## On the Hardness of Module Learning With Errors with Short Distributions

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## Reminder: Symmetric and Asymmetric Cryptography

Symmetric Cryptography

! Alice and Bob must agree on the same key

Asymmetric Cryptography


## The Need For Post-Quantum Cryptography

## What if we had a Cryptographically Relevant Quantum Computer ${ }^{1}$ ?

^ Quadratic quantum speed-up with Grover's algorithm [Gro96]: exhaustive key search of $\rho$ in $O(\sqrt{\# \text { key space }})$;

Exponential quantum speed-up with Shor's algorithm [Sho97]: factoring and discrete logarithm in poly $(\log n) \Longrightarrow \int$

The underlying hardness assumptions of modern cryptography (RSA, ECC) would no longer be valid.

Need: Design new cryptosystems from new mathematical problems that are hard to solve, even by a CRQC. And fast...
${ }^{1}$ NSA FAQ on Quantum Computing and Post-Quantum Cryptography

## Potential Candidates: NIST PQC Standardization

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

| Encryption | Signature |
| :--- | :--- |
| Crystals-Kyber | Crystals-Dilithium |
|  | Falcon |
|  | SPHINCS + |

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 <br> Proof Secret oaute <br> Key Field <br> Cryptography <br> Post-Quantum <br> Distribution <br> Security <br> Errofor

## You Said Lattice?



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

## You Said Lattice?

## Euclidean Lattice

$\mathcal{L}=\left\{B ; x \in \mathbb{Z}^{n}\right\}$ with basis $B \in \mathbb{R}^{n \times n}$.


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

Given $\boldsymbol{A} \in \mathbb{Z}_{q}^{m \times d}$ describing the lattice

$$
\mathcal{L}_{q}(\boldsymbol{A})=\left\{\boldsymbol{x} \in \mathbb{Z}^{m}: \exists \boldsymbol{s} \in \mathbb{Z}_{q}^{d}, \boldsymbol{A} \boldsymbol{s}=\boldsymbol{x} \bmod q\right\}
$$

and $\boldsymbol{t}=\boldsymbol{A s}+e \bmod q$, solve $\mathbf{C V P}_{\boldsymbol{t}}$ on $\mathcal{L}_{q}(\boldsymbol{A})$. This is LWE!

## Learning With Errors

Set $\mathbb{Z}_{q}=\mathbb{Z} / q \mathbb{Z}$ for some integer $q$.

where $\boldsymbol{A} \hookleftarrow \operatorname{Unif}\left(\mathbb{Z}_{q}^{m \times d}\right)$, $\boldsymbol{s} \hookleftarrow \mathcal{D}_{s}\left(\right.$ over $\left.\mathbb{Z}^{d}\right)$, and $e \hookleftarrow \mathcal{D}_{e}\left(\right.$ over $\left.\mathbb{Z}^{m}\right)$.

$$
\begin{array}{lll}
\text { Standard [Reg05]: } & \mathcal{D}_{s}=\operatorname{Unif}\left(\mathbb{Z}_{q}^{d}\right) & \mathcal{D}_{e}=\operatorname{Gauss}\left(\mathbb{Z}^{m}\right) \\
\text { Binary Secret }[B L P+13]: & \mathcal{D}_{s}=\operatorname{Unif}\left(\{0,1\}^{d}\right) & \mathcal{D}_{e}=\operatorname{Gauss}\left(\mathbb{Z}^{m}\right) \\
\text { Binary Error [MP13]: } & \mathcal{D}_{s}=\operatorname{Unif}\left(\mathbb{Z}_{q}^{d}\right) & \mathcal{D}_{e}=\operatorname{Unif}\left(\{0,1\}^{m}\right)
\end{array}
$$

[^0]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 $\mathcal{R}_{q}=\mathbb{Z}_{q}[x] /\langle f(x)\rangle$

$$
\sum_{i=0}^{n-1} a_{i} \cdot x^{i} \in \mathcal{R} \stackrel{\text { embedding }}{\longleftrightarrow}\left[\begin{array}{c}
a_{0} \\
\vdots \\
a_{n-1}
\end{array}\right] \in \mathbb{Z}^{n}
$$

$$
\left(\sum_{i=0}^{n-1} a_{i} \cdot x^{i}\right) \cdot\left(\sum_{i=0}^{n-1} b_{i} \cdot x^{i}\right) \longleftrightarrow \text { Rot (a) }\left[\begin{array}{c}
b_{0} \\
\vdots \\
b_{n-1}
\end{array}\right]
$$

Efficiency: FFT-like algorithms, use of structured matrices.
Storage: Structured matrices represented by a single vector.

## Module Learning With Errors as Structured LWE


where $\boldsymbol{A} \hookleftarrow \operatorname{Unif}\left(\mathcal{R}_{q}^{m \times d}\right), \boldsymbol{s} \hookleftarrow \mathcal{D}_{s}\left(\right.$ over $\left.\mathcal{R}^{d}\right)$, and $e \hookleftarrow \mathcal{D}_{e}\left(\right.$ over $\left.\mathcal{R}^{m}\right)$.
A good choice would be over $S_{1}=\{0,1\}[x] /\left\langle x^{n}+1\right\rangle$.


