

Exposé Inria

An LLL algorithm for module lattices

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eprint 2019/1035

I) Introduction:

* Lattice, SVP, CVP \rightarrow norme = $\|\cdot\|_2$

Schnorr87, Schnorr-Euchner94

and sometimes $\|\cdot\|_\infty$
Lenstra-Lenstra-Lovasz



* Lattice reduction : BKZ, LLL⁸².
(compute a short basis \rightarrow helps with SVP and CVP)

* module lattices :

- lattice : storage n^2 (cf codes)

\hookrightarrow structured lattices

en colonne \rightarrow $M_a = \begin{bmatrix} a_0 & \dots & a_{n-1} \\ a_{n-1} & \dots & a_0 \\ \vdots & & \vdots \end{bmatrix}^T$ (cf cyclic codes)

\hookrightarrow mult by $a_0 + a_1x + \dots + a_{n-1}x^{n-1} \bmod x^n - 1$

$$R = \mathbb{Z}[x]/x^n - 1$$

$$\sigma: R \rightarrow \mathbb{R}^n \text{ (or even } \mathbb{Z}^n)$$

$$a \rightarrow (a_0, \dots, a_{n-1})$$

$$M_a = \begin{bmatrix} \sigma(a) & \sigma(xa) & \dots & \sigma(x^{n-1}a) \end{bmatrix}$$

$$L(M_a) = \sigma(\langle a \rangle)$$

\hookrightarrow The lattice is an ideal in R
(via σ)
called "ideal lattice"

In more generality: ideal lattices $L = \sigma(I)$

* I ideal maybe not principal

↳ but in this talk always principal (for simplicity)

* R can be another ring

~~(recall)~~ ↳ for this work: $R = \text{ring of integers of } K$

for $K = \mathbb{Q}[x]/P(x)$
 ↑
 number field

P irreducible
 monic
 degree d

for simplicity: can think of $R = \mathbb{Z}[x]/p(x)$

Eg: ~~the~~ $R = \mathbb{Z}[x]/x^{d+1}$ $d=2^k$ (power of 2)
 cyclo

$R = \mathbb{Z}[x]/x^d - x - 1$ d prime (NTRU prime)

Module lattice: fix some K and $R = \mathbb{Z}[x]/p$

$$\mathcal{B} = \begin{bmatrix} M_{11} & M_{12} \\ \vdots & \vdots \\ M_{d1} & M_{d2} \end{bmatrix} \quad \begin{array}{c} H_{\text{arr}} \\ \uparrow d \\ \uparrow dr \end{array}$$

recall $H_{\alpha} = \begin{bmatrix} \sigma(\alpha) & \cdots & \sigma(\alpha^{d-1}) \end{bmatrix}$

$L(\mathcal{B})$ is a (free) module lattice

why the name? $\vec{b_i} = (a_{i1}, a_{i2}, \dots, a_{id}) \in R^d \rightarrow K\text{-linearly indep}$

$$M = \left\{ \sum x_i \vec{b_i} \mid x_i \in R \right\}$$

is an R -module in K^d

$$\text{and } L(\mathcal{B}) = \underbrace{\sigma(\vec{b_1}) \dots \sigma(\vec{b_d})}_{\text{(free)}}$$

$(\sigma(\vec{b_i}))$ is the concatenated vector $(\sigma(a_{i1}) \parallel \dots \parallel \sigma(a_{id}))$

More generally: A module lattice is $\mathcal{O}(M)$

for $M \subseteq K^r$ an R -module

↳ pseudo-basis instead of basis

For us today: a module lattice is $\mathcal{O}(M)$

where $M = \left\{ \sum_{i=1}^r x_i b_i \mid x_i \in R \right\}$

with $b_i \in K^r$ linearly indep.

* $(\vec{b}_1, \dots, \vec{b}_r)$ basis of M

* r rank of M

↳ basis of the module:

$$\begin{bmatrix} | & & | \\ \sigma(a_1) & \cdots & \sigma(a_n x) \\ | & & | \\ \vdots & & \vdots \\ | & \cdots & | \\ \sigma(a_1) & \cdots & \sigma(a_n x^{d-1}) \\ | & & | \\ | & \cdots & | \\ \sigma(a_1) & \cdots & \sigma(a_n x^d) \end{bmatrix}$$

Embeddings Here!
(next page)

Why do we care about module lattices?

RLWE, RSIS, Module-LWE, Module-SIS

are equivalent to SIVP in module lattices.
(which is no harder than SVP)

(And NTRU is no harder than SVP in module lattices)

NIST: 11 of the 12 lattice submissions

use NTRU, RLWE, RSIS, ...

↳ the modules involved have rank $\approx 3, 4$, always ≤ 10

⇒ Solving SVP in modules of small rank (≤ 10) has an impact on these constructions.

