

An LLL algorithm for module lattices

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and Alexandre Wallet³

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Lattices: Geometry, Algorithms and Hardness
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KU LEUVEN



What is this talk about?

LLL-type algorithm for modules over a ring of integers

- all number fields
- approx-factor \approx exponential in module rank
- quantum poly-time ... given a CVP oracle depending on K

Structured lattices

Motivation (crypto)

Improve efficiency of lattice-based schemes using structured lattices

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- 12 lattice-based
- 11 using structured lattices

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	Frodo (unstructured lattices)	Kyber (structured lattices)
public key size (in Bytes)	9 616	800
ciphertexts size (in Bytes)	9 720	736

(CCA2 KEMs, round 2 version, security level 1)

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All 11 schemes use **module lattices**

Module

- $K = \mathbb{Q}[X]/P(X)$
- $R = \mathbb{Z}[X]/P(X)$ (or $R = \mathcal{O}_K$)

with P monic and irreducible, degree d

Module

A (free) module M is a subset of K^k of the form $M = \{B\vec{x} \mid \vec{x} \in R^k\}$, with $B \in K^{k \times k}$ invertible. B is a **basis** of M ; k is the **rank** of M .

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A module is a 'lattice' of rank k over R .

Reminder:

- $K = \mathbb{Q}[X]/P(X)$ and $R = \mathbb{Z}[X]/P(X)$
- $M = \{B\vec{x} : \vec{x} \in R^d\}$, with $B = (\vec{b}_1, \dots, \vec{b}_k)$ linearly independent

Module lattice

Reminder:

- $K = \mathbb{Q}[X]/P(X)$ and $R = \mathbb{Z}[X]/P(X)$
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Canonical embedding

$$\begin{aligned}\sigma : K &\rightarrow \mathbb{C}^d \\ x &\mapsto (x(\alpha_1), \dots, x(\alpha_d))\end{aligned}$$

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$$\sigma : K^k \rightarrow \mathbb{C}^{kd}$$

$$(x_1, \dots, x_k) \mapsto (\sigma(x_1) \parallel \dots \parallel \sigma(x_k))$$

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$$\mathcal{L}(M) = \{\sigma(\vec{x}) : \vec{x} \in M\} \subset \mathbb{C}^{kd}$$

$\mathcal{L}(M)$ is a lattice of rank kd , spanned by $(\sigma(\vec{b}_1), \sigma(x\vec{b}_1), \dots, \sigma(x^{d-1}\vec{b}_k))$

Motivation

Module lattices

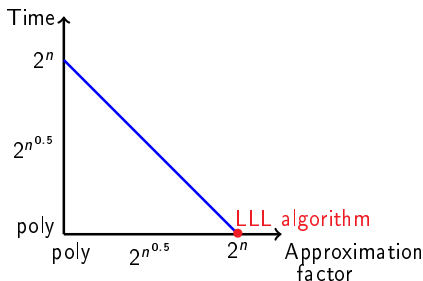
- dimension $n = kd$ over \mathbb{Z}
- dimension k over R

Motivation

Module lattices

- dimension $n = kd$ over \mathbb{Z}
typically $500 \leq kd \leq 1500$
- dimension k over R
typically $k \leq 10$

Motivation



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Lattice reduction over \mathbb{Z}

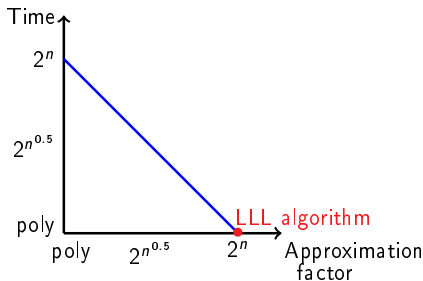
(in blue: BKZ trade-offs [Sch87, SE94])

[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.

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

Motivation



Lattice reduction over \mathbb{Z}
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Module lattices

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typically $500 \leq kd \leq 1500$
- dimension k over R
typically $k \leq 10$

Can we extend the LLL algorithm to lattices over R ?

