

# Modern Cryptography

## Post-Quantum Cycles - Lattice-based Cryptography on LWE and R-LWE problems

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- GGH cryptosystem
- Learning With Error (LWE) problem
- Ring-Learning With Error (R-LWE) problem

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## Lattice problems

# Lattice problems

## Problems on lattices:

- Shortest Vector Problem (SVP)
- Closest Vector Problem (CVP)
- Shortest Independent Vectors Problem (SIVP)

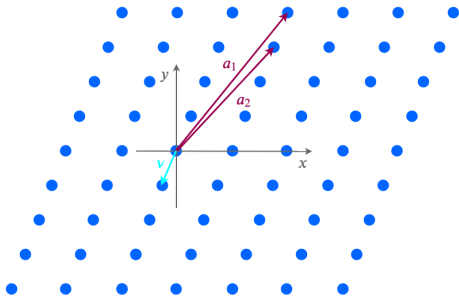
## Problems combined with lattices:

- Learning With Error (LWE) problem
- Ring-Learning With Error (R-LWE) problem
- Small Integer Solution (SIS) problem

# Lattice problems: SVP

## Note

Given  $\mathcal{B} = \{b_1, \dots, b_n\}$  linearly independent vectors, then  $\mathcal{L}(\mathcal{B})$  is the set of all  $t = \sum_i x_i b_i$  with  $x_i \in \mathbb{Z}$ .



## Shortest Vector Problem (SVP)

Given a lattice  $\mathcal{L}$ , find the **shortest non zero vector  $v$**  in  $\mathcal{L}$  with minimum  $\|v\|_2$ .

# SVP and solving algorithms

## Note

*There are no known polynomial-time algorithms for solving the SVP*

The algorithms that solves SVP

- deterministic (**exactly**), in  $n^{O(n)}$  [Kannan '83]
- randomized (**approximation**), in  $2^{O(n)}$  [AKS '01]

Faster algorithms are **very unlikely** because

- solve SVP **exactly** is NP-hard [Ajtai '98]

Nevertheless worst case hardness **is not enough for cryptography**.

## Theorem (Ajtai '98)

*Reduction of worst-case SVP to average-case SVP*

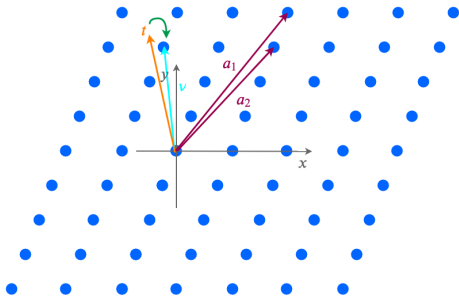
## Note (informally)

*Some worst-case SVP instances can be reduced to a SVP instance that can be generated with the solution (i.e., with the shortest vector).*

# Lattice problems: CVP

## Note

Given  $\mathcal{B} = \{b_1, \dots, b_n\}$  linearly independent vectors, then  $\mathcal{L}(\mathcal{B})$  is the set of all  $t = \sum_i x_i b_i$  with  $x_i \in \mathbb{Z}$ .



## Closest Vector Problem (CVP)

Given a lattice  $\mathcal{L}$  and a vector  $t \in \mathbb{R}^n$ , find  $v \in \mathcal{L}$  with minimum  $\|t - v\|_2$

# CVP and solving algorithms

The algorithms that solves SVP

- deterministic (**exactly**), in  $n^{O(n)}$  [Kannan '83]
- randomized (**approximation**), in  $2^{O(n)}$  [AKS '01]

... and for CVP

- deterministic (**exactly**), in  $n^{O(n)}$  [Micciancio, Voulgaris '10]

## Note

*SVP can be solved by a polynomial number of call to a CVP solver. Hence SVP hardness implies CVP hardness (not vice versa).*

However, it is proven that

- solve SVP **exactly** is NP-hard [Ajtai '98]
- solve CVP **exactly** is NP-hard [van Emde Boas '81]



# Beyond exact solutions

If we consider:

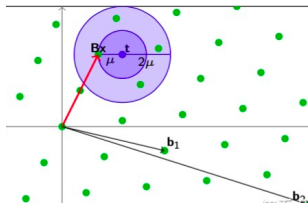
- $\text{dist}(t, \mathcal{L})$  be the minimum  $\|t - v\|_2$  for  $v \in \mathcal{L}$
- $\gamma > 1$  some **approximation factor**

The problem becomes:

**Approx-CVP <sub>$\gamma$</sub>**

Given  $t$  and  $\mathcal{L}$ , find  $v'$  such that

$$\|t - v'\|_2 \leq \gamma \text{dist}(t, \mathcal{L})$$



# Beyond exact solutions

If we consider:

- $\text{dist}(t, \mathcal{L})$  be the minimum  $\|t - v\|_2$  for  $v \in \mathcal{L}$
- $\gamma > 1$  some **approximation factor**

The problem becomes:

**Approx-CVP $_{\gamma}$**

Given  $t$  and  $\mathcal{L}$ , find  $v'$  such that

$$\|t - v'\|_2 \leq \gamma \text{dist}(t, \mathcal{L})$$

The algorithms that solves CVP

- deterministic (**exactly**), in  $n^{O(n)}$
- Approx-CVP $_{\gamma}$ , in  $2^{O(n)}$

[Micciancio, Voulgaris '10]

[LLL '82, Babai '86]

# Babai's Closest Vertex Algorithm

## Note

*If  $\mathcal{B}$  is an orthogonal basis of  $\mathcal{L}$ , approx-CVP can be solved in polynomial time.*

## Note

*Babai's algorithm also works with sufficiently orthogonal bases, i.e., it solves CVP and appr-CVP.*

## Theorem (Babai's Closest Vertex Algorithm)

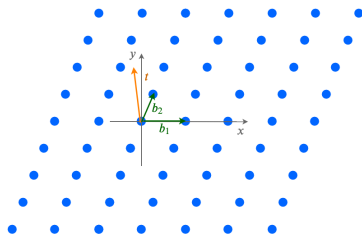
*Let  $\mathcal{L} \subset \mathbb{R}^n$  be a lattice with basis  $\mathbf{b}_1, \dots, \mathbf{b}_n$ , and let  $\mathbf{t} \in \mathbb{R}^n$  be an arbitrary vector. If the vectors in the basis are sufficiently orthogonal to one another, then the following algorithm solves CVP.*

*Write  $\mathbf{t} = t_1 \mathbf{b}_1 + t_2 \mathbf{b}_2 + \dots + t_n \mathbf{b}_n$  with  $t_1, \dots, t_n \in \mathbb{R}$ .  
Set  $a_i = \lceil t_i \rceil$  for  $i = 1, 2, \dots, n$ .  
Return the vector  $\mathbf{v} = a_1 \mathbf{b}_1 + a_2 \mathbf{b}_2 + \dots + a_n \mathbf{b}_n$ .*

# Babai's Closest Vertex Algorithm: Example

Write  $\mathbf{t} = t_1 \mathbf{b}_1 + t_2 \mathbf{b}_2 + \cdots + t_n \mathbf{b}_n$  with  $t_1, \dots, t_n \in \mathbb{R}$ .  
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Return the vector  $\mathbf{v} = a_1 \mathbf{b}_1 + a_2 \mathbf{b}_2 + \cdots + a_n \mathbf{b}_n$ .

Find the closest vector to  $\mathbf{t} = (-1.1, 6.2)$



Let  $\mathbf{b}_1 = (5, 0)$  and  $\mathbf{b}_2 = (1, 2)$  be the basis of  $\mathcal{L}$ .

$$t_1 \begin{pmatrix} 5 \\ 0 \end{pmatrix} + t_2 \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} -1.1 \\ 6.2 \end{pmatrix}$$

that is

$$t_2 = 3.1, \quad t_1 = \frac{-1.1 - t_2}{5} = -0.84$$

and

$$a_2 = \lceil 3.1 \rceil = 3, \quad a_1 = \lceil -0.84 \rceil = -1$$

We obtain:  $\mathbf{v} = -\mathbf{b}_1 + 3\mathbf{b}_2 = (-2, 6)$

# Babai's Closest Vertex Algorithm: Example

Write  $\mathbf{t} = t_1 \mathbf{b}_1 + t_2 \mathbf{b}_2 + \cdots + t_n \mathbf{b}_n$  with  $t_1, \dots, t_n \in \mathbb{R}$ .  
Set  $a_i = \lceil t_i \rceil$  for  $i = 1, 2, \dots, n$ .  
Return the vector  $\mathbf{v} = a_1 \mathbf{b}_1 + a_2 \mathbf{b}_2 + \cdots + a_n \mathbf{b}_n$ .