Structured version of LWE in dimensions $n m$ \& $n d$

[^1]
## What do we know so far?

| Distributions | LWE | M-LWE |
| :---: | :---: | :---: |
| $\mathcal{D}_{s}=\operatorname{Unif}\left(\mathcal{R}_{q}^{d}\right)$ | [Reg05] | [LS15] |
| $\mathcal{D}_{e}=\operatorname{Gauss}\left(\mathcal{R}^{m}\right)$ | [BLP $\left.{ }^{+13}\right]$ | ? |
| $\begin{aligned} & \mathcal{D}_{s}=\operatorname{Unif}\left(S_{1}^{d}\right) \\ & \mathcal{D}_{e}=\operatorname{Gauss}\left(\mathcal{R}^{m}\right) \end{aligned}$ | [GKPV10] | ? |
|  | [BLP $\left.{ }^{+13}\right]$ | ? |
|  | [Mic18] | ? |
| $\begin{aligned} & \mathcal{D}_{s}=\operatorname{Unif}\left(\mathcal{R}_{q}^{d}\right) \\ & \mathcal{D}_{e}=\operatorname{Unif}\left(S_{1}^{m}\right) \end{aligned}$ | [MP13] | ? |
| $\mathcal{D}_{s}$ arbitrary | [BD20a] | [LWW20] |
| $\mathcal{D}_{e}=\operatorname{Gauss}\left(\mathcal{R}^{m}\right)$ | [BD20b] (R-LWE) | ? |

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| $\mathcal{D}_{e}=\operatorname{Gauss}\left(\mathcal{R}^{m}\right)$ | [BLP $\left.{ }^{+13}\right]$ | (4) [BJRW20] |
| $\begin{aligned} & \mathcal{D}_{s}=\operatorname{Unif}\left(S_{1}^{d}\right) \\ & \mathcal{D}_{e}=\operatorname{Gauss}\left(\mathcal{R}^{m}\right) \end{aligned}$ | [GKPV10] | (1) [BJRW20] |
|  | [BLP $\left.{ }^{+13}\right]$ | (2) [BJRW21] |
|  | [Mic18] | ? |
| $\begin{aligned} & \mathcal{D}_{s}=\operatorname{Unif}\left(\mathcal{R}_{q}^{d}\right) \\ & \mathcal{D}_{e}=\operatorname{Unif}\left(S_{1}^{m}\right) \end{aligned}$ | [MP13] | (3) [BJRW22b] |
| $\mathcal{D}_{s}$ arbitrary | [BD20a] | [LWW20] |
| $\mathcal{D}_{e}=\operatorname{Gauss}\left(\mathcal{R}^{m}\right)$ | [BD20b] (R-LWE) | 5 [BJRW22a] |

(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 small $d$ and $e$, if $m$ is not too large.

## And now...



## (1) Computational Hardness of M-LWE with Short Secret

The secret $z$ is small $\left(S_{1}^{d}\right)$ and the secret $\boldsymbol{s}$ is large $\left(\mathcal{R}_{q}^{k}\right)$.


## (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 $\left(S_{1}^{d}\right)$ and the secret $\boldsymbol{s}$ is large $\left(\mathcal{R}_{q}^{k}\right)$.


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

## Standard M-LWE $\xrightarrow{\text { Reduction }}$ Short Secret M-LWE

$$
\text { modulus } q
$$

ring degree $n$
secret $s \in \mathcal{R}_{q}^{k}$
Gaussian width $\alpha$ rank $k$
modulus $q$
ring degree $n$
secret $z \in S_{1}^{d}$
Gaussian width $\beta$
rank d

| Property | Contribution (1) | Contribution (2) |
| :--- | :--- | :--- |
| Minimal rank $d$ | $k \log q+\Omega(\log n)$ | $(k+1) \log q+\omega(\log n)$ |
| Noise ratio $\beta / \alpha$ | $O\left(n^{2} \sqrt{m} d\right)$ | $O\left(n^{2} \sqrt{d}\right)$ |
| Conditions on $q$ | prime | other restrictions ${ }^{4}$ |
| Decision/Search | search | decision |

## Both proofs have their (dis)advantages

[^2]
## 3 Computational Hardness of M-LWE With Short Error

Idea: Prove that $(\boldsymbol{s}, \boldsymbol{e}) \mapsto \boldsymbol{A} \boldsymbol{s}+\boldsymbol{e}$ is one-way when $\boldsymbol{e}$ has small uniform coefficients. Reason on the dual function $\boldsymbol{e} \mapsto \boldsymbol{B}^{T} \boldsymbol{e}$.

## Uninvertible is not enough.

Result: It is one-way if $\boldsymbol{A}$ is not too tall, i.e., $m$ not too large. Why?


Lots of 0 if $e$ has small coefficients

## Wrapping Up

## Our contributions

$\checkmark$ Hardness of a main problem, with (close to) practical parameters.

## Lattice-based Cryptography

O Most promising PQC successor of RSA/ECC.
\$8 Mathematical problems on lattices that are (confidently assumed) hard to solve even for a quantum computer.

What's next?
? Keep closing the gap between provably secure parameter sets and the ones used in practice (small ones).
$\approx$ Use these stretched assumptions to design efficient PQC schemes (done, see NIST) with additional features (ok there is still work to do).

## Thank you for your attention!



Questions?

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[^0]:    ${ }^{2}$ The decision problem is to distinguish such $t$ from $\operatorname{Unif}\left(\mathbb{Z}_{q}^{m}\right)$

[^1]:    ${ }^{3}$ The decision problem is to distinguish such $t$ from $\operatorname{Unif}\left(\mathcal{R}_{q}^{m}\right)$

[^2]:    ${ }^{4}$ In power-of-two cyclotomic fields, $q$ must be prime such that $q=5 \bmod 8$.