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What is small in K ?

Remember

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Two generalization of $|\cdot|$:

- $\|x\| := \|\sigma(x)\|$ (Euclidean norm)
- $\mathcal{N}(x) := \prod_i |x(\alpha_i)|$ (algebraic norm)

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$$\|x\| \text{ small} \not\Rightarrow \mathcal{N}(x) \text{ small}$$

Previous works and result

	Bound on quality	Bound on runtime
Napias [Nap96]: <ul style="list-style-type: none">▶ specific number fields	✗	✗

[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] C. Fieker, M. E. Pohst. Lattices over number fields. ANTS.

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Kim-Lee [KL17]: <ul style="list-style-type: none">▶ norm Euclidean fields▶ biquadratic norm euclidean	✗ ✓	✓ ✓

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

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

Outline of the talk

- 1 The LLL algorithm
- 2 The Lagrange-Gauss algorithm
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High-level overview of LLL

LLL over \mathbb{Z}

- γ' -SVP in $\dim k$
 \leq 1-SVP in $\dim 2$
 - ▶ $\gamma' = 2^{O(k)}$
 - ▶ poly time

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

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LLL over $R = \mathbb{Z}[X]/(X^d + 1)$

- γ' -SVP in rank- k
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- Algorithm for γ -SVP in rank-2
 - ▶ $\gamma = 2^{(\log d)^{O(1)}}$
 - ▶ heuristic, quantum
 - ▶ poly time if oracle solving CVP in a
fixed lattice
(depending only on R)

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LLL over $R = \mathbb{Z}[X]/(X^d + 1)$

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 $\leq \gamma$ -SVP in rank- k

needs QR-factorisation

- Algorithm for γ -SVP in rank-2

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- ▶ poly oracle solving CVP in a fixed lattice
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next section

Inner product over R

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

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

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Properties

- $\|\sigma(\vec{a})\| = \|\vec{a}\|$

$$\|x\| = \|\sigma(x)\|$$

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Properties

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

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

QR factorization over R

Let $B = (\vec{b}_1, \dots, \vec{b}_k) \in K^k$, define

$$\vec{b}_i^* = \vec{b}_i - \sum_{j < i} \mu_{ij} \vec{b}_j^*, \quad \text{with } \mu_{ij} = \frac{\langle \vec{b}_i, \vec{b}_j^* \rangle_K}{\langle \vec{b}_j^*, \vec{b}_j^* \rangle_K}$$

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QR-factorisation: $B = QR$, with

- $r_{ii} = \|\vec{b}_i^*\|_K$, $r_{ij} = \mu_{ji} r_{ii}$ for $i < j$ and $r_{ij} = 0$ otherwise
- columns of Q are $\vec{b}_i^* / \|\vec{b}_i^*\|_K$

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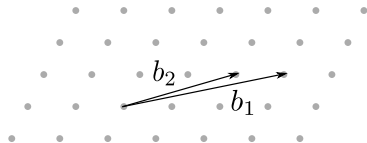
Properties

- $\forall \vec{x}, \vec{y}, \quad \langle B\vec{x}, B\vec{y} \rangle_K = \langle R\vec{x}, R\vec{y} \rangle_K$
- $\forall \vec{v} \in \mathcal{L}(B), \quad \mathcal{N}(\|\vec{v}\|_K) \geq \min \mathcal{N}(r_{ii})$

Outline of the talk

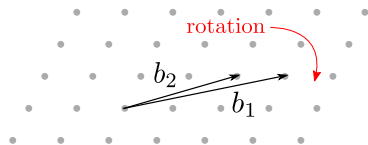
- 1 The LLL algorithm
- 2 The Lagrange-Gauss algorithm
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Lagrange-Gauss algorithm (over \mathbb{Z})



