

On the Secret Distributions in Module Learning With Errors

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The Need For Post-Quantum Cryptography

The security of currently deployed public-key cryptography relies on Factoring and Discrete Logarithm.



? What if we had a powerful **Quantum Computer** ?

The Need For Post-Quantum Cryptography

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? What if we had a powerful **Quantum Computer** ?



Exponential quantum speed-up with **Shor's** algorithm [Sho94]: factoring and discrete logarithm solvable in $poly(\lambda)$:  \implies 

 **Hardness assumptions underlying RSA/ECC no longer valid.** 

Need: Design new cryptosystems from new mathematical problems that are hard to solve, even quantumly. And fast...

Future NIST PQC Standards

NIST **PQC standardization process** launched in 2016. First round of standardized algorithms announced in July 2022:

Encryption	Signature
Crystals-Kyber	Crystals-Dilithium
	Falcon
	SPHINCS+

M-LWE

lattice-based

NSA has already announced its CNSA Suite 2.0 for Quantum-Resistant algorithms. It includes **Kyber** and **Dilithium**.



How robust is Module Learning With Errors with such short distributions? **Let's see**

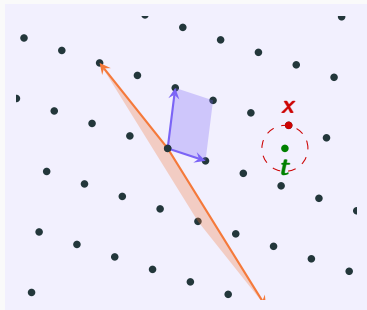
Problem
Reduction
Proof
Secret
Module
Key
Field Attack
Euclidean
Lattice
Encryption Signature
Cryptography
Post-Quantum
Distribution
Security
Bounded
Error
Vector

You Said Lattice?

Euclidean Lattice

$$\mathcal{L} = \left\{ \begin{array}{|c} \boxed{B} \\ \hline \mathbf{x} \end{array} ; \mathbf{x} \in \mathbb{Z}^n \right\}$$

with basis $B \in \mathbb{R}^{n \times n}$.



CVP:

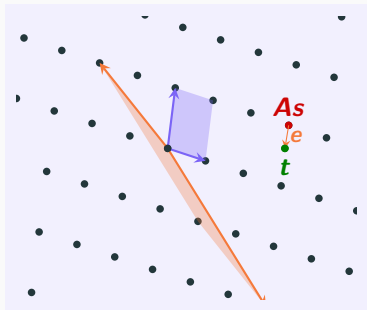
Given a target \mathbf{t} , find $\mathbf{x} \in \mathcal{L}$ that minimizes $\|\mathbf{x} - \mathbf{t}\|$.

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CVP:

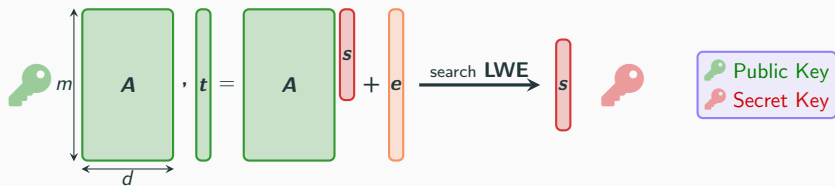
Given a target \mathbf{t} , find $\mathbf{x} \in \mathcal{L}$ that minimizes $\|\mathbf{x} - \mathbf{t}\|$.

Given $\mathbf{A} \in \mathbb{Z}_q^{m \times d}$ describing the lattice

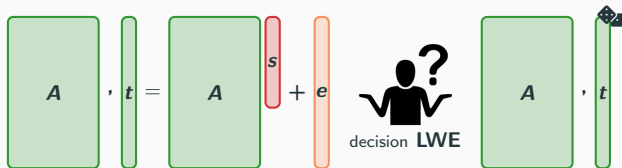
$$\mathcal{L}_q(\mathbf{A}) = \{ \mathbf{x} \in \mathbb{Z}^m : \exists \mathbf{s} \in \mathbb{Z}_q^d, \mathbf{A}\mathbf{s} = \mathbf{x} \bmod q \}$$

and $\mathbf{t} = \mathbf{A}\mathbf{s} + \mathbf{e} \bmod q$, solve $\text{CVP}_{\mathbf{t}}$ on $\mathcal{L}_q(\mathbf{A})$. This is **LWE**!

