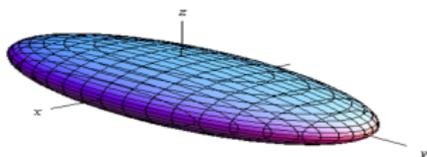


# On error distributions in ring-based LWE



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<sup>2</sup> Ghent University

<sup>3</sup> Open Security Research

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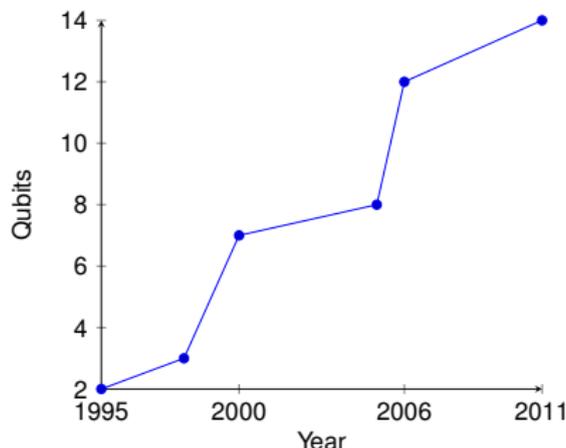


# Motivation for LWE

- 1981 A basic concept of a quantum computer by Feynman
- 1994 Shor's algorithm
  - ▶ Factorization and DLP are easy
  - ▶ **Broken**: RSA, Diffie-Hellman, ECDLP etc.

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- 1995 First quantum logic gate by Monroe, Meekhof, King, Itano and Wineland



# Motivation for LWE

## 2016 CNSA Suite and Quantum Computing FAQ by NSA

“Many experts predict a quantum computer capable of effectively breaking public key cryptography within a few decades, and therefore NSA believes it is important to address that concern.”

## NIST report on post-quantum crypto

“We must begin now to prepare our information security systems to be able to resist quantum computing.”

# Learning With Errors (LWE)

The LWE problem (Regev, '05): solve a linear system with noise

$$\begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1,n} \\ a_{21} & a_{22} & \dots & a_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{m,n} \end{pmatrix} \cdot \begin{pmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{pmatrix} + \begin{pmatrix} e_1 \\ e_2 \\ \vdots \\ e_m \end{pmatrix}$$

over a finite field  $\mathbb{F}_q$  for a secret  $(s_1, s_2, \dots, s_n) \in \mathbb{F}_q^n$  where

- ▶ a modulus  $q = \text{poly}(n)$
- ▶ the  $a_{ij} \in \mathbb{F}_q$  are chosen uniformly randomly,
- ▶ an adversary can ask for new equations ( $m > n$ ).

# Learning With Errors (LWE)

The LWE problem is easy when  $\forall e_i = 0$ .

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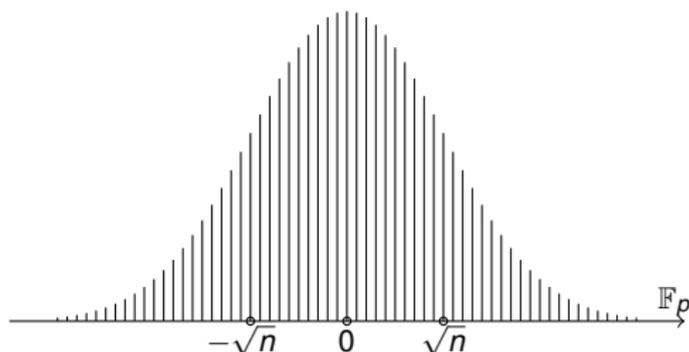
Gaussian elimination solves the problem.  
Otherwise, **LWE** might be hard.

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Gaussian elimination amplifies errors.

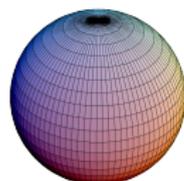
## Learning With Errors (LWE)

The errors  $e_j$  are sampled independently from a Gaussian with standard deviation  $\sigma > 2\sqrt{n}$ :



When viewed jointly, the error vector

$$\begin{pmatrix} e_1 \\ \vdots \\ e_m \end{pmatrix}$$

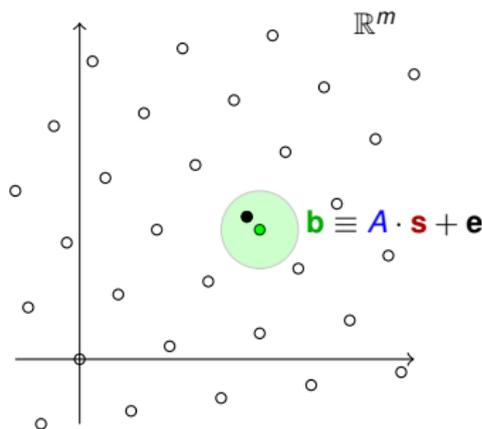


is sampled from a **spherical** Gaussian.

