

An LLL algorithm for module lattices

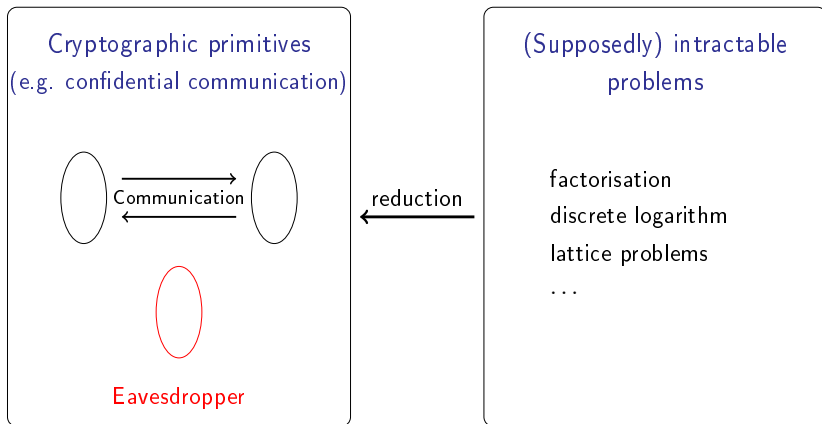
Changmin Lee¹, **Alice Pellet-Mary**², Damien Stehlé¹
and Alexandre Wallet³