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

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

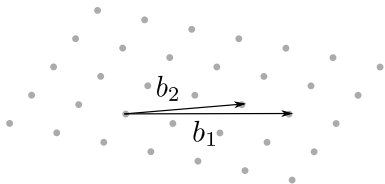


rotation

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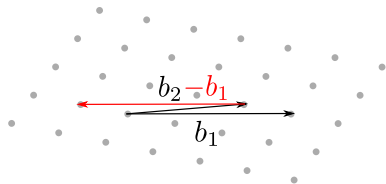
Compute QR factorization

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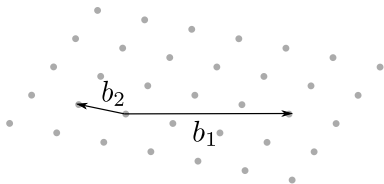


reduce b_2 with b_1

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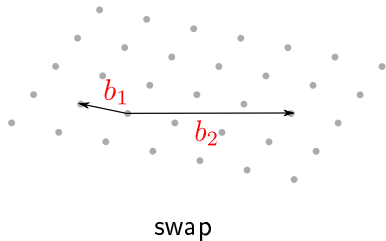
“Euclidean division” (over \mathbb{R})
of 7.3 by 10.2

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



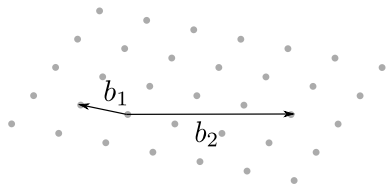
$$M = \begin{pmatrix} 10.2 & -2.9 \\ 0 & 0.6 \end{pmatrix}$$

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



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

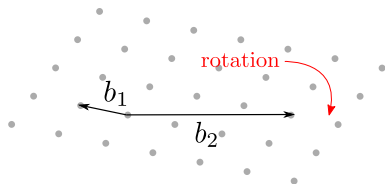
Lagrange-Gauss algorithm (over \mathbb{Z})



start again

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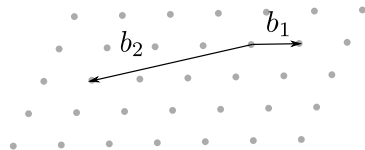
Lagrange-Gauss algorithm (over \mathbb{Z})



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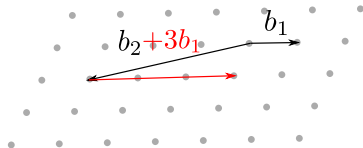
Lagrange-Gauss algorithm (over \mathbb{Z})



rotation

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

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

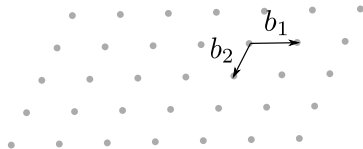


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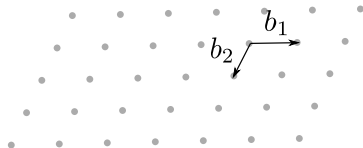
“Euclidean division” (over \mathbb{R})
of -10 by 3

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



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

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

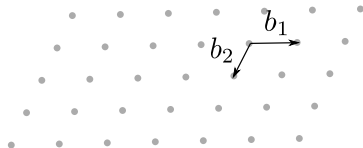


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For the Lagrange-Gauss algorithm over R , we need

- Rotation
- Euclidean division

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 \Rightarrow ok
- Euclidean division \Rightarrow ?

Euclidean division

Over \mathbb{Z}

Input: $a, b \in \mathbb{Z}, a \neq 0$

Output: $r \in \mathbb{Z}$

such that $|b + ra| \leq |a|/2$

Euclidean division

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CVP in \mathbb{Z} with target $-b/a$.



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Over R

CVP in R with target $-b/a$

\Rightarrow output $r \in R$



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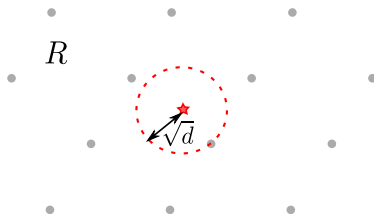


Over R

CVP in R with target $-b/a$
 \Rightarrow output $r \in R$

Difficulty: Typically

$\|b + ra\| \approx \sqrt{d} \cdot \|a\| \gg \|a\|$.



