

# An LLL algorithm for module lattices

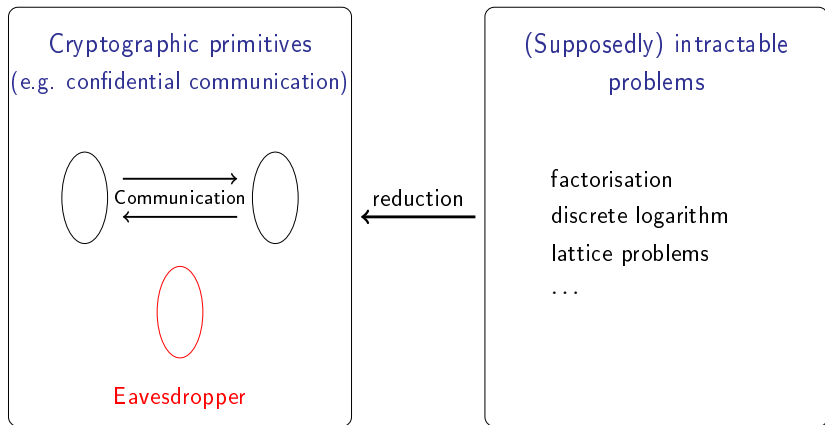
Changmin Lee<sup>1</sup>, **Alice Pellet-Mary**<sup>2</sup>, Damien Stehlé<sup>1</sup>  
and Alexandre Wallet<sup>3</sup>