¹ ENS de Lyon, ² KU Leuven, ³ NTT Tokyo

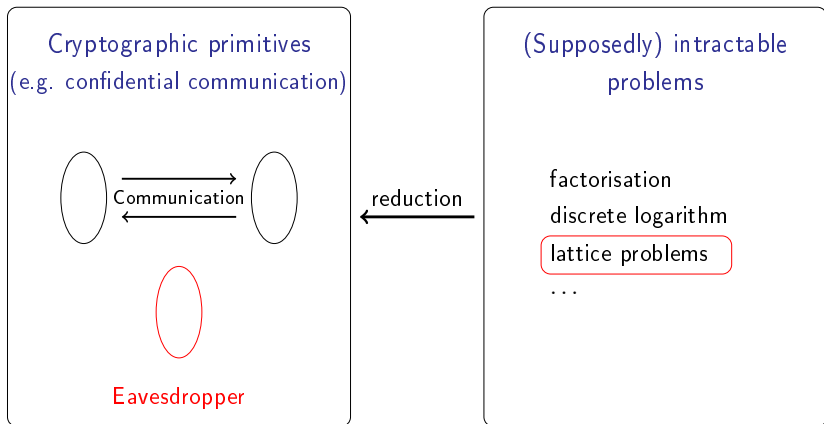
Institut Fourier,
January 16, 2020

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

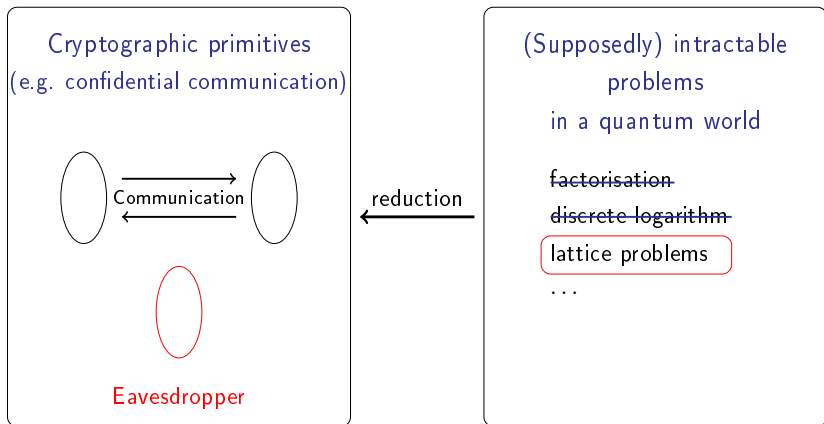
Cryptography and hard problems



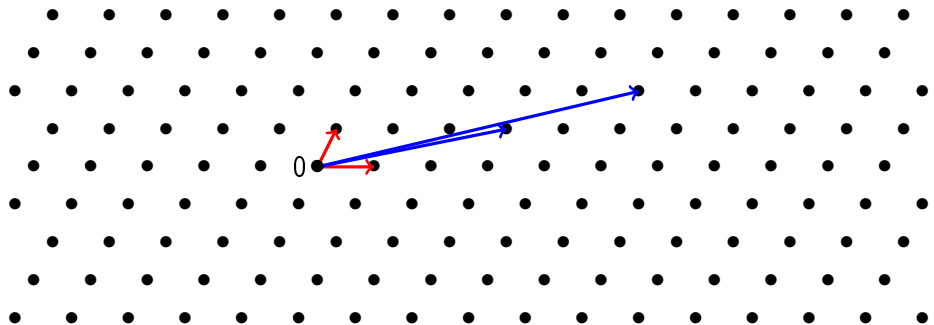
Cryptography and hard problems



Cryptography and hard problems



Lattices

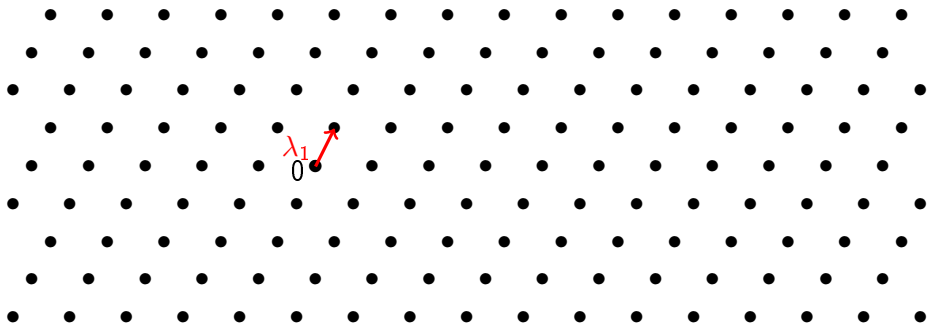


Lattice

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.

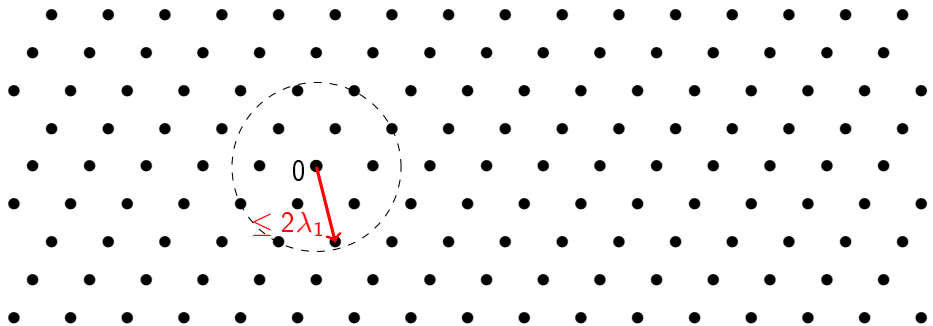
Lattice problems



Shortest Vector Problem (SVP)

Find a shortest (in Euclidean norm) non-zero vector.
Its Euclidean norm is denoted λ_1 .

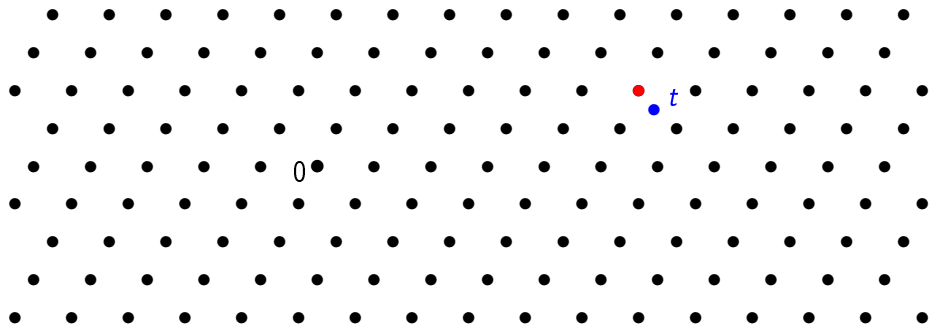
Lattice problems



Approximate Shortest Vector Problem (approx-SVP)