Learning With Errors

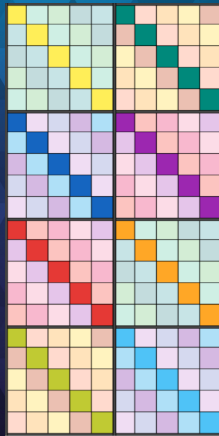
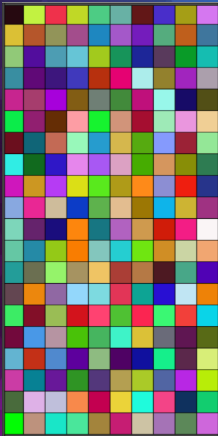


where $A \leftarrow \text{Unif}(\mathbb{Z}_q^{m \times d})$, $s \leftarrow \mathcal{D}_s$ (over \mathbb{Z}^d), and $e \leftarrow \text{Gauss}(\mathbb{Z}^m)$.



Standard Secret [Reg05]: $\mathcal{D}_s = \text{Unif}(\mathbb{Z}_q^d)$
Binary Secret [BLP+13]: $\mathcal{D}_s = \text{Unif}(\{0, 1\}^d)$
General Secret [BD20a]: \mathcal{D}_s arbitrary, with enough entropy

Reduce needed storage and
speed-up computations by
adding **Structure**



Adding an Algebraic Structure for More Efficiency



Replace \mathbb{Z} with a ring $\mathcal{R} = \mathbb{Z}[x]/\langle f(x) \rangle$, e.g., $f(x) = x^n + 1$ with $n = 2^\ell$ and \mathbb{Z}_q by $\mathcal{R}_q = \mathbb{Z}_q[x]/\langle f(x) \rangle$

$$\sum_{i=0}^{n-1} a_i \cdot x^i \in \mathcal{R} \xleftrightarrow{\text{embedding}} \begin{bmatrix} a_0 \\ \vdots \\ a_{n-1} \end{bmatrix} \in \mathbb{Z}^n$$

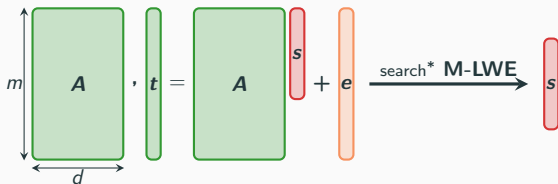
$$\left(\sum_{i=0}^{n-1} a_i \cdot x^i \right) \cdot \left(\sum_{i=0}^{n-1} b_i \cdot x^i \right) \xleftrightarrow{\text{embedding}} \begin{bmatrix} \text{Rot}(\mathbf{a}) \\ \vdots \\ \text{Rot}(\mathbf{a}) \end{bmatrix} \cdot \begin{bmatrix} b_0 \\ \vdots \\ b_{n-1} \end{bmatrix}$$

The diagram illustrates the mapping of polynomial multiplication in the ring \mathcal{R} to matrix-vector multiplication in \mathbb{Z}^n . The top part shows a polynomial $\sum_{i=0}^{n-1} a_i \cdot x^i$ being mapped via an embedding to a column vector $\begin{bmatrix} a_0 \\ \vdots \\ a_{n-1} \end{bmatrix} \in \mathbb{Z}^n$. The bottom part shows the product of two polynomials $\left(\sum_{i=0}^{n-1} a_i \cdot x^i \right) \cdot \left(\sum_{i=0}^{n-1} b_i \cdot x^i \right)$ being mapped to a matrix-vector product. The matrix is a circulant matrix, represented as a grid of colored cells (red, pink, blue, green, yellow, orange) with a central box labeled $\text{Rot}(\mathbf{a})$. The vector is $\begin{bmatrix} b_0 \\ \vdots \\ b_{n-1} \end{bmatrix}$.

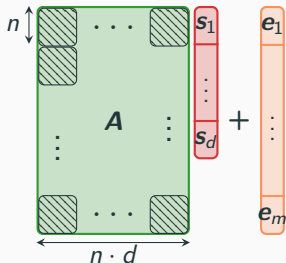
Efficiency: FFT-like algorithms, use of structured matrices.

Storage: Structured matrices represented by a single vector.

Module Learning With Errors as Structured LWE



where $\mathbf{A} \leftarrow \text{Unif}(\mathcal{R}_q^{m \times d})$, $\mathbf{s} \leftarrow \mathcal{D}_s$ (over \mathcal{R}^d), and $\mathbf{e} \leftarrow \text{Gauss}(\mathcal{R}^m)$.



Structured version of LWE
in dimensions nm & nd

*The decision problem is to distinguish such \mathbf{t} from $\text{Unif}(\mathcal{R}_q^m)$

What do we know so far?

