# Hermitian lattices reduction

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• Part I: LLL algorithm for hermitian lattices

• Part II: Representations of fractional ideals

The theory of euclidean lattices and its algorithmic approach are well-known, but there are few studies of the algorithmic side for hermitian lattices.



H. Lenstra



A. Lenstra



L. Lovász

## The inventors of the LLL algorithm

# Part I: LLL algorithm for hermitian lattices

# 1 Introduction

2 Hermitian lattices over a quadratic euclidean number field

# LLL-reduction LLL-reduction for hermitian lattices Usefulness for the SVP Computing LLL-reduced basis

## Probabilistic analysis

Average case Experimental results Let  $K = \mathbb{Q}(i\sqrt{d})$  with  $d \in \{1, 2, 3, 7, 11\}$  and  $\mathbb{Z}_K$  be its maximal order.

#### Definition

A subgroup  $\Lambda$  of  $\mathbb{C}^m$  is called a  $\mathbb{Z}_K$ -lattice if there exists  $(e_1, \ldots, e_m)$  a  $\mathbb{C}$ -basis of  $\mathbb{C}^m$  such that  $\Lambda = \mathbb{Z}_K e_1 \oplus \cdots \oplus \mathbb{Z}_K e_m$ .

A  $\mathbb{Z}_{\mathcal{K}}$ -lattice in  $\mathbb{C}^m$  may be described as a  $\mathbb{Z}$ -lattice in  $\mathbb{R}^{2m}$ .

#### Definition

The minimal norm of  $\Lambda$  is  $\lambda_1(\Lambda) = \min_{x \in \Lambda \setminus \{0\}} ||x||^2$ .

# How to compute $\lambda_1(\Lambda)$ and a minimal vector of $\Lambda$ ?

# LLL-reduction for hermitian lattices

Let  $\mathcal{E} = (e_1, \dots, e_m)$  be a  $\mathbb{C}$ -basis of  $\mathbb{C}^m$ . We denote by  $e_i^*$  and  $\mu_{i,j}$  its Gram-Schmidt orthogonalization.

Let  $0 < m_K < \delta < 1$ , where  $m_K$  is the euclidean minima of K:

$$m_{\mathcal{K}} = \sup_{x \in \mathbb{C}} \inf_{y \in \mathbb{Z}_{\mathcal{K}}} |x - y|^2,$$

#### Definition

The basis  $\mathcal{E}$  is said  $\delta$ -LLL-reduced if:

$$\begin{cases} |\mu_{i,j}|^2 \leqslant m_{\mathcal{K}} & \text{for } 1 \leqslant j < i \leqslant m, \\ \|e_i^*\|^2 \geqslant (\delta - |\mu_{i,i-1}|^2) \|e_{i-1}^*\|^2 & \text{for } 2 \leqslant i \leqslant m. \end{cases}$$

Computing a LLL-reduced basis of a  $\mathbb{Z}_K$ -lattice allow to approximate its minimal norm by giving a quasi-minimal vector.

#### Theorem

Let  $\mathcal{E}$  be a  $\delta$ -LLL-reduced basis of a  $\mathbb{Z}_{\mathcal{K}}$ -lattice  $\Lambda$  in  $\mathbb{C}^m$ . Then

$$\|e_1\|^2 \leq \left(\frac{1}{\delta - m_K}\right)^{m-1} \lambda_1(\Lambda).$$

#### Idea [Napias, Gan/Ling/Mow]

The original LLL algorithm (over  $\mathbb{Z}$ ) can be generalised for  $\mathbb{Z}_{K}$ -lattices.

Therefore, one may compute a  $\delta$ -LLL-reduced basis of a  $\mathbb{Z}_{K}$ -lattice  $\Lambda$  from one of its basis  $\mathcal{E} = (e_1, \dots, e_m)$  using

$$\mathcal{O}\left(m^4\log_{\delta}\left(rac{\lambda_1(\Lambda)^{1/2}}{\|\mathcal{E}\|_{\infty}}
ight)
ight)$$

operations in  $\mathbb{C}$ .

The bound  $||e_1||^2 \leq \left(\frac{1}{\delta - m_K}\right)^{m-1} \lambda_1(\Lambda)$  has been proven using  $|\mu_{i,i-1}|^2 = m_K$ : this is the worst case, which is unrealistic.