Euclidean division

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Over R

CVP in R with target $-b/a$

\Rightarrow output $r \in R$

Difficulty: Typically

$\|b + ra\| \approx \sqrt{d} \cdot \|a\| \gg \|a\|$.

Relax the requirement

Find $x, y \in R$ such that

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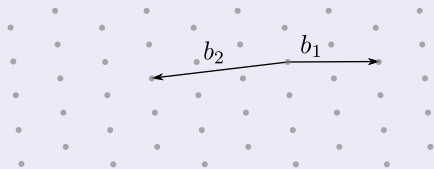
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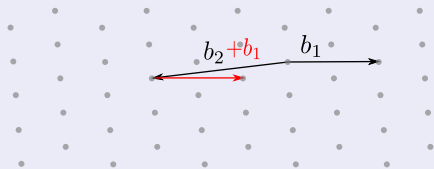
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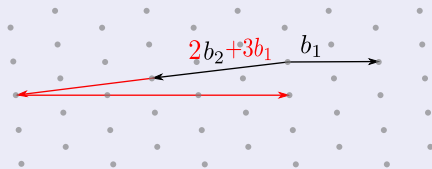
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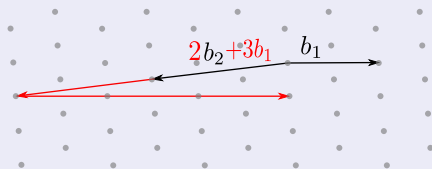
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\Rightarrow sufficient for Gauss' algo



Outline of the talk

- 1 The LLL algorithm
- 2 The Lagrange-Gauss algorithm
- 3 Computing the relaxed Euclidean division

The Log space

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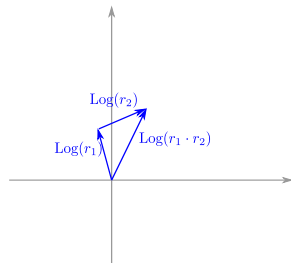
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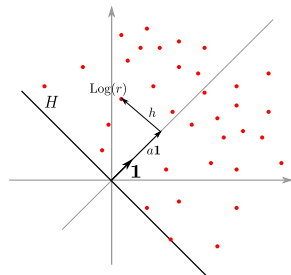
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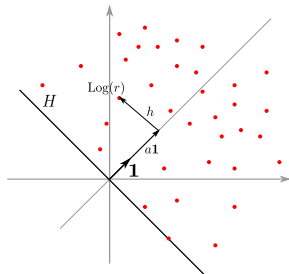
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Extended logarithm

Let's say $d = 1$

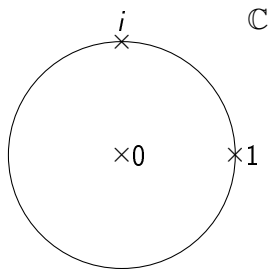
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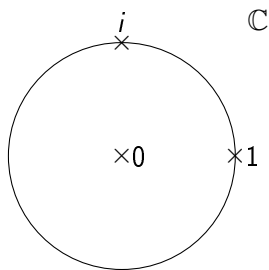


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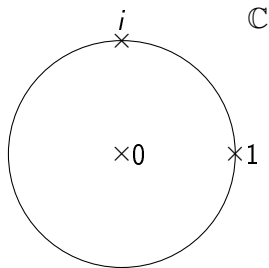
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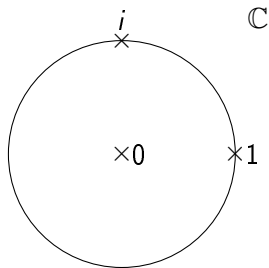
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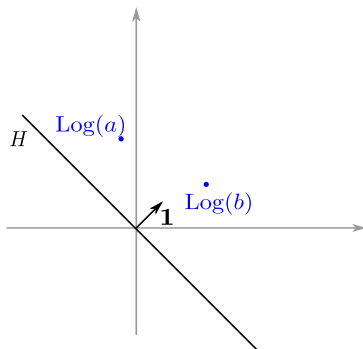
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Idea of the algorithm