<sup>1</sup> ENS de Lyon, <sup>2</sup> KU Leuven, <sup>3</sup> NTT Tokyo

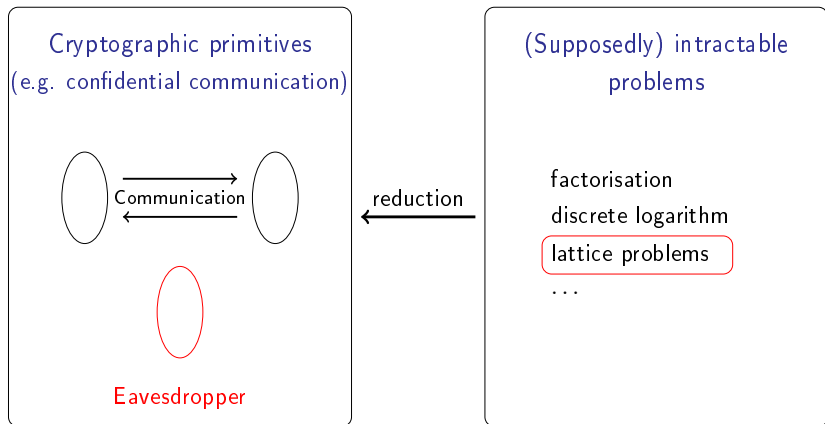
Institut Fourier,  
January 16, 2020

<https://eprint.iacr.org/2019/1035>

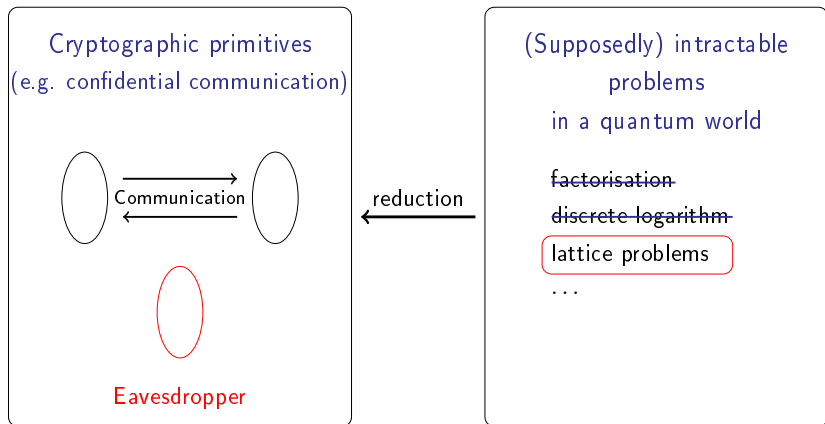
# Cryptography and hard problems



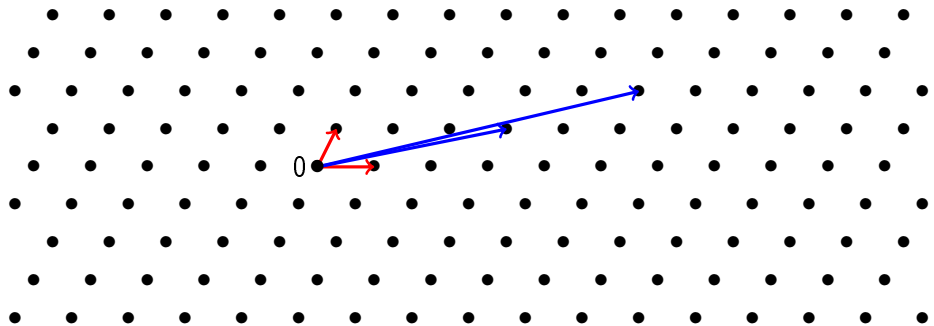
# Cryptography and hard problems



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# Lattices

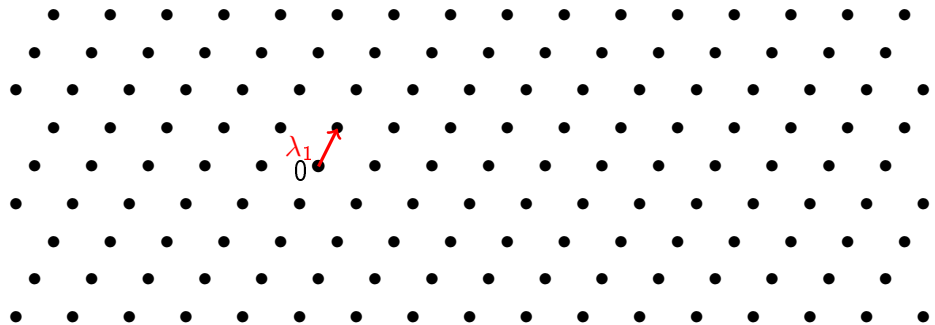


## Lattice

A (full-rank) lattice  $L$  is a subset of  $\mathbb{R}^n$  of the form  $L = \{Bx \mid x \in \mathbb{Z}^n\}$ , with  $B \in \mathbb{R}^{n \times n}$  invertible.  $B$  is a **basis** of  $L$ .

$\begin{pmatrix} 3 & 1 \\ 0 & 2 \end{pmatrix}$  and  $\begin{pmatrix} 17 & 10 \\ 4 & 2 \end{pmatrix}$  are two bases of the above lattice.

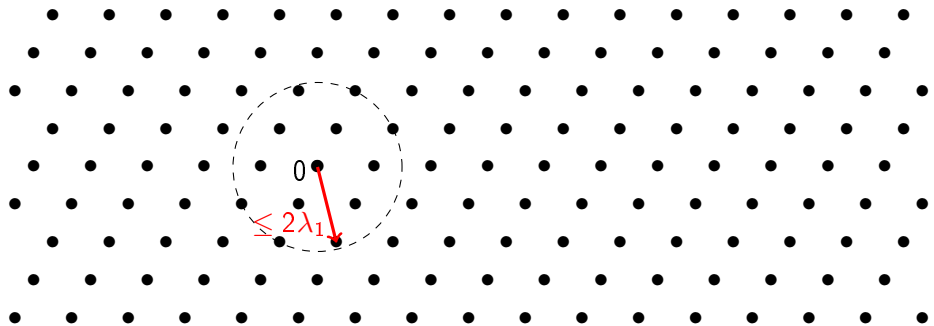
# Lattice problems



## Shortest Vector Problem (SVP)

Find a shortest (in Euclidean norm) non-zero vector.  
Its Euclidean norm is denoted  $\lambda_1$ .

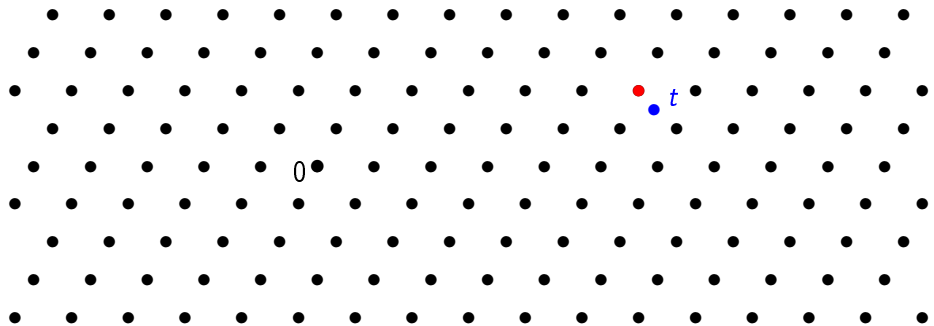
# Lattice problems



## Approximate Shortest Vector Problem (approx-SVP)

Find a short (in Euclidean norm) non-zero vector.  
(e.g. of norm  $\leq 2\lambda_1$ ).