Find the closest vector to  $\mathbf{t} = (-1.1, 6.2)$

Let  $\mathbf{b}_1 = (9, 8)$  and  $\mathbf{b}_2 = (8, 6)$  be the basis of  $\mathcal{L}$ .

$$t_1 \begin{pmatrix} 9 \\ 8 \end{pmatrix} + t_2 \begin{pmatrix} 8 \\ 6 \end{pmatrix} = \begin{pmatrix} -1.1 \\ 6.2 \end{pmatrix}$$

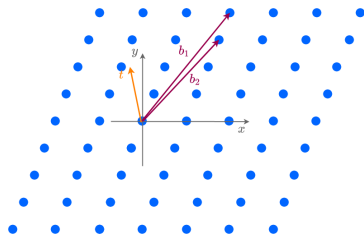
that is

$$t_2 = \frac{6.2 - 8t_1}{6}, \quad t_1 = \frac{3.3 + 4 * 6.2}{5} = 5.62$$

and

$$a_2 = \lceil -6.46 \rceil = -7, \quad a_1 = \lceil 5.62 \rceil = 6$$

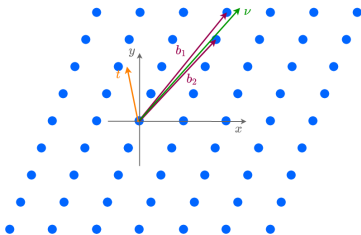
We obtain:  $\mathbf{v} = 6\mathbf{b}_1 - 7\mathbf{b}_2 = (6, 12)$



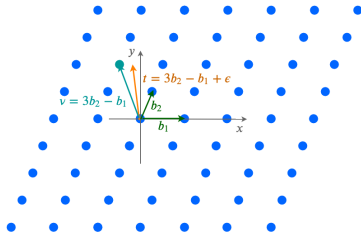
# Babai's Closest Vertex Algorithm: Example

Write  $\mathbf{t} = t_1 \mathbf{b}_1 + t_2 \mathbf{b}_2 + \cdots + t_n \mathbf{b}_n$  with  $t_1, \dots, t_n \in \mathbb{R}$ .  
Set  $a_i = \lceil t_i \rceil$  for  $i = 1, 2, \dots, n$ .  
Return the vector  $\mathbf{v} = a_1 \mathbf{b}_1 + a_2 \mathbf{b}_2 + \cdots + a_n \mathbf{b}_n$ .

Find the closest vector to  $\mathbf{t} = (-1.1, 6.2)$



If  $\mathbf{b}_1 = (9, 8)$  and  $\mathbf{b}_2 = (8, 6)$ , then  
 $\mathbf{v} = 6\mathbf{b}_1 - 7\mathbf{b}_2 = (6, 12)$



If  $\mathbf{b}_1 = (5, 0)$  and  $\mathbf{b}_2 = (1, 2)$ , then  
 $\mathbf{v} = -\mathbf{b}_1 + 3\mathbf{b}_2 = (-2, 6)$

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Goldreich, Goldwasser, Halevi (GGH) cryptosystem

# GGH public-key cryptosystem: example



## Key Generation

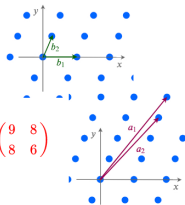
Choose a **good basis**:  $V = [v_1 | v_2] = \begin{pmatrix} 5 & 1 \\ 0 & 2 \end{pmatrix}$

Choose  $U$  such that  $\det(U) = \pm 1$

$$U = \begin{pmatrix} 1 & 1 \\ 4 & 3 \end{pmatrix}$$

Compute a **bad basis**:  $W = UV = [w_1 | w_2] = \begin{pmatrix} 9 & 8 \\ 8 & 6 \end{pmatrix}$

Publish the public key  $W$



## Encryption

Choose a message:  $m = (x_1, x_2) = (1, -1)$

Choose a small random vector:  $r = \begin{pmatrix} 1.1 \\ -0.5 \end{pmatrix}$

Compute:  $t = x_1 w_1 + x_2 w_2 + r = \begin{pmatrix} 2.1 \\ 1.5 \end{pmatrix}$

Send  $t$  to Alice

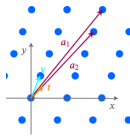
## Decryption

Use Babai's algorithm to solve CVP

$$\begin{aligned} t &= t_1 v_1 + t_2 v_2 \\ a_1 &= \lceil t_1 \rceil = \lceil 0.27 \rceil = 0 \\ a_2 &= \lceil t_2 \rceil = \lceil 0.75 \rceil = 1 \\ \nu &= a_1 v_1 + a_2 v_2 = \begin{pmatrix} 1 \\ 2 \end{pmatrix} \end{aligned}$$