Distributions	LWE	M-LWE
$\mathcal{D}_s = \text{Unif}(\mathcal{R}_q^d)$ $\mathcal{D}_e = \text{Gauss}(\mathcal{R}^m)$	[Reg05] [BLP+13]	[LS15] ?
$\mathcal{D}_s = \text{Unif}(S_1^d)$ $\mathcal{D}_e = \text{Gauss}(\mathcal{R}^m)$	[GKPV10] [BLP+13] [Mic18]	? ? ?
$\mathcal{D}_s = \text{Unif}(\mathcal{R}_q^d)$ $\mathcal{D}_e = \text{Unif}(S_1^m)$	[MP13]	?
\mathcal{D}_s arbitrary $\mathcal{D}_e = \text{Gauss}(\mathcal{R}^m)$	[BD20a] [BD20b] (R-LWE)	[LWW20] ?

$$S_1 = \{0, 1\}[x]/\langle x^n + 1 \rangle$$

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$\mathcal{D}_s = \text{Unif}(\mathcal{R}_q^d)$ $\mathcal{D}_e = \text{Gauss}(\mathcal{R}^m)$	[Reg05] [BLP+13]	[LS15] ④ [BJRW20]
$\mathcal{D}_s = \text{Unif}(S_1^d)$ $\mathcal{D}_e = \text{Gauss}(\mathcal{R}^m)$	[GKPV10] [BLP+13] [Mic18]	① [BJRW20] ② [BJRW21] ?
$\mathcal{D}_s = \text{Unif}(\mathcal{R}_q^d)$ $\mathcal{D}_e = \text{Unif}(S_1^m)$	[MP13]	⑤ [BJRW23]
\mathcal{D}_s arbitrary $\mathcal{D}_e = \text{Gauss}(\mathcal{R}^m)$	[BD20a] [BD20b] (R-LWE)	[LWW20] ③ [BJRW22]

$$S_1 = \{0, 1\}^{\langle x \rangle} / \langle x^n + 1 \rangle$$

① M-LWE is still hard with **small** s and **Gaussian** e ;

Today

② Decisional M-LWE is still hard with **small** s and **Gaussian** e ;

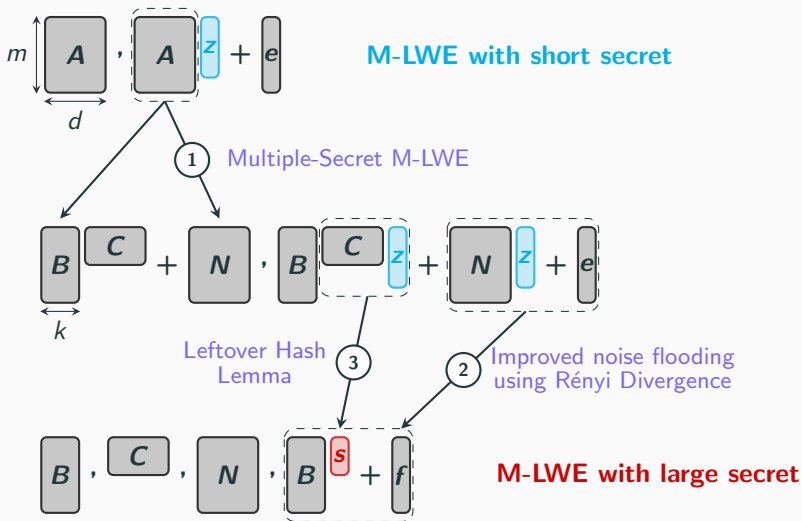
③ M-LWE is still hard with **arbitrary** s , if it has enough entropy.

And now...

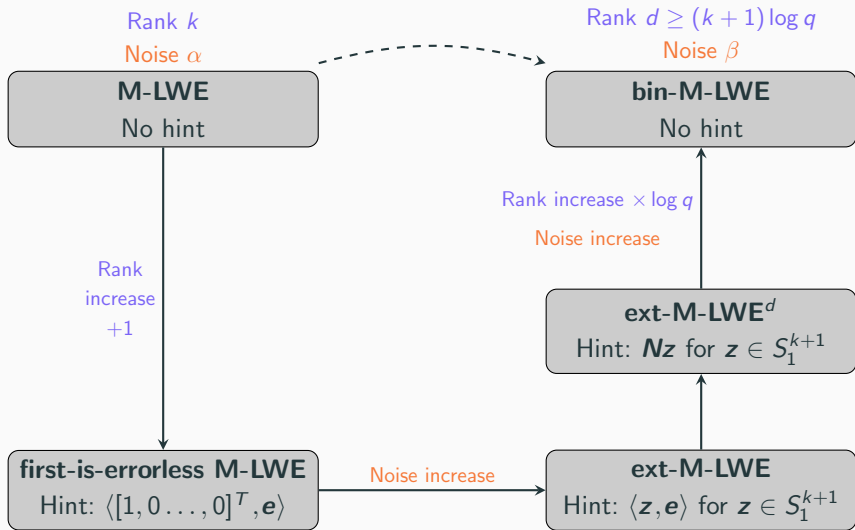
90%
DE RÉDUCTION

① Computational Hardness of M-LWE with Short Secret

The secret z is small (S_1^d) and the secret s is large (\mathcal{R}_q^k).