# Lattice problems

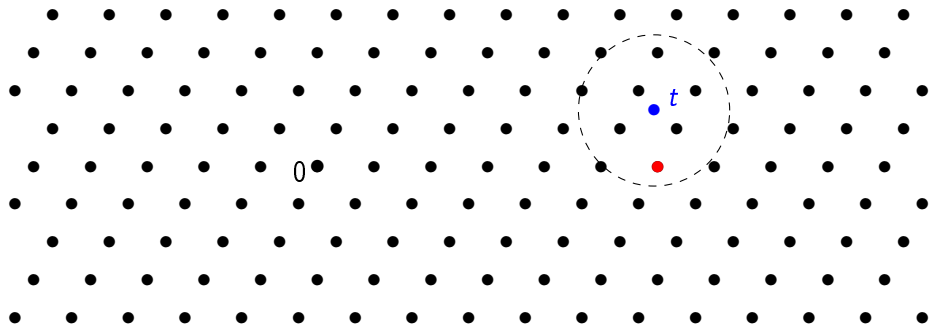


## Closest Vector Problem (CVP)

Given a target point  $t$ , find a point of the lattice closest to  $t$ .



## Lattice problems

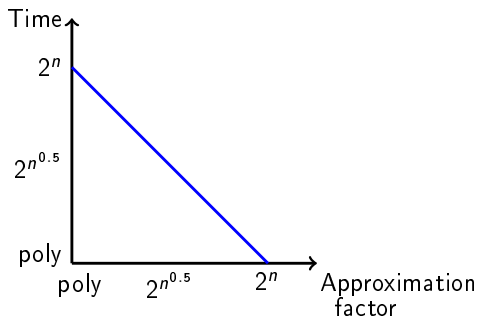


### Approximate Closest Vector Problem (approx-CVP)

Given a target point  $t$ , find a point of the lattice close to  $t$ .

# Hardness of lattice problems

Best Time/Approximation trade-off for SVP and CVP (even quantumly):  
BKZ algorithm [Sch87,SE94]

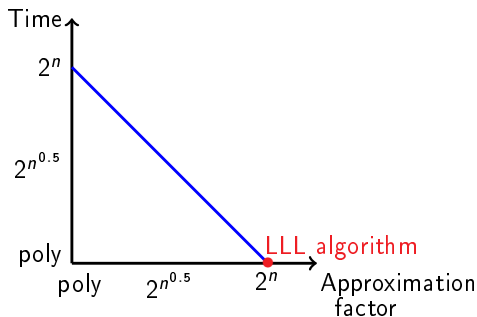


[Sch87] C.-P. Schnorr. A hierarchy of polynomial time lattice basis reduction algorithms. TCS.

[SE94] C.-P. Schnorr and M. Euchner. Lattice basis reduction: improved practical algorithms and solving subset sum problems. Mathematical programming.

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[LLL82] A. K. Lenstra, H. W. Lenstra, L. Lovász. Factoring polynomials with rational coefficients. *Mathematische Annalen*.

# Structured lattices

## Motivation

Schemes using lattices are usually not very efficient

(storage:  $n^2$ , matrix-vector mult:  $n^2$ )

⇒ improve efficiency using **structured lattices**

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**Example:** NIST post-quantum standardization process

- 26 candidates (2nd round)
- 12 lattice-based
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	Frodo (lvl 1) (unstructured lattices)	Kyber (lvl 1) (structured lattices)
secret key size (in Bytes)	19 888	1 632
public key size (in Bytes)	9 616	800

## Ideal lattices

$$M_{\mathbf{a}} = \begin{pmatrix} a_0 & a_{n-1} & \cdots & a_1 \\ a_1 & a_0 & \cdots & a_2 \\ \vdots & \ddots & \ddots & \vdots \\ a_{n-1} & a_{n-2} & \cdots & a_0 \end{pmatrix}$$

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 $a_0 + a_1X + \cdots + a_{n-1}X^{n-1}$   
mod  $X^n - 1$



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$$R = \mathbb{Z}[X]/(X^n - 1)$$

$$\sigma : \quad R \rightarrow \mathbb{R}^n$$

$$\sum_{i=0}^{n-1} a_i X^i \mapsto (a_0, \dots, a_{n-1})$$

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$$\mathcal{L}(M_{\mathbf{a}}) = \sigma(\langle \mathbf{a} \rangle)$$

(principal) ideal lattice

## Ideal lattices (2)

### Notations

- $P$  irreducible monic polynomial of degree  $d$
- $K = \mathbb{Q}[X]/P$
- $R$  ring of integers of  $K$

An ideal lattice is  $\sigma(I)$  for  $I$  an ideal of  $R$

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### For simplicity in this talk

- only principal ideals  $I = \langle a \rangle$
- $P = X^d + 1$  ( $d = 2^k$ )  
or  $P = X^d - X - 1$  ( $d$  prime)

# Module lattices

## Definition

A (full rank) free module  $\mathcal{M} \subset K^k$  is

$$\mathcal{M} = \left\{ \sum_i x_i \vec{b}_i : x_i \in R \right\},$$

where the  $\vec{b}_i \in K^k$  are linearly independent vectors and  $1 \leq i \leq k$ .  
 $k$  is the **rank** of  $\mathcal{M}$ .