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

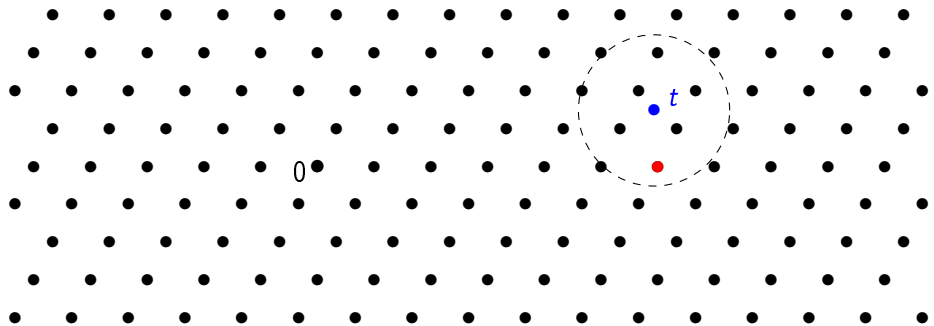
Lattice problems



Closest Vector Problem (CVP)

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

Lattice problems

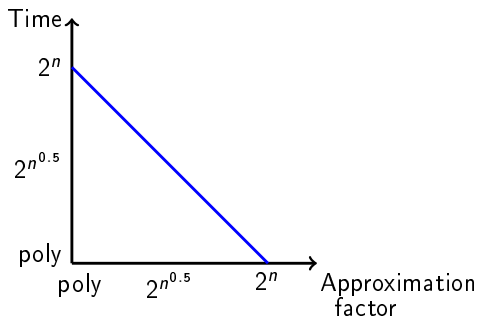


Approximate Closest Vector Problem (approx-CVP)

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]

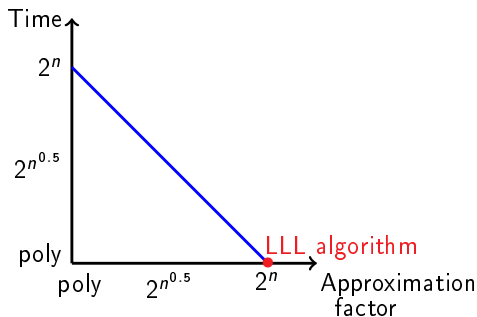


[Sch87] C.-P. Schnorr. A hierarchy of polynomial time lattice basis reduction algorithms. TCS.

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

Hardness of lattice problems

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



[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
- 11 using structured lattices

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Example: NIST post-quantum standardization process

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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

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

Ideal lattices

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

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

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

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

Ideal lattices

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

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

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

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} .

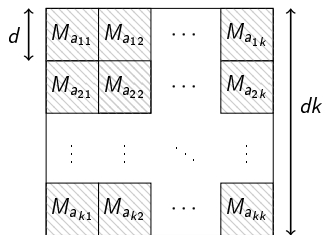
A (free) module lattice is $\sigma(\mathcal{M}) \subset \mathbb{R}^{kd}$

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 $\vec{b}_i = (a_{1i}, \dots, a_{ki})$.

