 \cdot v_2 = (\mu_1 \cdot u_2) \cdot s + \mu_1 \cdot e_2 + e^\perp \cdot (\mu_1 \cdot \mu_2)$$

Multiplication

$$\text{Eval}(\times, (\mathbf{u}_1, \mathbf{v}_1), (\mathbf{u}_2, \mathbf{v}_2)) = (\mathbf{u}_1 \cdot \mathbf{u}_2, \mathbf{u}_1 \cdot \mathbf{v}_2 + \mathbf{u}_2 \cdot \mathbf{v}_1, \mathbf{v}_1 \cdot \mathbf{v}_2)$$

$$\begin{aligned}\mathbf{u}_1 \cdot \mathbf{u}_2 \cdot \mathbf{s}^2 - (\mathbf{u}_1 \cdot \mathbf{v}_2 + \mathbf{u}_2 \cdot \mathbf{v}_1) \cdot \mathbf{s} + \mathbf{v}_1 \cdot \mathbf{v}_2 \\= (\mathbf{v}_1 - \mathbf{u}_1 \cdot \mathbf{s}) \cdot (\mathbf{v}_2 - \mathbf{u}_2 \cdot \mathbf{s}) \\= (\mathbf{e}_1 + \mathbf{e}^\perp \cdot \boldsymbol{\mu}_1) \cdot (\mathbf{e}_2 + \mathbf{e}^\perp \cdot \boldsymbol{\mu}_2) \\= \underbrace{\mathbf{e}_1 \cdot \mathbf{e}_2 + \mathbf{e}^\perp \cdot (\mathbf{e}_1 \cdot \boldsymbol{\mu}_2 + \mathbf{e}_2 \cdot \boldsymbol{\mu}_1)}_{\mathbf{e}', \text{Supp}(\mathbf{e}') \subset E^2 \oplus \mathbf{e}^\perp E} + (\mathbf{e}^\perp)^2 \cdot \boldsymbol{\mu}_1 \cdot \boldsymbol{\mu}_2\end{aligned}$$

$$\text{DecAfterMul}((\mathbf{u}, \mathbf{v}, \mathbf{w}), \text{sk}) = \varphi_{(\mathbf{e}^\perp)^2}(\mathbf{u} \cdot \mathbf{s}^2 - \mathbf{v} \cdot \mathbf{s} + \mathbf{w})$$

Summary

- Encryption scheme based on ideal random rank metric codes
- Unlimited additions
- Multiplication adds a component to the ciphertext and increases noise quadratically

Reducing ciphertext noise

$(\mathbf{u}, \mathbf{v}, \mathbf{w})$ decryptable under sk_1



(\mathbf{u}, \mathbf{v}) decryptable under sk_2

Bootstrapping

$$\widehat{\text{Eval}}(\text{DecAfterMul}(\cdot, \cdot), \underbrace{\text{ct}_1}_{\text{Enc}((\mathbf{u}, \mathbf{v}, \mathbf{w}), \text{sk}_2)}, \underbrace{\text{ct}_2}_{\text{Enc}(\text{sk}_1, \text{sk}_2)}) \approx \text{Enc}(\mu, \text{sk}_2)$$

$$\widehat{\text{Eval}}(\text{DecAfterMul}(\cdot, \cdot), \underbrace{\text{ct}_1}_{(\mathbf{u}, \mathbf{v}, \mathbf{w})}, \underbrace{\text{ct}_2}_{\text{Enc}(\varphi_{e^\perp}(\text{sk}_1), \text{sk}_2)}) \approx \text{Enc}(\mu, \text{sk}_2)$$

Bootstrapping

$$\text{DecAfterMul}((\mathbf{u}, \mathbf{v}, \mathbf{w}), \text{sk}_1) = \varphi_{(\mathbf{e}^\perp)^2}(\mathbf{u} \cdot \mathbf{s}_1^2 - \mathbf{v} \cdot \mathbf{s}_1 + \mathbf{w})$$

$$\mathbf{u} = \sum_i \gamma_i \mathbf{u}^{(i)}$$

$$\mathbf{v} = \sum_i \gamma_i \mathbf{v}^{(i)}$$

$$\mathbf{w} = \sum_i \gamma_i \mathbf{w}^{(i)}$$

$$\text{DecAfterMul}((\mathbf{u}, \mathbf{v}, \mathbf{w}), \text{sk}_1) = \sum_i \underbrace{\mathbf{u}^{(i)}}_{\in \mathbb{F}_p} \cdot \underbrace{\mathbf{a}^{(i)}}_{\in \mathbb{F}_q} - \mathbf{v}^{(i)} \cdot \mathbf{b}^{(i)} + \mathbf{w}^{(i)} \cdot \mathbf{c}^{(i)}$$

Bootstrapping

$$\begin{aligned}\varphi_{(\mathbf{e}^\perp)^2}(\mathbf{u} \cdot \mathbf{s}_1^2) &= \sum_i \varphi_{(\mathbf{e}^\perp)^2}(\gamma_i \mathbf{u}^{(i)} \cdot \mathbf{s}_1^2) \\ &= \sum_i \varphi_{(\mathbf{e}^\perp)^2}(\mathbf{u}^{(i)} \cdot \gamma_i \mathbf{s}_1^2) \\ &= \sum_i \mathbf{u}^{(i)} \cdot \varphi_{(\mathbf{e}^\perp)^2}(\gamma_i \mathbf{s}_1^2) \\ &= \sum_i \mathbf{u}^{(i)} \cdot \mathbf{a}^{(i)}\end{aligned}$$