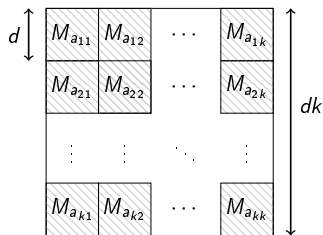
A (free) module lattice is  $\sigma(\mathcal{M}) \subset \mathbb{R}^{kd}$

where  $\sigma((x_1, \dots, x_k)) = \sigma(x_1) \parallel \dots \parallel \sigma(x_k)$

# Duality of module lattices

## Recall

- $M_a$  = matrix of multiplication by  $a$
- $\mathcal{M} = \{\sum_i x_i \vec{b}_i : x_i \in R\}$ , with  $\vec{b}_i = (a_{1i}, \dots, a_{ki})$ .

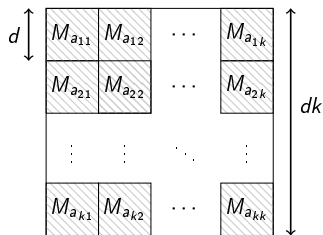


Basis of  $\mathcal{L}(\mathcal{M})$  over  $\mathbb{Z}$   
( $\dim n = dk$ )

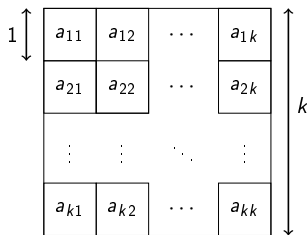
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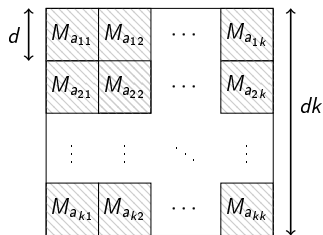


Basis of  $\mathcal{M}$  over  $R$   
( $\dim k$ )

# Duality of module lattices

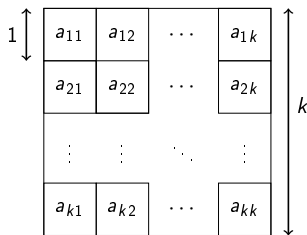
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Basis of  $\mathcal{L}(\mathcal{M})$  over  $\mathbb{Z}$   
(dim  $n = dk$ )

Typically  $500 \leq dk \leq 1000$



Basis of  $\mathcal{M}$  over  $R$   
(dim  $k$ )

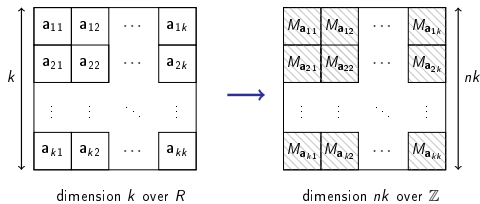
Typically  $k \leq 10$



# Canonical embedding

## Reminder

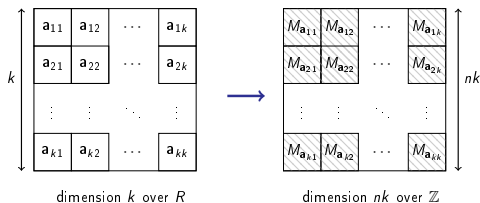
$$K = \mathbb{Q}[X]/P(X)$$



# Canonical embedding

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## Coefficient embedding

$$\sigma : K \rightarrow \mathbb{R}^n$$

$$\mathbf{a} = a_0 + a_1X + \cdots + a_{n-1}X^{n-1} \mapsto (a_0, a_1, \dots, a_{n-1})^T$$