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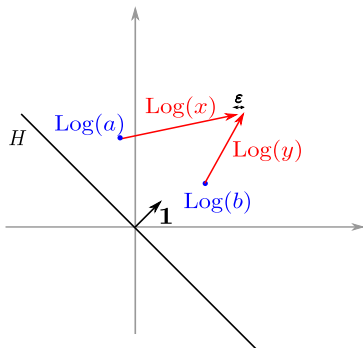
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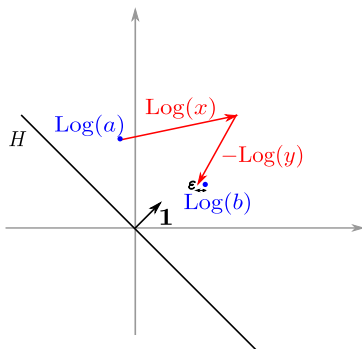
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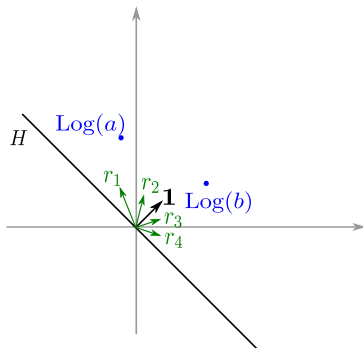
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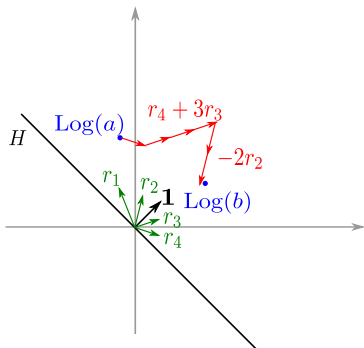


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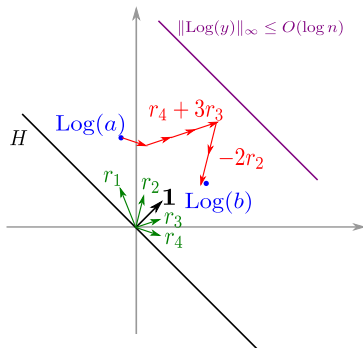


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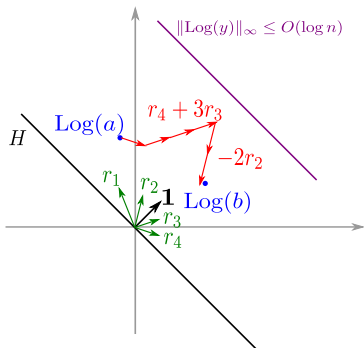
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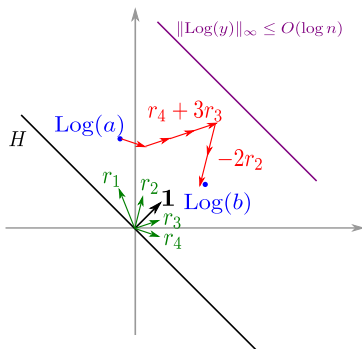
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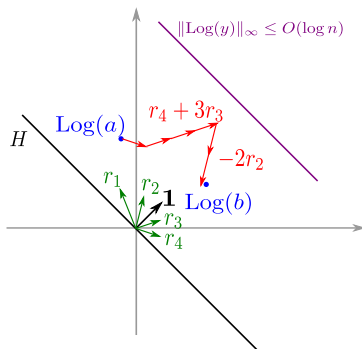
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Complexity