# Learning With Errors (LWE)

LWE is tightly related to classical lattice problems.

- ▶ Bounded Distance Decoding (BDD)



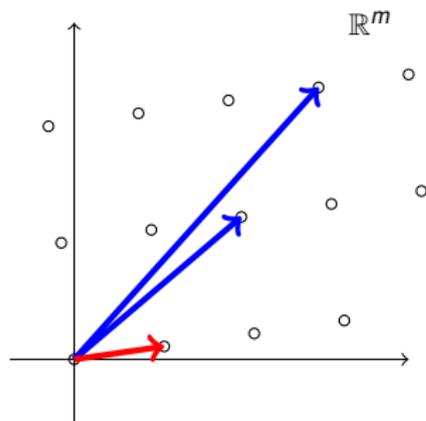
Given  $\mathbf{b}$ , find the closest point of the  $q$ -ary lattice

$$\{\mathbf{w} \in \mathbb{Z}^m \mid \exists \mathbf{s} \in \mathbb{Z}^n : \mathbf{w} \equiv \mathbf{A} \cdot \mathbf{s} \pmod{q}\}$$

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LWE is tightly related to classical lattice problems.

- ▶ Shortest Vector Problem (SVP)

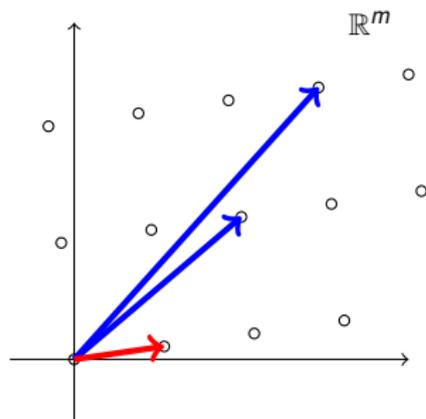


Given a **basis**, find **a shortest non-zero vector** of the lattice.

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Given a **basis**, find a **shortest non-zero vector** of the lattice.

- ▶ LWE is at least as hard as worst-case SVP-type problems (Regev'05, Peikert'09).
- ▶ Not known to be broken by quantum computers.

# Learning With Errors (LWE)

Known attacks for  $q = \text{poly}(n)$ :

	Time	Samples
Trial and error	$2^{O(n \log n)}$	$O(n)$
Blum, Kalai, Wasserman '03	$2^{O(n)}$	$2^{O(n)}$
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Idea: if all errors (almost) certainly lie in  $\{-T, \dots, T\}$ , then

$$\prod_{i=-T}^T (a_1 s_1 + a_2 s_2 + \dots + a_n s_n - b + i) = 0.$$

View as linear system of equations in  $\approx n^{2T}$  monomials.

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Application: public-key encryption of a bit (Regev'05).

- ▶ Private key:  $\mathbf{s} \in \mathbb{F}_q^n$ .
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$$\mathbf{c}^T := \mathbf{r}^T \cdot \mathbf{A} \quad \text{and} \quad \mathbf{d} := \begin{cases} \mathbf{r}^T \cdot \mathbf{b} & \text{if the bit is 0,} \\ \mathbf{r}^T \cdot \mathbf{b} + \lfloor q/2 \rfloor & \text{if the bit is 1.} \end{cases}$$

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- ▶ **Decryption** of pair  $\mathbf{c}^T, \mathbf{d}$ : compute

$$\mathbf{d} - \mathbf{c}^T \cdot \mathbf{s} = \mathbf{d} - \mathbf{r}^T \cdot \mathbf{A} \cdot \mathbf{s} = \mathbf{d} - \mathbf{r}^T \mathbf{b} - \mathbf{r}^T \mathbf{e} \approx \begin{cases} 0 & \text{if bit was 0,} \\ \lfloor q/2 \rfloor & \text{if bit was 1.} \end{cases}$$