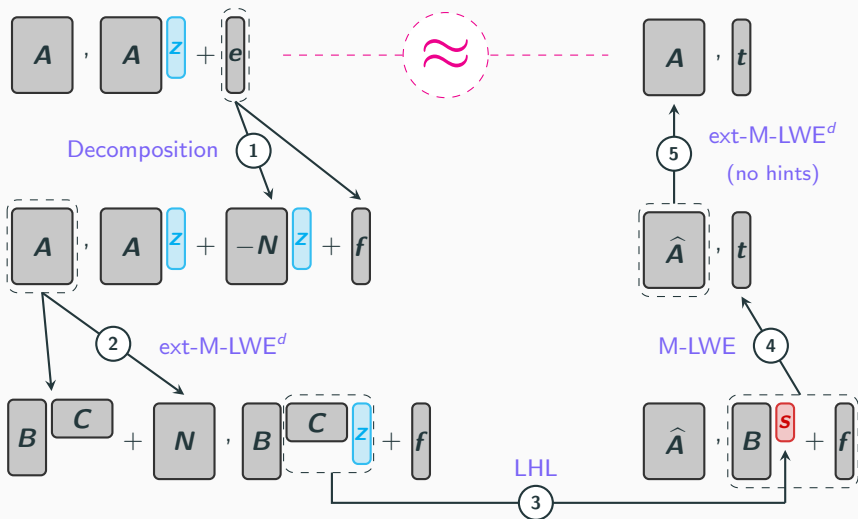


2 Pseudorandomness of M-LWE with Short Secret (1/2)

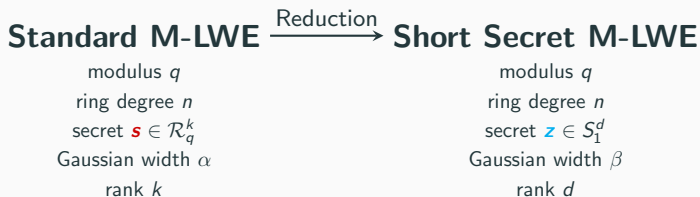


2 Pseudorandomness of M-LWE with Short Secret (2/2)

The secret z is small (S_1^d) and the secret s is large (\mathcal{R}_q^k).



Hardness of Module-LWE with Short Secret: Sum-Up



Property	Contribution ①	Contribution ②
Minimal rank d	$k \log q + \Omega(\log n)$	$(k + 1) \log q + \omega(\log n)$
Noise ratio β/α	$O(n^2 \sqrt{md})$	$O(n^2 \sqrt{d})$
Conditions on q	prime	other restrictions*
Decision/Search	search	decision



Both proofs have their (dis)advantages

*In power-of-two cyclotomic fields, q must be prime such that $q = 5 \pmod{8}$.



- What about **non-uniform** secrets? •
- What about **smaller ranks**? •

Hardness of Module-LWE with Entropic Secret

Motivation: **Leakage resilience** of M-LWE-based systems

1. Physical attack to recover a noisy secret \tilde{s} .



2. Target a new M-LWE instance

$$\Delta t = A\tilde{s} - t = A \begin{matrix} 0 \\ \tilde{s} \end{matrix} - e$$



Under what condition on s' is the problem still hard?
 s' must have enough **entropy** \rightarrow **Entropic hardness**

Intuition: **Lossiness**

$H_\infty(s' | A, As' + e)$ large \implies M-LWE instance with secret s' hard

What About Module-NTRU?