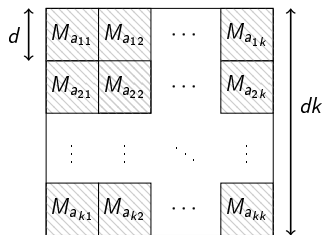


Basis of $\mathcal{L}(\mathcal{M})$ over \mathbb{Z}
($\dim n = dk$)

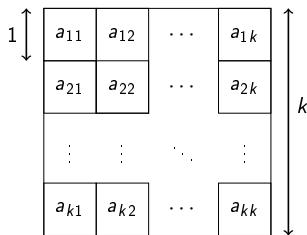
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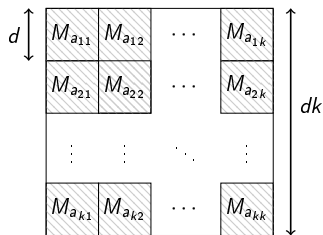


Basis of \mathcal{M} over R
($\dim k$)

Duality of module lattices

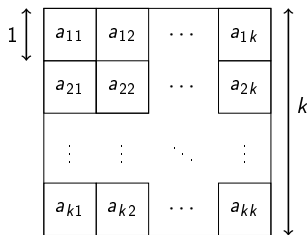
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Basis of $\mathcal{L}(\mathcal{M})$ over \mathbb{Z}
($\dim n = dk$)

Typically $500 \leq dk \leq 1000$



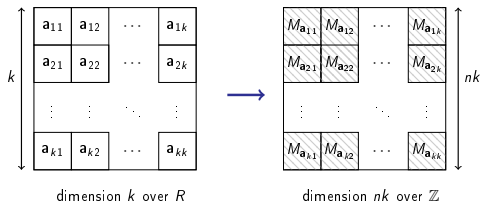
Basis of \mathcal{M} over R
($\dim k$)

Typically $k \leq 10$

Canonical embedding

Reminder

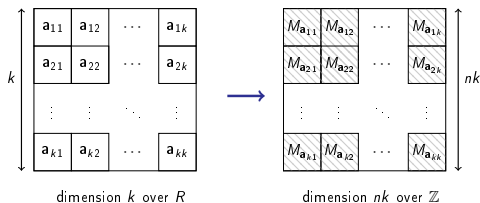
$$K = \mathbb{Q}[X]/P(X)$$



Canonical embedding

Reminder

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



Coefficient embedding

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

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

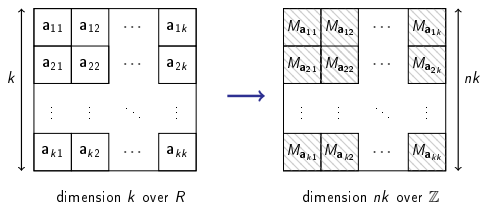
$$\mathbf{a} \longrightarrow M_{\mathbf{a}} = \begin{pmatrix} \left| \begin{array}{c} \sigma(\mathbf{a}) \\ \sigma(X\mathbf{a}) \\ \vdots \end{array} \right| & \left| \begin{array}{c} \sigma(X\mathbf{a}) \\ \sigma(X^2\mathbf{a}) \\ \vdots \end{array} \right| & \cdots & \left| \begin{array}{c} \sigma(X^{n-1}\mathbf{a}) \\ \sigma(X^n\mathbf{a}) \\ \vdots \end{array} \right| \end{pmatrix}$$

Canonical embedding

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

$\alpha_1, \dots, \alpha_n$ 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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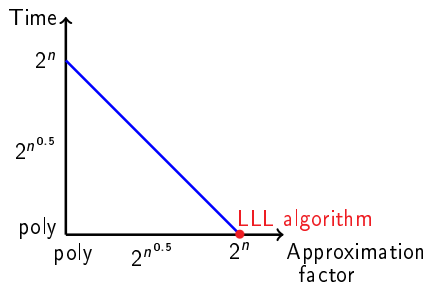
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Module lattices

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

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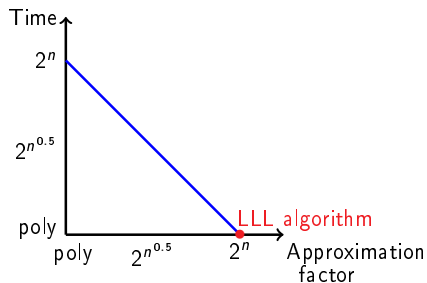
Lattice reduction over \mathbb{Z}

Module lattices

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

Objective

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Lattice reduction over \mathbb{Z}