$$W^{-1} = \begin{pmatrix} -3/5 & 4/5 \\ 12/15 & -9/10 \end{pmatrix}$$

$$m = \nu^T W^{-1} = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$





# GGH public-key cryptosystem

Alice	Bob
<b>Key Creation</b>	
Choose a <b>good basis</b> $v_1, \dots, v_n$ Choose an integer matrix $U$ satisfying $\det(U) = \pm 1$ . Compute a <b>bad basis</b> $w_1, \dots, w_n$ as rows of $W = UV$ . Publish the <b>public key</b> $w_1, \dots, w_n$ .	
<b>Encryption</b>	
	Choose small plaintext vector $m = (x_1, \dots, x_n)$ . Choose random small vector $r$ . Use Alice public key to compute $t = x_1 w_1 + \dots + x_n w_n + r$ . Send the ciphertext $t$ to Alice.
<b>Decryption</b>	
Use Babai's algorithm to compute the vector $v$ closest to $t$ . Compute $v^T W^{-1}$ to recover $m$ .	

# GGH public-key cryptosystem: a bit more

## Note

*GGH is a **probabilistic cryptosystem** since a single plaintext leads to many different ciphertexts due to the choice of the noise  $r$ .*

## Note (Attacks)

*No **asymptotically good attack** to GGH is known. Known attacks break the cryptosystem in practice for moderately large values of the security parameter.*

## Note (Security)

*Therefore, it is enough to **increase the security parameter**  $n$  (dimension of the lattice) to make the cryptosystem secure. However, this makes the cryptosystem impractical.*

so how much big  $n$ ?

Here is the [link](#) to the lattice challenge, i.e. SVP computation.

## Summary

GGH scheme is **secure** but **impractical**.

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## Learning With Errors (LWE) problem

# Problems combined with lattice

	Base	Ring	Module
Random error	Learning With Errors (LWE)	Ring-Learning With Errors(RLWE)	Module Learning With Errors (MLWE)
Rounding	Learning With Rounding (LWR)	Ring-Learning With Rounding(RLWR)	Module Learning With Rounding (MLWR)

Are all equivalent?

**In theory:** yes, but not strictly (one may need to change the parameters in order to have the same level of security).

**In practice:** At the moment there does not exist any attack which exploits the additional structures.

**Error or Rounding?:** in “error” we add a small noise, in “rounding” we round with a modulus.

# Solving a system of equations

Given a system of equations

$$\begin{cases} s_1 + 5s_2 + 3s_3 + 2s_4 \equiv 1 \pmod{7} \\ s_1 + 4s_2 + 2s_3 + 6s_4 \equiv 2 \pmod{7} \\ 2s_1 + s_2 + 3s_3 + s_4 \equiv 4 \pmod{7} \\ 3s_1 + 4s_2 + 4s_3 + 6s_4 \equiv 0 \pmod{7} \end{cases}$$

We want to find  $s = (s_1, \dots, s_4) \in \mathbb{Z}_7^4$

# Solving a system of equations

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We want to find  $\mathbf{s} = (s_1, \dots, s_4) \in \mathbb{Z}_7^4$

Note

*This problem is easy to solve via Gaussian Elimination (GE)*

$$\begin{pmatrix} 1 & 5 & 3 & 2 \\ 1 & 4 & 2 & 6 \\ 2 & 1 & 3 & 1 \\ 3 & 4 & 4 & 6 \end{pmatrix} \begin{pmatrix} s_1 \\ s_2 \\ s_3 \\ s_4 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \\ 4 \\ 0 \end{pmatrix} \xrightarrow{GE} \begin{pmatrix} 1 & 5 & 3 & 2 \\ 0 & 1 & 1 & -4 \\ 0 & 0 & 1 & 4 \\ 0 & 0 & 0 & 2 \end{pmatrix} \begin{pmatrix} s_1 \\ s_2 \\ s_3 \\ s_4 \end{pmatrix} = \begin{pmatrix} 1 \\ -1 \\ 0 \\ 0 \end{pmatrix}$$

Solution

In this case,  $\mathbf{s} = (-1, -1, 0, 0)$ .

# Solving a system of equations

Given a system of equations

$$\begin{cases} s_1 + 5s_2 + 3s_3 + 2s_4 \equiv 1 \pmod{7} \\ s_1 + 4s_2 + 2s_3 + 6s_4 \equiv 2 \pmod{7} \\ 2s_1 + s_2 + 3s_3 + s_4 \equiv 4 \pmod{7} \\ 3s_1 + 4s_2 + 4s_3 + 6s_4 \equiv 0 \pmod{7} \end{cases}$$

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Note

*This problem is easy to solve via Gaussian Elimination (GE)*

$$\begin{cases} s_1 + 5s_2 + 3s_3 + 2s_4 = 1 \pmod{7} \\ s_2 + s_3 + (-4)s_4 = -1 \pmod{7} \\ s_3 + 4s_4 = 0 \pmod{7} \\ 2s_4 = 0 \pmod{7} \end{cases}$$

Solution

In this case,  $s = (-1, -1, 0, 0)$ .