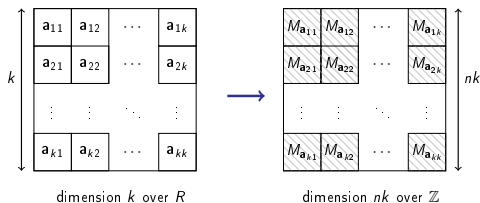
$$\mathbf{a} \rightarrow M_{\mathbf{a}} = \begin{pmatrix} \left| \begin{array}{c} \sigma(\mathbf{a}) \\ \hline \end{array} \right| & \left| \begin{array}{c} \sigma(X\mathbf{a}) \\ \hline \end{array} \right| & \cdots & \left| \begin{array}{c} \sigma(X^{n-1}\mathbf{a}) \\ \hline \end{array} \right| \end{pmatrix}$$

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## Reminder

$$K = \mathbb{Q}[X]/P(X)$$

$\alpha_1, \dots, \alpha_n$  roots of  $P$



## Canonical embedding

$$\sigma : K \rightarrow \mathbb{C}^n$$

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## Objective

*Is SVP still hard when restricted to module lattices?*

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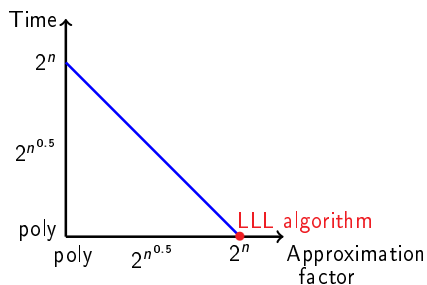
*Is SVP still hard when restricted to module lattices?*

## Module lattices

- large dimension over  $\mathbb{Z}$
- small dimension over  $R$

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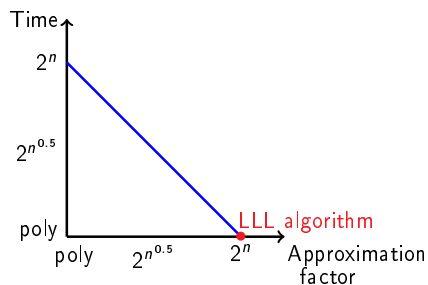
Lattice reduction over  $\mathbb{Z}$

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Lattice reduction over  $\mathbb{Z}$

## Module lattices

- large dimension over  $\mathbb{Z}$
- small dimension over  $R$

Can we extend the LLL algorithm to lattices over  $R$ ?

## Previous works and result

- [Nap96] LLL for some specific number fields  
no bound on quality / run-time

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[Nap96] H. Napias. A generalization of the LLL-algorithm over Euclidean rings or orders. *Journal de théorie des nombres de Bordeaux*.



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- [FP96] LLL for any number fields  
no bound on quality / run-time  
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[FP96] C. Fieker, M. E. Pohst. Lattices over number fields. ANTS.

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- [Cam16] LLL for Euclidean (for  $\|\cdot\|_2$  norm) imaginary quadratic fields  
bound on run-time and on quality

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[Cam17] T. Camus. Méthodes algorithmiques pour les réseaux algébriques. Thèse de doctorat.

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- [KL17] LLL for norm-Euclidean fields  
bound on run-time but not on quality  
bound on quality for biquadratic fields

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[KL17] T. Kim, C. Lee. Lattice reductions over euclidean rings with applications to cryptanalysis. IMACC.

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- [LPSW19] LLL for any number field  
bound on quality and run-time if oracle solving CVP in a fixed lattice (depending on  $R$ )

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[LPSW19] C. Lee, A. Pellet-Mary, D. Stehlé, A. Wallet. An LLL algorithm for module lattices. To appear at Asiacrypt 2019.

# Outline of the talk

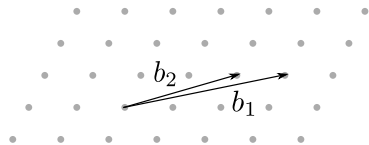
- 1 LLL over  $K^2$ 
  - QR factorization
  - Euclidean division
  