↑  
small enough

# Learning With Errors (LWE)

- ▶ Features:
  - ▶ Hardness reduction from classical lattice problems
  - ▶ Linear operations
    - ▶ simple and efficient implementation
    - ▶ highly parallelizable
  - ▶ Source of exciting applications
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    - ▶ FHE, attribute-based encryption for arbitrary access policies, general-purpose code obfuscation
- ▶ Drawback: key size.
  - ▶ To hide the **secret** one needs an entire **linear system**:

$$\begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1,n} \\ a_{21} & a_{22} & \dots & a_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{m,n} \end{pmatrix} \cdot \begin{pmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{pmatrix} + \begin{pmatrix} e_1 \\ e_2 \\ \vdots \\ e_m \end{pmatrix}$$

$\uparrow$   $m \log p$                        $\uparrow$   $mn \log p$                        $\uparrow$   $n \log p$

# Ring-based LWE

- ▶ Identify vector space

$$\mathbb{F}_q^n \quad \text{with} \quad \mathcal{R}_q = \mathbb{Z}[x]/(q, f(x))$$

for some irreducible monic  $f(x) \in \mathbb{Z}[x]$  s.t.  $\deg f = n$ ,  
by viewing

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- ▶ Store  $\mathbf{a}(x)$  rather than  $A_{\mathbf{a}}$ : saves factor  $n$ .

# Ring-based LWE

Example:

- ▶ if  $f(x) = x^n + 1$ , then  $A_a$  is the anti-circulant matrix

$$\begin{pmatrix} a_1 & -a_n & \dots & -a_3 & -a_2 \\ a_2 & a_1 & \dots & -a_4 & -a_3 \\ a_3 & a_2 & \dots & -a_5 & -a_4 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ a_n & a_{n-1} & \dots & a_2 & a_1 \end{pmatrix}$$

of which it suffices to store the first column.

## Ring-based LWE

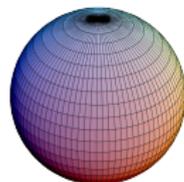
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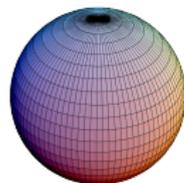
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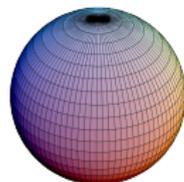
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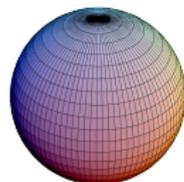
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- ▶ Not backed up by hardness statement.
- ▶ Sometimes called **Poly-LWE**.

# Ring-LWE

So what is Ring-LWE according to [LPR10]? Samples look like

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where

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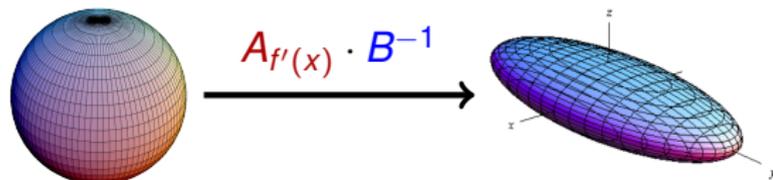
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Hardness reduction from ideal lattice problems.

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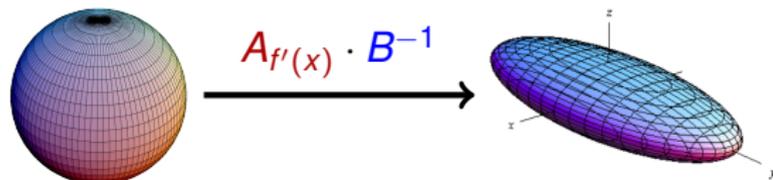
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- ▶ but also scales it!
  - ▶  $\det A_{f'(x)} = \Delta$  with

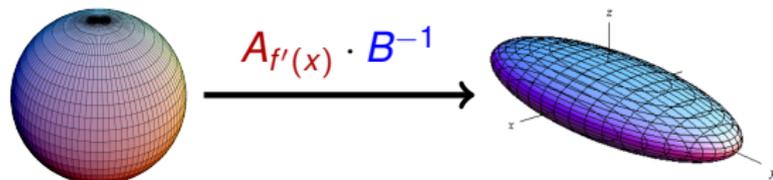
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So “on average”, each  $e_i$  is scaled up by  $\sqrt{\Delta}^{1/n} \dots$

- ▶ ... but remember: skewness.