# Passing to LWE problem (Regev, 2005)

Given a system of equations

$$\begin{cases} s_1 + 5s_2 + 3s_3 + 2s_4 \equiv 1 \pmod{7} \\ s_1 + 4s_2 + 2s_3 + 6s_4 \equiv 2 \pmod{7} \\ 2s_1 + s_2 + 3s_3 + s_4 \equiv 4 \pmod{7} \\ 3s_1 + 4s_2 + 4s_3 + 6s_4 \equiv 0 \pmod{7} \end{cases}$$

We want to find  $\mathbf{s} = (s_1, \dots, s_4) \in \mathbb{Z}_7^4$

Note

If we add a small error  $\mathbf{e} \in \{-1, 0, 1\}$ ,

$$\begin{cases} s_1 + 5s_2 + 3s_3 + 2s_4 \approx 2 (= 1 + 1) \pmod{7} \\ s_1 + 4s_2 + 2s_3 + 6s_4 \approx 2 \pmod{7} \\ 2s_1 + s_2 + 3s_3 + s_4 \approx 3 \pmod{7} \\ 3s_1 + 4s_2 + 4s_3 + 6s_4 \approx -1 \pmod{7} \end{cases}$$

the Gaussian elimination does not work anymore, i.e., we cannot find  $\mathbf{s}$ .

Learning with errors is hard.



# LWE problem: ingredients

LWE ingredients:

- integers  $n, q = \text{poly}(n), m$ ,
- an **error probability distribution**  $\chi$  over  $\mathbb{Z}_q$ ,
- $m$  random vectors  $a_i \in \mathbb{Z}_q^n$
- a **secret**  $s \in \mathbb{Z}_q^n$  and  $b_1, \dots, b_m \in \mathbb{Z}_q$  such that

$$\begin{array}{c} \left( \begin{array}{cccc} 14 & 15 & 5 & 2 \\ 13 & 14 & 14 & 6 \\ 6 & 10 & 13 & 1 \\ & \vdots & & \\ 6 & 7 & 16 & 2 \end{array} \right) \\ \mathbf{A} \end{array} + \begin{array}{c} \left( \begin{array}{c} s_1 \\ s_2 \\ s_3 \\ s_4 \end{array} \right) \\ \mathbf{s} \end{array} = \begin{array}{c} \left( \begin{array}{c} 1 \\ 0 \\ -1 \\ \vdots \\ 1 \end{array} \right) \\ \mathbf{e} \end{array} = \begin{array}{c} \left( \begin{array}{c} 8 \\ 16 \\ 3 \\ \vdots \\ 3 \end{array} \right) \\ \mathbf{b} \end{array}$$

# Is finding $s$ even solvable?

## Note

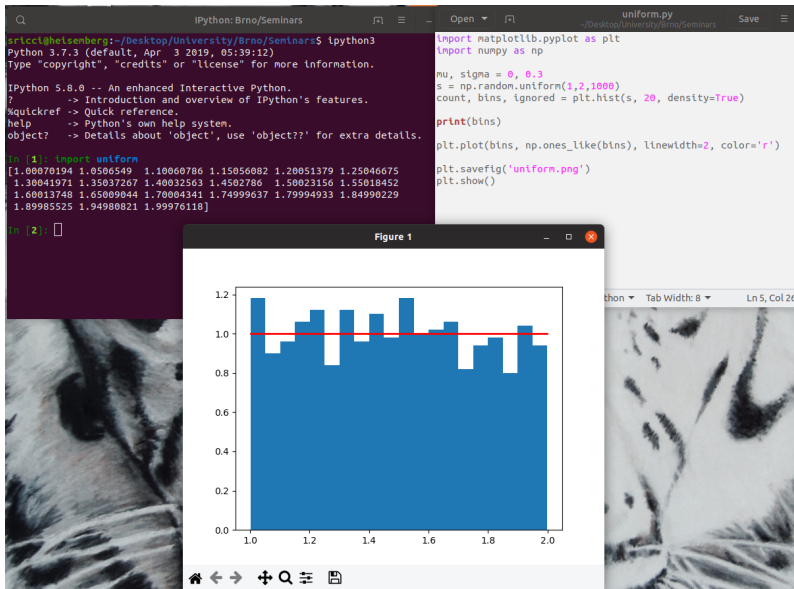
*Short answer: it depends on the choice of the parameters.*

- if  $n = 1$  and  $e \in \{-1, 0, 1\}$ , i.e.,  $\chi$  is an uniform distribution over  $\mathbb{Z}_2$ ,  
LWE can be solved (not easily) by linearization.
- if  $\chi$  is uniform over  $\mathbb{Z}_q$ ,  
there is no information on the related LWE problem solution.