- 2 LLL over  $K^k$

# Outline of the talk

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  - Euclidean division

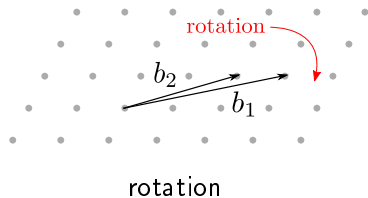
- 2 LLL over  $K^k$

## Lagrange-Gauss algorithm (over $\mathbb{Z}$ )



$$M = \begin{pmatrix} 10 & 7 \\ 2 & 2 \end{pmatrix}$$

# Lagrange-Gauss algorithm (over $\mathbb{Z}$ )

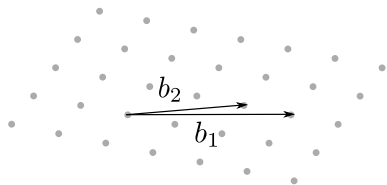


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Compute QR factorization

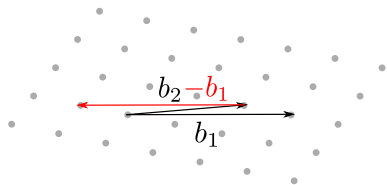


## Lagrange-Gauss algorithm (over $\mathbb{Z}$ )



$$M = \begin{pmatrix} 10.2 & 7.3 \\ 0 & 0.6 \end{pmatrix}$$

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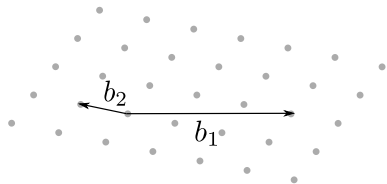


reduce  $b_2$  with  $b_1$

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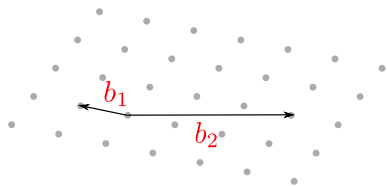
“Euclidean division” (over  $\mathbb{R}$ )  
of 7.3 by 10.2

## Lagrange-Gauss algorithm (over $\mathbb{Z}$ )



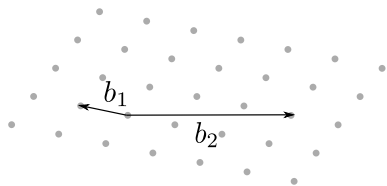
$$M = \begin{pmatrix} 10.2 & -2.9 \\ 0 & 0.6 \end{pmatrix}$$

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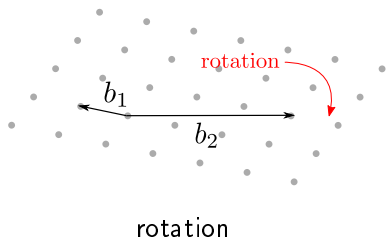
## Lagrange-Gauss algorithm (over $\mathbb{Z}$ )



start again

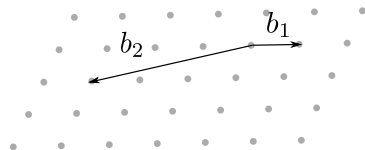
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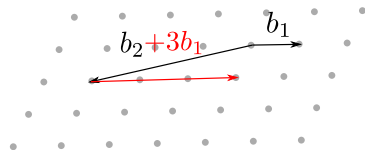
## Lagrange-Gauss algorithm (over $\mathbb{Z}$ )



rotation

$$M = \begin{pmatrix} 3 & -10 \\ 0 & -2 \end{pmatrix}$$

## Lagrange-Gauss algorithm (over $\mathbb{Z}$ )



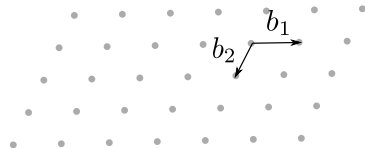
reduce  $b_2$  with  $b_1$

$$M = \begin{pmatrix} 3 & -10 \\ 0 & -2 \end{pmatrix}$$

“Euclidean division” (over  $\mathbb{R}$ )  
of  $-10$  by  $3$

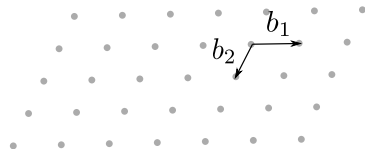


## Lagrange-Gauss algorithm (over $\mathbb{Z}$ )



$$M = \begin{pmatrix} 3 & -1 \\ 0 & -2 \end{pmatrix}$$

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$$M = \begin{pmatrix} 3 & -1 \\ 0 & -2 \end{pmatrix}$$

For the Lagrange-Gauss algorithm over  $R$ , we need

- Rotation (i.e., QR factorization)
- Euclidean division

## QR factorization over $R$

## Inner product over $R$

For  $\vec{a} = (a_1, \dots, a_k) \in K^k$  and  $\vec{b} = (b_1, \dots, b_k) \in K^k$ ,

$$\langle \vec{a}, \vec{b} \rangle_K = \sum_i a_i \bar{b}_i \in K \quad (\text{or } K_{\mathbb{R}} = K \otimes \mathbb{R})$$

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### Properties

- $\text{Tr}(\|\vec{a}\|_K^2) = \|\sigma(\vec{a})\|_2^2$

$$\text{Tr}(x) = \sum_{i=1}^d \sigma(x)_i$$

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For  $\vec{a} = (a_1, \dots, a_k) \in K^k$  and  $\vec{b} = (b_1, \dots, b_k) \in K^k$ ,