# Scaled Canonical Gaussian ring-based LWE

$A_{f'(x)}$  is changed to a scalar  $\lambda$

$$\begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{pmatrix} = A_a \cdot \begin{pmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{pmatrix} + \lambda \cdot B^{-1} \cdot \begin{pmatrix} e_1 \\ e_2 \\ \vdots \\ e_n \end{pmatrix}.$$

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SCG-LWE = Ring-LWE for  $2^m$ -cyclotomic fields:

▶  $f'(x) = 2^{m-1} x^{2^{m-1}-1} = nx^{n-1}$ ,

▶  $\lambda = 2^{m-1} = n$ ,

▶ So  $A_{f'(x)} = A_{x^{n-1}} \cdot \lambda$ .

# Main result

For SCG ring-based LWE with parameters:

- ▶  $n = 2^\ell$  for some  $\ell \in \mathbb{N}$ ,
- ▶ a modulus  $q = \text{poly}(n)$ ,
- ▶ an error distribution with  $\sigma = \text{poly}(n)$ ,
- ▶ an underlying field  $K = \mathbb{Q}(\sqrt{p_1}, \sqrt{p_2}, \dots, \sqrt{p_\ell})$ ,
  - ▶ a square-free  $m = \prod p_i \geq (2\sigma\sqrt{n \log n})^{2/\varepsilon}$  for some  $\varepsilon > 0$ ,
  - ▶  $\forall i : p_i \equiv 1 \pmod{4}$ , so  $\Delta_K = m^{n/2}$ ,
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$\lambda' = \lambda/|\Delta_K|^{1/2n}$  appears in ELOS'15, CLS'15, CLS'16.

# Main result

Tensor structure:

- ▶  $K = K_1 \otimes_{\mathbb{Q}} K_2 \otimes_{\mathbb{Q}} \cdots \otimes_{\mathbb{Q}} K_\ell$ ,
  - ▶ where  $K_i = \mathbb{Q}(\sqrt{p_i})$
- ▶ The ring of integers  $R = R_1 \otimes_{\mathbb{Z}} R_2 \otimes_{\mathbb{Z}} \cdots \otimes_{\mathbb{Z}} R_\ell$ ,
  - ▶ where  $R_i = \mathbb{Z}[(1 + \sqrt{p_i})/2]$
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# Main result

Tensor structure:

- ▶  $K = K_1 \otimes_{\mathbb{Q}} K_2 \otimes_{\mathbb{Q}} \cdots \otimes_{\mathbb{Q}} K_\ell$ ,
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So  $\lambda \cdot B^{-1}$  is a Kronecker product of corresponding matrices in underlying quadratic fields  $K_i$

$$\begin{pmatrix} \frac{-1+\sqrt{p_i}}{2} & \frac{1+\sqrt{p_i}}{2} \\ 1 & -1 \end{pmatrix}$$

# Main result

Note

$$(0 \ 1) \cdot \begin{pmatrix} \frac{-1+\sqrt{p_i}}{2} & \frac{1+\sqrt{p_i}}{2} \\ 1 & -1 \end{pmatrix} = (1 \ -1)$$

and through the Kronecker product

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Applying to an error term of

$$\mathbf{b} = A_{\mathbf{a}} \cdot \mathbf{s} + \lambda' \cdot B^{-1} \cdot \mathbf{e}$$

we have

$$|\Delta_K|^{-\varepsilon/n} \cdot \mathbf{d} \cdot (\mathbf{e}_1 \ \mathbf{e}_2 \ \dots \ \mathbf{e}_n)^T = \omega.$$

## Main result

$\omega$  is distributed by Gaussian with the standard deviation

$$\frac{\sqrt{n} \cdot \sigma}{|\Delta_K|^{\varepsilon/n}} = \frac{\sqrt{n} \cdot \sigma}{\sqrt{m^\varepsilon}} \leq \frac{1}{2\sqrt{\log n}}.$$

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So a **SCG-LWE** sample

$$\begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{pmatrix} = A_{\mathbf{a}} \cdot \begin{pmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{pmatrix} + \lambda' \cdot B^{-1} \cdot \begin{pmatrix} e_1 \\ e_2 \\ \vdots \\ e_n \end{pmatrix}$$

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The attack works for the corresponding Ring-LWE problem with

$$\sigma' = \frac{\sigma}{|\Delta|^{\varepsilon/n}}.$$

# Conclusion

- ▶ No threat to the security proof of Ring-LWE.  
The standard deviation is far less than needed.

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Thank you for your attention!