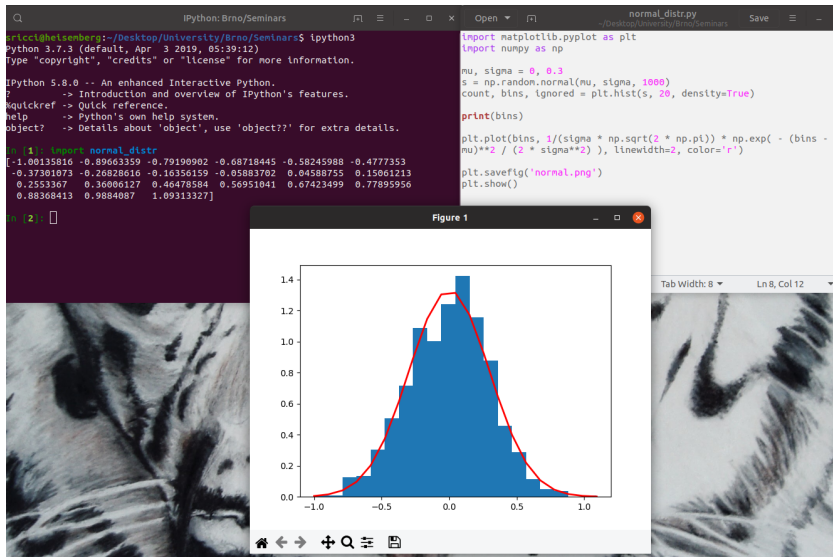
## Note

*It is important to choose (smartly) the error, i.e., the probability distribution  $\chi$  that creates the error.*

# Uniform distribution: example



# Normal distribution: example



# LWE problem: formulation

## Problem (Computational-LWE problem)

Given  $n, m, q \in \mathbb{Z}$  and  $\chi$  a distribution in  $\mathbb{Z}_q$  (typically “rounded” normal distribution).

**Input:** a pair  $(A, As + e)$ , where

- $A \in_{UR} \mathbb{Z}_q^{m \times n}$ .
- $e$  chosen in  $\mathbb{Z}_q^m$  according to  $\chi^m$ .

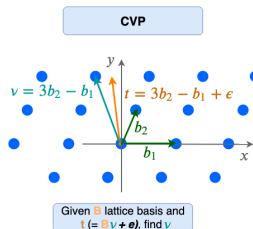
**Goal:** For a vector  $s \in_{UR} \mathbb{Z}_q^n$ , given arbitrarily many samples  $(A, As + e)$ , compute  $s$ .

Given  $(A, As + e)$

$$\begin{pmatrix} 14 & 15 & 5 & 2 \\ 13 & 14 & 14 & 6 \\ 6 & 10 & 13 & 1 \\ \vdots & & & \\ 6 & 7 & 16 & 2 \end{pmatrix} \begin{pmatrix} s_1 \\ s_2 \\ s_3 \\ s_4 \end{pmatrix} = \begin{pmatrix} 8 \\ 16 \\ 3 \\ \vdots \\ 3 \end{pmatrix}$$

$A \qquad \qquad s \qquad \qquad As + e$

our goal is to recover  $s$



# LWE problem: formulation

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## Note (Hardness)

Breaking the problem (or finding an efficient algorithm for LWE) implies having an efficient quantum algorithm for approximating SVP.

## Theorem (Regev)

The LWE problems are as hard as worst-case assumptions in general lattices when  $\chi$  is a discrete Gaussian with standard deviation  $\sigma = \alpha q$  for some fixed real number  $0 < \alpha < 1$ .