#### Theorem

Let  $\mathcal{E} = (e_1, \ldots, e_m)$  be a basis of a  $\mathbb{Z}_K$ -lattice  $\Lambda$  in  $\mathbb{C}^m$ , to which the  $\delta$ -LLL algorithm is applied. Assuming that the coefficients  $|\mu_{i,i-1}|^2$  of the GSOP of  $\mathcal{E}$  are identically distributed random variables of density p, we get that:

$$\mathbb{E}(\log(\|e_1\|^2)) \leqslant \log(\lambda_1(\Lambda)) - (m-1)\int_0^{m_{\kappa}} \log(\delta - x)p(x)dx.$$

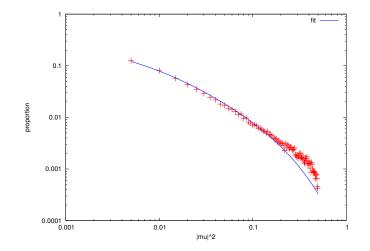
The density p has been approximated using experimental data.

Simple implementation in GP ( $\approx$  400 lines). Tested on 500 bases in various dimension (50 to 150).

D	1	2	3	7	11
m <sub>K</sub>	0.5	0.75	0.3333333	0.5714286	0.8181818
$\int_0^{m_\kappa} \log(\delta - x) p(x) dx$	- 0.0765100	- 0.09183234	- 0.0708416	- 0.0796641	- 0.0927955
$\frac{1/(\delta - m_k)}{\exp\left(-\int_0^{m_K}\log(\delta - x)\tilde{p}(x)dx\right)}$	1.8904972	3.8010754	1.4186946	2.2061385	6.3860367

$$p(x) = \begin{cases} \frac{a}{x+b}e^{-x/c} & \text{if } x \in [0, m_K], \\ 0 & \text{otherwise.} \end{cases}$$

# Distribution and interpolation obtained in $\mathbb{Q}(i)$ for $\delta = 0.99$ (logarithmic scale)



Similar results for other fields.

• Part I: LLL algorithm for hermitian lattices

• Part II: Representations of fractional ideals

Let K be a number field of degree d and  $\mathbb{Z}_K$  be its ring of integers.

#### Definition

A fractional ideal of K is a  $\mathbb{Z}_K$ -submodule  $\mathfrak{a}$  of K for which one may find  $\zeta \in \mathbb{Z}_K$  such that  $\zeta \mathfrak{a} \subset \mathbb{Z}_K$ . In this case, one may find a  $\mathbb{Q}$ -basis of K which is a  $\mathbb{Z}$ -basis of  $\mathfrak{a}$ .

# How to represent ideals in an algorithmic setting?

In PARI/GP:

- HNF representation (idealhnf)  $\rightarrow$  easy to use.
- Two-element representation (idealtwoelt)  $\rightarrow$  memory-friendly.

# Part II: representations of a fractional ideal

# **6** Introduction

### 6 Matrix representation

#### Two-element representation

naive algorithm Strong reduction, variable success rate Weak reduction, bounded failure rate Experimental results Let a be an integral ideal of K and  $\omega = (\omega_1, \ldots, \omega_d)$  be an integral basis of K. We consider  $\mathcal{E} = (e_1, \ldots, e_d)$  a  $\mathbb{Z}$ -basis of a.

#### Matrix representation of a

The ideal  $\mathfrak a$  may be represented  $\mathfrak a$  by the coordinates matrix of  $\mathcal E$  with respect to  $\omega.$ 

It gives a representation of  $\mathfrak{a}$  as an element of  $M_d(\mathbb{Z}) \cap GL_d(\mathbb{Q})$ .

Uniqueness of such a representation is achieved by choosing a specific basis of  $\mathfrak a$  (i.e HNF).

# Two-element representation: naive algorithm

Let  $\mathfrak{a}$  be an integral ideal of K.