Module lattices

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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

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

Previous works and result

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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

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

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

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

Outline of the talk

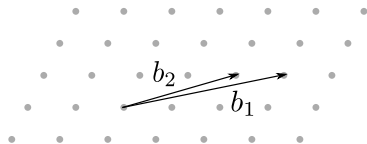
- 1 LLL over R^2
 - QR factorization
 - Euclidean division
- 2 LLL over R^k

Outline of the talk

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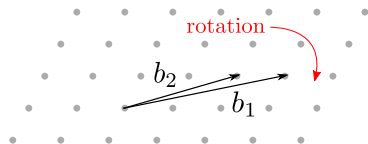
- 2 LLL over R^k

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



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

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

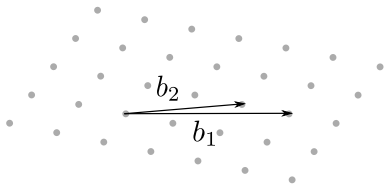


rotation

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

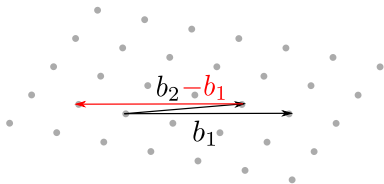
Compute QR factorization

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



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

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

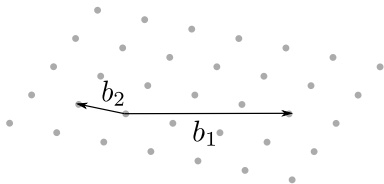


reduce b_2 with b_1

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

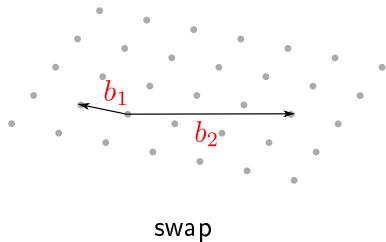
“Euclidean division” (over \mathbb{R})
of 7.3 by 10.2

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



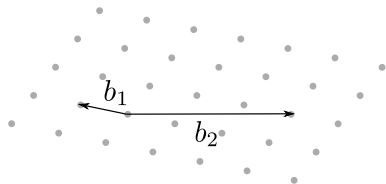
$$M = \begin{pmatrix} 10.2 & -2.9 \\ 0 & 0.6 \end{pmatrix}$$

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



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

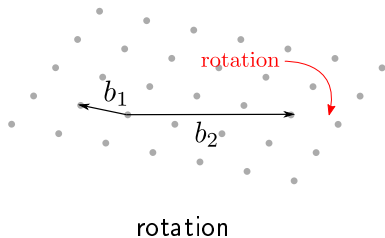
Lagrange-Gauss algorithm (over \mathbb{Z})



start again

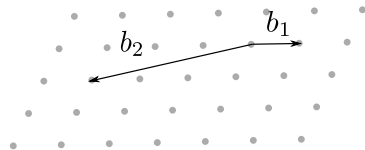
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Lagrange-Gauss algorithm (over \mathbb{Z})



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

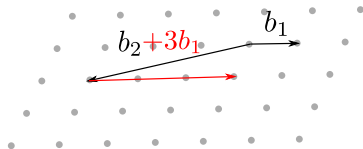
Lagrange-Gauss algorithm (over \mathbb{Z})



rotation

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

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

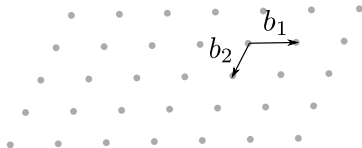


reduce b_2 with b_1

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

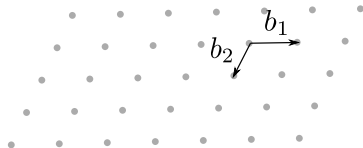
“Euclidean division” (over \mathbb{R})
of -10 by 3

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



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

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



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

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 $\vec{b} = (b_1, \dots, b_k) \in K^k$,