Quantum poly time
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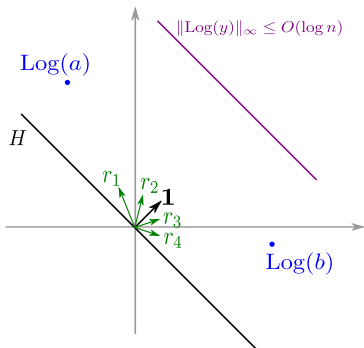
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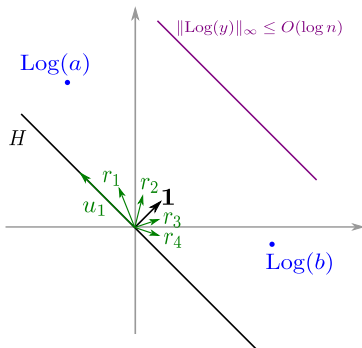
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 - ▶ assuming $(\beta \cdot x \| \vec{v})$ with $v \stackrel{\$}{\leftarrow} \{\vec{w} \in \{0, 1\}^{d^2} \mid \|\vec{w}\|_1 = d\}$ implies x (almost) uniform mod U .

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 - ▶ related to Arakelov random walk \Rightarrow see Léo and Benjamin's talk

Under the carpet

- Any module
 - ▶ use pseudo-basis
 - ▶ add class group to L (cf [Buc88])
- Proving correctness
 - ▶ Lovász' swap condition
 - ▶ switch between $\mathcal{N}(\cdot)$ and $\|\cdot\|$
 - ▶ handling bit sizes

[Buc88] J. Buchmann. A subexponential algorithm for the determination of class groups and regulators of algebraic number fields. Séminaire de théorie des nombres.

Summary and impact

LLL algorithm for cyclotomic fields

- Approx: $\text{quasi-poly}(d)^{O(k)} = 2^{(\log d)^{O(1)} \cdot k}$
- Time: quantum polynomial time
if **oracle** solving CVP in L (of dim $O(d^{2+\epsilon})$)

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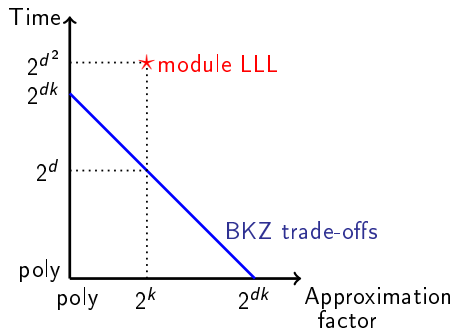
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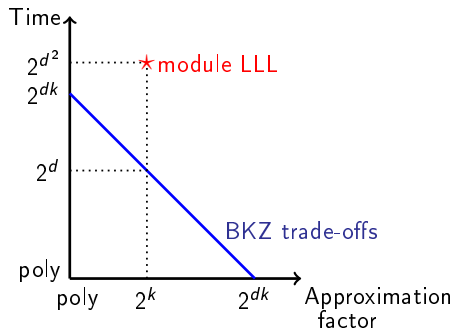


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\Rightarrow theoretical result
(not practical)

Open problems

(1) Improving CVP in L

- ▶ reduce its dimension to $\tilde{O}(d)$?
- ▶ use the structure of L to improve CVP solvers?

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- ▶ use the structure of L to improve CVP solvers?
- ▶ CVP with pre-processing is NP-hard [Mic01]

“CVPP in NP-hard”: $\forall n, \exists L, \forall \phi$ (SAT instance with n variables), $\exists \vec{t}$ and α s.t.

$$\phi \text{ is satisfiable} \Leftrightarrow \text{dist}(L, \vec{t}) \leq \alpha.$$

[Mic01] D. Micciancio. The hardness of the closest vector problem with preprocessing. Transaction on Information Theory.

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- ▶ use the structure of L to improve CVP solvers?
- ▶ CVP with pre-processing is NP-hard [Mic01]

(2) Generalizing LLL to all the BKZ trade-offs?

- ▶ reduction rank k to rank β (cf [MS19])
- ▶ sieving/enumeration in modules?

[MS19] T. Mukherjee, N. Stephens-Davidowitz. Lattice Reduction for Modules, or How to Reduce ModuleSVP to ModuleSVP, ePrint.

Open problems

(1) Improving CVP in L

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Thank you