Bootstrapping

$$\mu = \text{DecAfterMul}((\mathbf{u}, \mathbf{v}, \mathbf{w}), \text{sk}_1) = \sum_i \mathbf{u}^{(i)} \cdot \mathbf{a}^{(i)} - \mathbf{v}^{(i)} \cdot \mathbf{b}^{(i)} + \mathbf{w}^{(i)} \cdot \mathbf{c}^{(i)}$$

with

$$\mathbf{a}^{(i)} = \varphi_{(\mathbf{e}^\perp)^2}(\gamma_i \mathbf{s}_1^2)$$

$$\mathbf{b}^{(i)} = \varphi_{(\mathbf{e}^\perp)^2}(\gamma_i \mathbf{s}_1)$$

$$\mathbf{c}^{(i)} = \varphi_{(\mathbf{e}^\perp)^2}(\gamma_i (1, 0, \dots, 0))$$

Bootstrapping

$$\text{ct}_{\mathbf{a}^{(i)}} = \text{Enc}(\mathbf{a}^{(i)}, \text{sk}_2) = \text{Enc}(\varphi_{(\mathbf{e}^\perp)^2}(\gamma_i \mathbf{s}_1^2), \text{sk}_2)$$

$$\text{ct}_{\mathbf{b}^{(i)}} = \text{Enc}(\mathbf{b}^{(i)}, \text{sk}_2) = \text{Enc}(\varphi_{(\mathbf{e}^\perp)^2}(\gamma_i \mathbf{s}_1), \text{sk}_2)$$

$$\text{ct}_{\mathbf{c}^{(i)}} = \text{Enc}(\mathbf{c}^{(i)}, \text{sk}_2) = \text{Enc}(\varphi_{(\mathbf{e}^\perp)^2}(\gamma_i(1, 0, \dots, 0)), \text{sk}_2)$$

$$\begin{aligned}\widehat{\text{Eval}}(\text{DecAfterMul}, (\mathbf{u}, \mathbf{v}, \mathbf{w}), \text{ct}_{\mathbf{a}^{(i)}}, \text{ct}_{\mathbf{b}^{(i)}}, \text{ct}_{\mathbf{c}^{(i)}}) \\ := \sum_i \mathbf{u}^{(i)} \cdot \text{ct}_{\mathbf{a}^{(i)}} - \mathbf{v}^{(i)} \cdot \text{ct}_{\mathbf{b}^{(i)}} + \mathbf{w}^{(i)} \cdot \text{ct}_{\mathbf{c}^{(i)}} \\ = \text{Enc}\left(\sum_i \mathbf{u}^{(i)} \cdot \mathbf{a}^{(i)} - \mathbf{v}^{(i)} \cdot \mathbf{b}^{(i)} + \mathbf{w}^{(i)} \cdot \mathbf{c}^{(i)}, \text{sk}_2\right) \\ = \text{Enc}(\boldsymbol{\mu}, \text{sk}_2)\end{aligned}$$

Bootstrapping

Our bootstrapping algorithm:

- Transforms a three-component ciphertext into a two-component ciphertext;
- reduces the noise from $\approx E_1^2$ to E_2 ;
- has no multiplicative cost;
- but... requires $3m$ independent ciphertexts under sk_2 .

Security reduction ($3m$ ciphertexts)

The attacker needs to solve the RSD problem in an ideal $[3mn, n]_{p^m}$ code.

There exists a polynomial attack [GRS13] in an $[n, k]_{p^m}$ code when

$$(k + 1)(\sigma m + 1) \leq n + 1.$$

\implies maximal number of independent ciphertexts $\approx \sigma m$.

Reducing the number of bootstrapping ciphertexts

We pack several plaintexts into a single ciphertext:

$$\text{Enc}((\mu_1, \dots, \mu_t) \in (\mathbb{F}_p^n)^t, \text{sk}) = (\mathbf{u} \in \mathbb{F}_q^n, \mathbf{v} = \mathbf{u} \cdot \mathbf{s} + \underbrace{\mathbf{e}}_{\|\mathbf{e}\| \leq w} + \sum_{i=1}^t \chi^i \mathbf{e}^\perp \cdot \mu_i)$$

with $\chi \in \mathbb{F}_q$ s.t. $\chi^t = 1$.

Maximal packing index $t = \frac{1}{\sigma}$.

\implies reduces the number of bootstrapping plaintexts to $\frac{3m}{t} = 3\sigma m$.

Summary

- Encryption scheme based on ideal random rank metric codes
- Unlimited additions
- Multiplication adds a component to the ciphertext and increases noise quadratically
- Limited number of ciphertexts (sublinear in ciphertext size)

Parameters

d	q	m	n	w	ℓ	Security	Key size	ct size	Add	Mul	Bootstrap
1	2	172	20	13	9	128	3.7 kB	0.9 kB	0.002 ms	0.5 ms	2 ms
2	2	367	183	7	5	128	17.0 kB	16.8 kB	0.04 ms	52 ms	374 ms
3	2	1296	314	6	4	128	210 kB	102 kB	0.3 ms	944 ms	11 s
4	2	3125	713	5	3	128	1.22 MB	557 kB	1 ms	14.3 s	239 s

Table: Example of parameters for our SHE scheme, with associated sizes and execution timings. d is the number of possible multiplications. q , m and n are parameters of the rank linear code and w is the rank weight of the error. ℓ is the number of independant ciphertexts that can be published.

Comparison

Scheme	ct size	Bootstrap ct size	Mul time	Bootstrap time
TFHE [CGGI20] [AAPS11]	2 kB 18.5 kB	15.6 MB -	0.03 ms 10 ms	13 ms -
This work	0.9 kB	35 kB	0.5 ms	2 ms

Table: Parameters are taken for 128-bit security, and for SHE schemes, with a single multiplication allowed.

Open problems

- Reduce the number of bootstrapping ciphertexts
- Demonstrate a practical application

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Thank you for your attention!

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