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

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

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Properties

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

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

Inner product over R

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Properties

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

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

QR factorization over R

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

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

QR factorization over R

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QR-factorisation: $B = QR$, with

- $r_{ii} = \|b_i^*\|_K$, $r_{ij} = \mu_{ji} r_{ii}$ for $i < j$ and $r_{ij} = 0$ otherwise
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QR factorization over R

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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

Non-Euclidean rings

Objective

Input: $a, b \in K$, $a \neq 0$

Output: $r \in R$ such that $\|b + ra\| \leq \|a\|/2$

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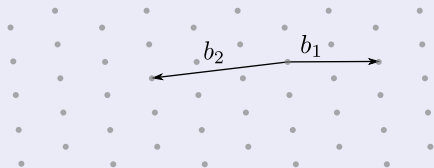
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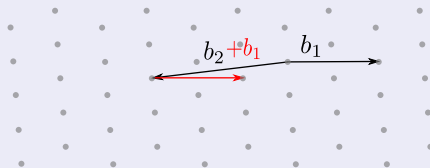
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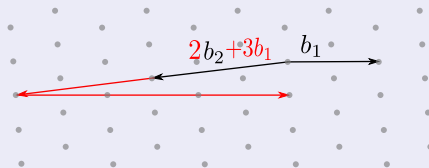
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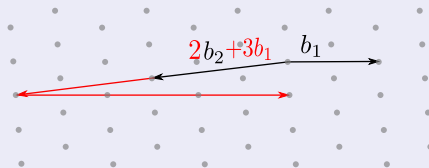
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\Rightarrow sufficient for Gauss' algo



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Reminder: $\sigma(r) = (r(\alpha_1), \dots, r(\alpha_d))^T$

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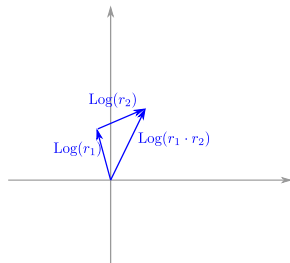
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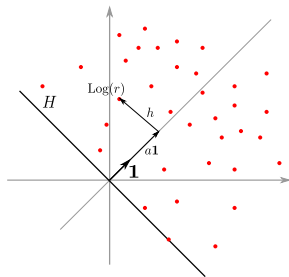
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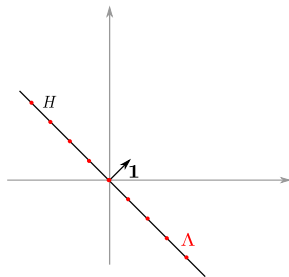
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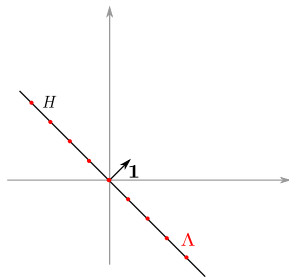
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- $\|r\| \simeq 2^{\|\text{Log } r\|_\infty}$