# LWE problem in matrix form

$$\begin{aligned}14s_1 + 15s_2 + 5s_3 + 2s_4 &\approx 8 \bmod 17 \\13s_1 + 14s_2 + 14s_3 + 6s_4 &\approx 16 \bmod 17 \\6s_1 + 10s_2 + 13s_3 + 1s_4 &\approx 3 \bmod 17 \\10s_1 + 4s_2 + 12s_3 + 16s_4 &\approx 12 \bmod 17 \\&\vdots \\6s_1 + 7s_2 + 16s_3 + 2s_4 &\approx 3 \bmod 17\end{aligned}$$

In this case:  $\mathbf{s} \in \mathbb{Z}_{17}^4$

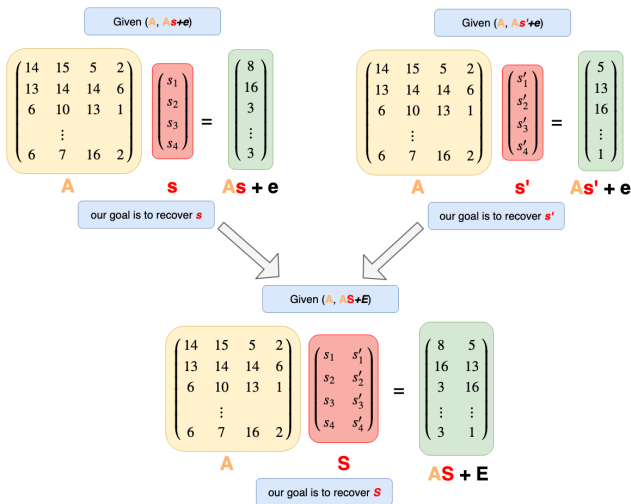
and the equation is correct up to some small additive error (say,  $\pm 1$ )

our goal is to recover  $\mathbf{s}$

$$\begin{pmatrix} 14 & 15 & 5 & 2 \\ 13 & 14 & 14 & 6 \\ 6 & 10 & 13 & 1 \\ \vdots & & & \\ 6 & 7 & 16 & 2 \end{pmatrix} \begin{pmatrix} s_1 \\ s_2 \\ s_3 \\ s_4 \end{pmatrix} + \begin{pmatrix} 1 \\ 0 \\ -1 \\ \vdots \\ 1 \end{pmatrix} = \begin{pmatrix} 8 \\ 16 \\ 3 \\ \vdots \\ 3 \end{pmatrix}$$

$\mathbf{A} \qquad \mathbf{s} \qquad \mathbf{e} \qquad \mathbf{As} + \mathbf{e}$

# LWE problem: all matrices!





# When the error is “enough small”?

## Simplified problem

We want to encrypt one-bit message  $m \in \{0, 1\}$ .

If  $p \in \mathbb{Z}_{17}$  is our noise, then

$$c = p + m$$

For which value of  $p$  are we able to decrypt?

# When the error is “enough small”?

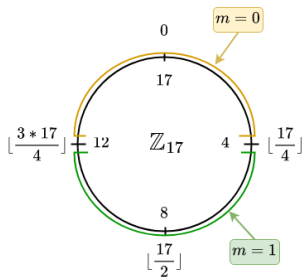
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For which value of  $p$  are we able to decrypt?



Therefore, the encryption becomes

$$c = p + m \lfloor \frac{q}{2} \rfloor$$

and the decryption is

$$m = \begin{cases} 1 & \text{if } p \in [5, 11] \\ 0 & \text{if } p \in [-4, 3] \end{cases}$$

# (Regev) one-bit encryption scheme



Alice



Bob

## Public Parameter

Security parameter:  $n=3$

Modulus :  $q=17$

Number of eqs: 4

## Key Generation

Private key:  $s = \begin{pmatrix} 2 \\ 3 \\ -2 \end{pmatrix} \in_{UR} \mathbb{Z}_{17}^3$   $A = \begin{pmatrix} 5 & 3 & -1 \\ 6 & 1 & -13 \\ 1 & -11 & 1 \\ -7 & 4 & 0 \end{pmatrix} \in_{UR} \mathbb{Z}_{17}^{4 \times 3}$

$$e = \begin{pmatrix} 1 \\ -1 \\ 0 \\ 1 \end{pmatrix} \in_{\psi_0} \mathbb{Z}_{17}^4 \quad p = As + e = \begin{pmatrix} 21 \\ -11 \\ -33 \\ -2 \end{pmatrix} + e \bmod 17 = \begin{pmatrix} 5 \\ 1 \\ -1 \end{pmatrix}$$

Public key:  $(A, p)$

## Encryption

Message:  $b = 1$  is a bit  $\{0, 1\}$

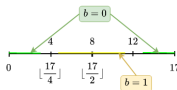
$$\begin{aligned} Enc_{A,p}(b) &= (a', p') = \left( \sum a_i, \sum p_i + b \lfloor \frac{q}{2} \rfloor \right) \\ &= (a_1 + a_4 = -2, 7, -1), p_1 + p_4 = 12) \end{aligned}$$