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$$\|\vec{a}\|_K := \sqrt{\langle \vec{a}, \vec{a} \rangle_K} \in K \quad (\text{or } K_{\mathbb{R}})$$

### Properties

- $\text{Tr}(\|\vec{a}\|_K^2) = \|\sigma(\vec{a})\|_2^2$
- $\mathcal{N}(\|\vec{a}\|_K^2) = \Delta_K^{-1} \cdot \det(\mathcal{L}(\vec{a}))^2$

$$\mathcal{N}(x) = \prod_{i=1}^d \sigma(x)_i$$

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Let  $B = (b_1, \dots, b_k) \in K^k$ , define

$$b_i^* = b_i - \sum_{j < i} \mu_{ij} b_j^*, \quad \text{with } \mu_{ij} = \frac{\langle b_i, b_j^* \rangle_K}{\langle b_j^*, b_j^* \rangle_K}$$



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### Properties

- $R$  is triangular
- $\overline{Q}^T Q = I_k$
- $\langle R\vec{u}, R\vec{v} \rangle = \langle B\vec{u}, B\vec{v} \rangle$

## Euclidean division over $R$

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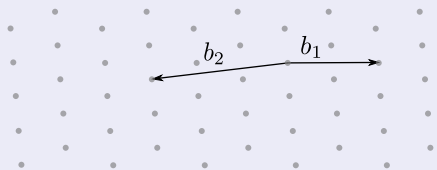
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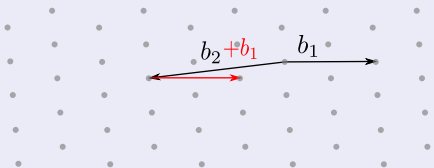
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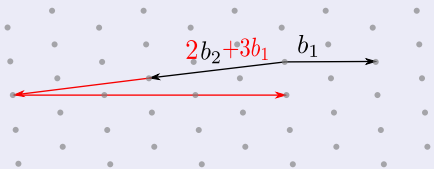
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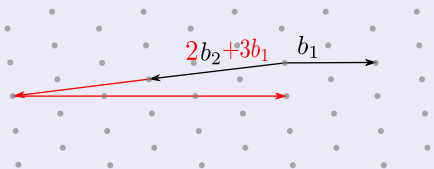
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$\Rightarrow$  sufficient for Gauss' algo



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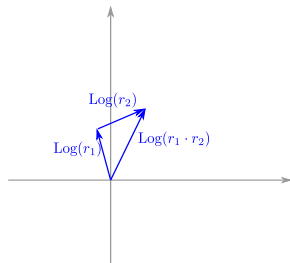
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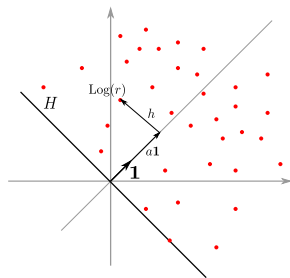
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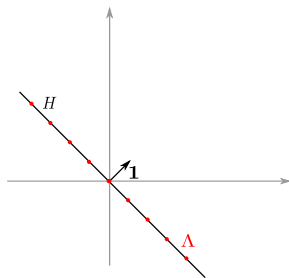
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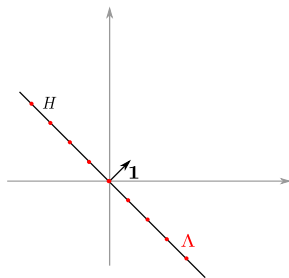
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### New objective

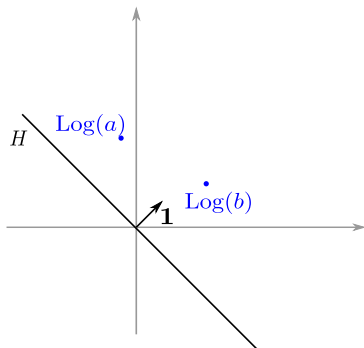
Find  $x, y \in R$  such that

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# Idea

Objective: find  $x, y \in R$  s.t.