#### Classical result

Let x be a non-zero element of a. There exists  $y \in \mathfrak{a}$  such that  $\mathfrak{a} = (x, y)$ . Moreover, an element y chosen uniformly at random in  $\mathfrak{a}/(x)$  satisfies  $(x, y) = \mathfrak{a}$  with probability:

$$\mathbb{P}[(x,y) = \mathfrak{a}] = \prod_{\mathfrak{p} : v_{\mathfrak{p}}(x) > v_{\mathfrak{p}}(\mathfrak{a})} \left(1 - \frac{1}{\mathcal{N}(\mathfrak{p})}\right) \geq \prod_{\mathfrak{p} \mid \mathfrak{a}} \left(1 - \frac{1}{\mathcal{N}(\mathfrak{p})}\right).$$

#### **Problems:**

- Maximise the shortness of such a representation.
- Success rate depends on a.

# Strong reduction, variable success rate

Lets add a size-reduction condition to the naive algorithm:

### Algorithm 1

- **1** Choose  $x \in \mathfrak{a}$  short (w.r.t the  $T_2$  norm), using the LLL-algorithm.
- **2** Find  $y \in \mathfrak{a}$  such that  $(x, y) = \mathfrak{a}$ , using naïve algorithm.
- **3** Size-reduce *y*.

It produces a representation  $(x, y) = \mathfrak{a}$  such that:

```
\max\{\|x\|,\|y\|\} \in \mathcal{O}(\mathcal{N}(\mathfrak{a})^{1/d}).
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 $\rightarrow$  Strong reduction, but no changes on the success rate.

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Implemented in GP(2C) (\approx 100 lines in C).
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# Weak reduction, bounded failure rate

Lets add a size-reduction to the algorithm used in the function idealtwoelt of  $\mathsf{GP}$ :

# Algorithm 2 [Fieker/Sthelé]

- Find b ⊂ a such that p|b implies N(p) ≥ y, for y a well-chosen constant.
- ${\it 2}$  Find a small two-element representation of  ${\frak b},$  using the previous algorithm.
- **3** Recover a two-element representation of  $\mathfrak{a}$  from the one of  $\mathfrak{b}$ .

It produces a representation  $(x, y) = \mathfrak{a}$  such that:

 $\max\{\|x\|,\|y\|\}\in \mathcal{O}(\mathcal{N}(\mathfrak{a})^{4/d}).$ 

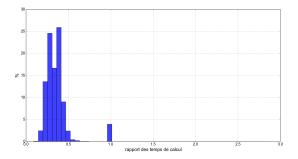
 $\rightarrow$  Weaker size-reduction, increase of the overall complexity, but the failure rate is bounded (depending on a "success parameter" *t*):

 $\mathbb{P}[\mathsf{failure}] \leqslant 0.8^t$ 

Implemented in GP(2C) ( $\approx$  500 lines in C).

# Heuristic remarks (WiP)

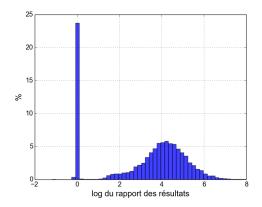
Ratio  $\frac{\rm time~algorithm~1}{\rm time~algorithm~2}$  over all integral ideals of norm  $\leqslant 5\cdot 10^4$  in a field of degree 25:



Despite the bounded failure rate, algorithm 2 tends to be way slower than algorithm 1. It seems that the control of the success rate does not outweigh the complexity explosion.

# Heuristic remarks (WiP)

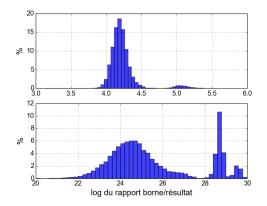
Ratio log  $\frac{\rm result\ algorithm\ 2}{\rm result\ algorithm\ 1}$  over all integral ideals of norm  $\leqslant 5\cdot 10^4$  in a field of degree 25:



As foreseen, algorithm 1 usually produces shorter representations than algorithm 2.

# Heuristic remarks (WiP)

Ratio log  $\frac{\rm theoretical\ bound}{\rm result\ algorithm}$  over all integral ideals of norm  $\leqslant 5\cdot 10^4$  in a field of degree 25:



The theoretical bounds on the size of the elements seem to be quite large for both algorithms.

# Thanks for listening!

References:

- Napias: A generalization of the LLL-algorithm over euclidean rings or orders(Journal de théorie des nombres de Bordeaux, 1996).
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- Fieker/Stehlé: Short bases of lattices over number fields (ANT, 2010).