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New objective

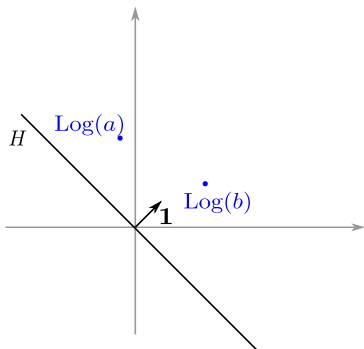
Find $x, y \in R$ such that

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Idea

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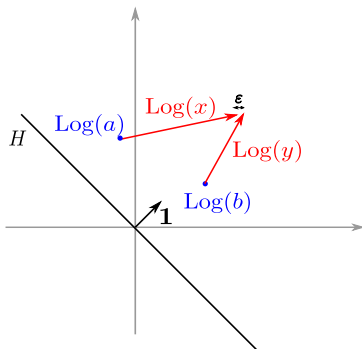
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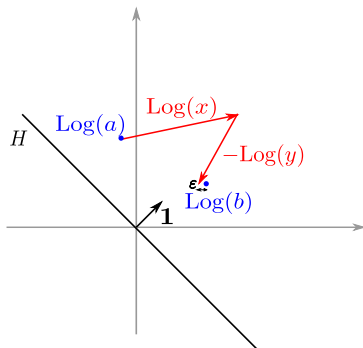
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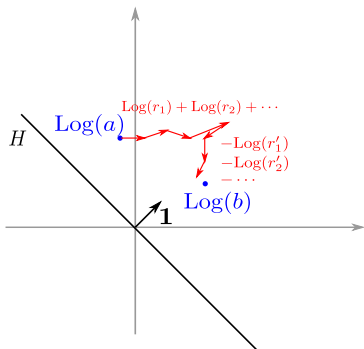
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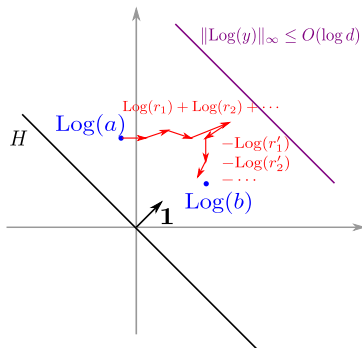
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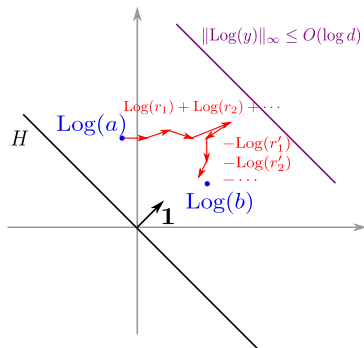


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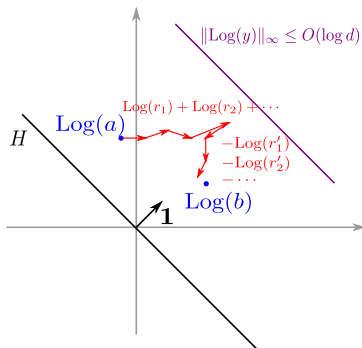
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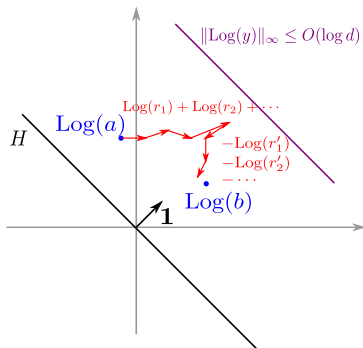
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Complexity

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 R^2
 - ▶ Lovász' swap condition
 - ▶ switch between $\mathcal{N}(\cdot)$ and $\|\cdot\|$
 - ▶ handling bit sizes

Outline of the talk

- 1 LLL over R^2
 - QR factorization
 - Euclidean division
- 2 LLL over R^k

Going to dimension k

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 - ▶ heuristic, quantum
 - ▶ poly time if oracle solving CVP in a fixed lattice
(depending only on R)

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

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Summary and impact

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

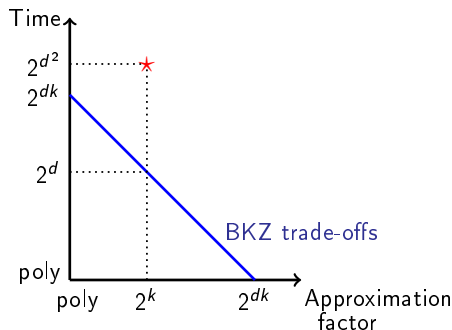
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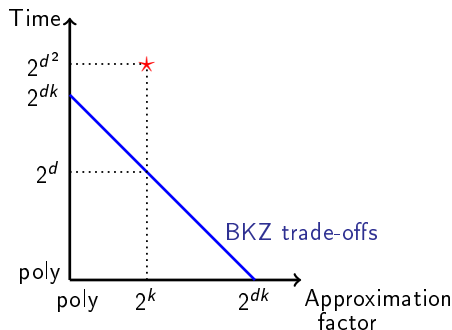


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\Rightarrow theoretical result
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