NTRU

$$a \approx g \cdot f_q^{-1}$$

$$a \sim \text{Unif}(\mathcal{R}_q), f, g \sim \text{Gauss}(\mathcal{R})$$

f_q^{-1} inverse of f in \mathcal{R}_q

NTRU Learning

m

$$a \approx g \cdot f_q^{-1}$$

$g \sim \text{Gauss}(\mathcal{R}^m), f \sim \text{Gauss}(\mathcal{R})$

Module-NTRU

m

$A \approx G$

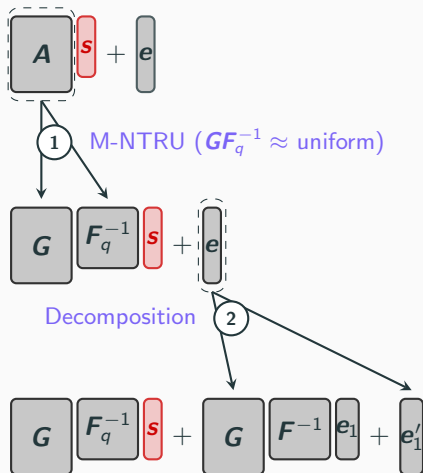
F_q^{-1}

d

$G \sim \text{Gauss}(\mathcal{R}^{m \times d}), F \sim \text{Gauss}(\mathcal{R}^{d \times d})$

Entropic Hardness of M-LWE from M-NTRU

Replacing \mathbf{A} by \mathbf{GF}_q^{-1} , with \mathbf{F}, \mathbf{G} Gaussian and $\mathbf{F}_q^{-1} = (\mathbf{F} \bmod q\mathcal{R})^{-1}$.
The secret \mathbf{s} is only assumed to have **large enough entropy**.



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Replacing \mathbf{A} by \mathbf{GF}_q^{-1} , with \mathbf{F}, \mathbf{G} Gaussian and $\mathbf{F}_q^{-1} = (\mathbf{F} \bmod q\mathcal{R})^{-1}$.
 The secret \mathbf{s} is only assumed to have **large enough entropy**.

$$\boxed{\mathbf{A}} \boxed{\mathbf{s}} + \boxed{\mathbf{e}}$$

① M-NTRU ($\mathbf{GF}_q^{-1} \approx \text{uniform}$)

$$\boxed{\mathbf{G}} \boxed{\mathbf{F}_q^{-1}} \boxed{\mathbf{s}} + \boxed{\mathbf{e}}$$

Decomposition

$$\boxed{\mathbf{G}} \boxed{\mathbf{F}_q^{-1}} \boxed{\mathbf{s}} + \boxed{\mathbf{G}} \boxed{\mathbf{F}^{-1}} \boxed{\mathbf{e}_1} + \boxed{\mathbf{e}'_1}$$

$$\boxed{\mathbf{G}} \left(\boxed{\mathbf{F}_q^{-1}} \boxed{\mathbf{s}} + \boxed{\mathbf{F}^{-1}} \boxed{\mathbf{e}_1} \right) + \boxed{\mathbf{e}'_1}$$

$$\begin{aligned} & H_\infty(\mathbf{s} | \mathbf{G}(\mathbf{F}_q^{-1}\mathbf{s} + \mathbf{F}^{-1}\mathbf{e}_1) + \mathbf{e}'_1) \\ & \geq H_\infty(\mathbf{s} | \mathbf{F}_q^{-1}\mathbf{s} + \mathbf{F}^{-1}\mathbf{e}_1) \\ & \geq H_\infty(\mathbf{s} | \mathbf{F}_q^{-1}\mathbf{s} + \mathbf{F}^{-1}\mathbf{e}_2) \quad (\mathbf{e}_2 \in \mathcal{L}(\mathbf{F})) \\ & = H_\infty(\mathbf{s} | \mathbf{s} + \mathbf{e}_2) \\ & \geq H_\infty(\mathbf{s} | \mathbf{s} + \mathbf{e}') - nd \log_2 \|\mathbf{F}\|_2 \\ & \geq H_\infty(\mathbf{s}) - nd \log_2 \frac{q}{\sigma_{\mathbf{e}'}} - nd \log_2 \|\mathbf{F}\|_2 \end{aligned}$$

② $\sigma_{\mathbf{e}} > \sigma_{\mathbf{e}'} \|\mathbf{GF}^{-1}\|_2$ ↑ Singular Values to optimize

Our contributions

- ✓ **Hardness** of a main problem, with (close to) **practical parameters**.
- ✓ Sufficient conditions on the secret distribution for **leakage resilience** of M-LWE.

Related Work

- 📄 Other reduction in [LWW20] from Module-LWE (uniform secret) to Module-LWE (entropic secret).
 - ✗ Not rank-preserving.
 - ✓ Assumption proven on module lattices.
 - ▢ Parameter regimes with sometimes better or worse results.

Open Questions

- ? Prove the hardness of Module-LWE with **low-entropy secret** distributions without increasing the rank

Thank you for your
attention!



Questions?



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Singular Values of Multiplication Matrices