Ciphertext:  $(a', p')$

## Decryption

$$e' = p' - a' * s^T = 12 - 2 = 10$$

$$Dec_s(a', p') = \begin{cases} 0 & \text{if } e' \sim 0 \\ 1 & \text{if } e' \sim \frac{17}{2} \end{cases} = 1$$



# (Generalized) Regev cryptosystem

## Note

*The one-bit scheme can be generalized to work with  $m \in \mathbb{Z}_q$ .*

## Efficiency:

- only multiplications and additions modulus  $q$ .
- parallelization.
- **Public/Private key** size:  $(n/\log q, m(n + l) \log q)$ .
- Operation encryption/decryption per bit:  $(O(m(1 + n/l)), O(n))$  (ignoring logarithmic factors).

## Security:

- distinguishing between public keys  $(A, P)$  as generated by the cryptosystem and pairs  $(A, P)$  chosen uniformly at random implies a [solution to the LWE problem](#).

## Decryption errors:

- the error has to be “[small enough](#)”.
- Using an appropriate set of parameters and an error correcting code to encode  $m$  help.

## Note

*It is [necessary to choose the parameters](#) so that the [LWE problem is hard](#).*

# LWE applications to cryptography ...

... include but are not limited to

- secret key encryption
- public key encryption
- key exchange
- digital signature
- (fully) homomorphic encryption
- identity-based encryption
- zero-knowledge proofs

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## Ring-Learning With Errors (R-LWE) problem

# R-LWE problem: why?

## Note

*Cryptographic schemes based **LWE problems** (and SIS) tend to require rather **large key sizes** ( $\sim n^2$ ).*

From a practical point of view, it will be good to **reduce** the key size to **almost linear size**.

# R-LWE problem: why?

## Note

Cryptographic schemes based *LWE problems* (and SIS) tend to require rather *large key sizes* ( $\sim n^2$ ).

From a practical point of view, it will be good to *reduce* the key size to *almost linear size*.

- Need: *more compactness*.
- Idea: replace a random matrix with a *structured* “circulant” matrix with a random column

$$\begin{pmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{pmatrix} \Rightarrow \begin{pmatrix} a_1 & a_4 & a_3 & a_2 \\ a_2 & a_1 & a_4 & a_3 \\ a_3 & a_2 & a_1 & a_4 \\ a_4 & a_3 & a_2 & a_1 \end{pmatrix}$$

- We need to memorize only one column.

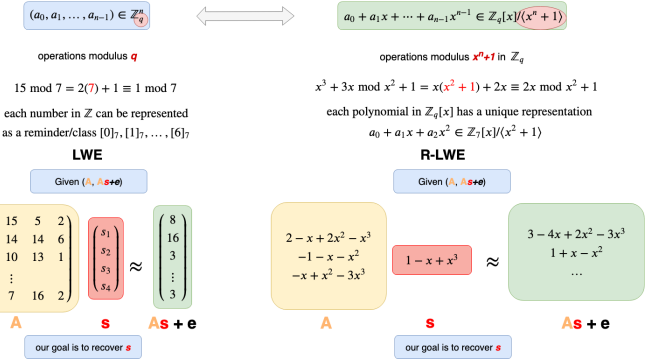


# Ideal lattice: on the other side

## Note

It is possible to achieve this goal assuming that there is *some structure* in the LWE samples.

The structure is the *ideal lattice*, i.e. the group  $\mathbb{Z}_q^n$  is replaced by the ring  $\mathbb{Z}_q[x]/\langle x^n + 1 \rangle$ .



# Summary

## Problems on lattice:

- CVP and SVP problems are NP-hard, i.e., secure against quantum attacks.
- Cryptographic protocols are based on hard problems that are easy to solve if you have a trapdoor (e.g., a good basis of a lattice)
- CVP and appr-CVP can be solved by Babai's algorithm with a lattice good basis
- GGH cryptosystem is based on the CVP problem and uses Babai's algorithm to decrypt the message.

## LWE and R-LWE problems:

- solving a system of equations is easy (Gaussian elimination).
- if a small error is added, then Gaussian elimination does not work anymore
- LWE problem is NP-Hard with the right choice of parameters.
- LWE has too big key size (matrix  $A$  and secret  $s$ )
- R-LWE adds a structure to  $A$  that allows memorizing only one column of  $A$ , i.e. smaller keys.
- R-LWE problem is based on polynomials.

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# Thank you for attention!

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