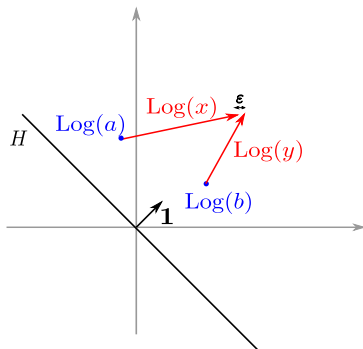
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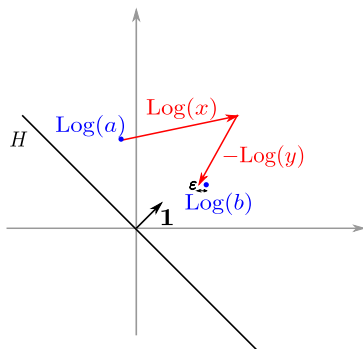
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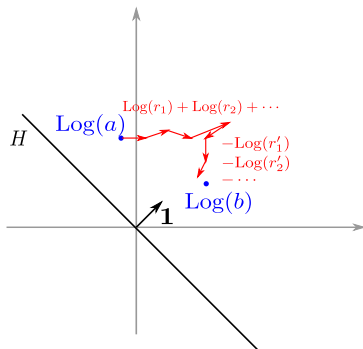




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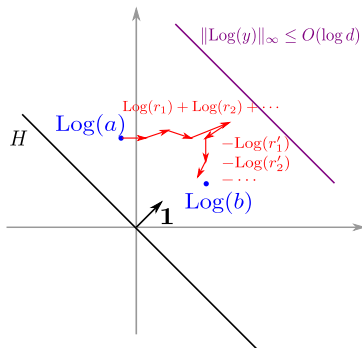
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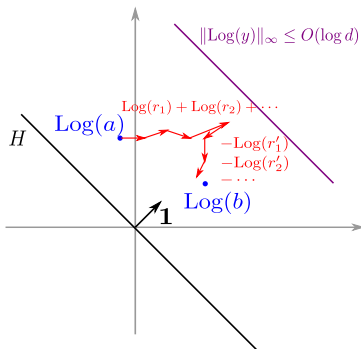
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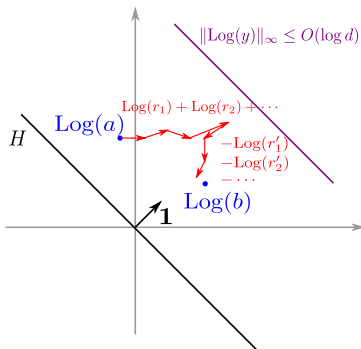
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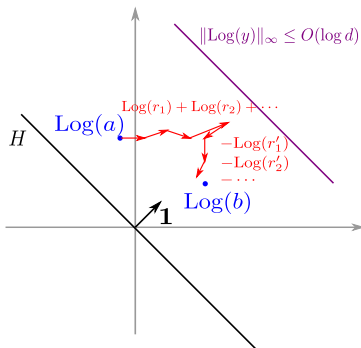
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Complexity

Poly time  
(with the oracle)

# Under the carpet

- Heuristics
  - ▶ maths justification
  - ▶ numerical experiments (in very small dimension)
- Any module / ideal
  - ▶ use pseudo-basis
  - ▶ add class group to  $L$  (cf [Buc88])
  - ▶ decompose ideals in class group  $\Rightarrow$  quantum algorithm
- Full LLL algo over  $K^2$ 
  - ▶ Lovász' swap condition
  - ▶ switch between  $\mathcal{N}(\cdot)$  and  $\|\cdot\|$

# Outline of the talk

- 1 LLL over  $K^2$ 
  - QR factorization
  - Euclidean division
- 2 LLL over  $K^k$

# Going to dimension $k$

## LLL over $\mathbb{Z}$

- $\gamma'$ -SVP in dim  $k$   
     $\leq$  1-SVP in dim 2
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  - ▶ poly time if oracle solving CVP in a fixed lattice  
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## Summary and impact

### LLL algorithm for power-of-two cyclotomic fields (or NTRUPrime)

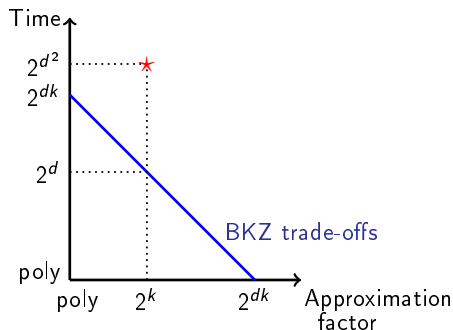
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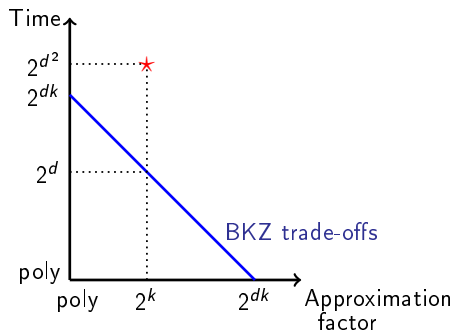


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$\Rightarrow$  theoretical result  
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# Conclusion

Open problems:

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