Embeddings: $\sigma: K \rightarrow \mathbb{R}^d$

coefficient embedding

$$a_0 + a_1x + \dots + a_{d-1}x^{d-1} \mapsto (a_0, \dots, a_{d-1})$$

canonical embedding:

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

$$a(x) \mapsto (a(\alpha_1), \dots, a(\alpha_d))$$

with $\alpha_1, \dots, \alpha_d$ the complex roots
of P . ($K = \mathbb{Q}[x]/P$)

A module lattice is $\sigma(M)$ with
the same basis as before, but we can
change the embedding σ .

* Constructions usually use σ_{coeff}
(easier to handle: elements in \mathbb{Q} or even in \mathbb{Z})

* Cryptanalyse usually use $\sigma_{\text{canonical}}$
(nice algebraic properties: eg mult is coordinate-wise)

$\sigma_{\text{coeff}}(M)$ is usually not the same lattice as $\sigma_{\text{canonical}}$

But: * for power-of-2 cyclo \rightarrow same lattice (up to rotation and scaling)

* NTRU Prime and other "nice" fields
(that we want to use) \rightarrow similar geometry
(not exactly the same, but ok).

From now on: $\sigma = \sigma_{\text{canonical}}$

A module is a lattice over R

module: $M = \{ \sum x_i \vec{b}_i \mid x_i \in R \}$

$\vec{b}_i \in K$
 \vec{b}_i linearly indep

$$= \boxed{\vec{b}_1 \dots \vec{b}_r} \times R^r$$

lattice: $L = \{ \sum x_i \vec{b}_i \mid x_i \in \mathbb{Z} \}$

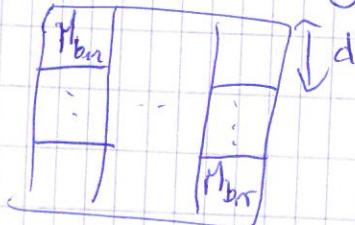
$$= \boxed{\vec{b}_1 \dots \vec{b}_n} \mathbb{Z}^n$$

$\vec{b}_i \in IR^r$
linearly indep

A module is both a "lattice" of rank r over R

$$\boxed{\vec{b}_1 \dots \vec{b}_r} R^r$$

• Lattice of rank d_r over \mathbb{Z}

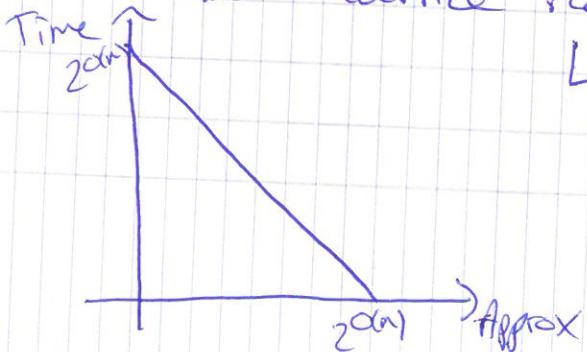


in practice (NIST): $r \approx 3, 4 \quad (\leq 10)$

$d \approx 512, 1024, 256$

s.t. $\frac{s}{500} \leq d_r \leq 1000$

Remember lattice reduction



↳ This works for lattices over \mathbb{Z} . What about lattices over R ?

Objective : Adapt LLL to lattices over \mathbb{R} .

Why LLL not BKZ?

* easier \rightarrow this is the true reason

* sufficient for modules of rank 3, 4 ...

History : Nopas '96 : specific number fields
no bound

	bound on quality	bound on run time
Nopas '96 specific number fields	X	X
Fieker-Schost '96 All number fields totally real fields	X X	X ✓
Kim-Lee '17 norm Euclidean fields biquadratic fields	X ✓	✓ ✓
This work any field	✓	$\approx (\checkmark \text{ if we have an oracle solving CVP in a fixed lattice})$

II LLL in dimension 2 = Lagrange-Gauss algorithm

Over \mathbb{Z} : animation on slides

1) What is $\|\cdot\|$ over \mathbb{R} ?

$$\cdot a \in \mathbb{R} \quad \text{def: } \|a\|_2 = \|\sigma(a)\|_2$$

$$\vec{b} \in \mathbb{R}^r \quad \|\vec{b}\|_2 = \|\sigma(b_1)\|_2 \cdots \|\sigma(b_r)\|_2 \\ = \sqrt{\sum_i \|\sigma(b_i)\|^2} \quad (\text{Pythagore})$$

This is what we want: small in $\mathbb{R}^r \Rightarrow$ small in \mathbb{Z}^{rd}

Two notions of number theory: $\text{Tr}(a) = \sum_i \sigma(a)_i$;

$$(\underbrace{\mathcal{N}(a) = \prod_i \sigma(a)_i}_{\text{well defined}})$$

~~*Assume \mathcal{N} is well defined (e.g. $R = \mathbb{Z}[x]/(x^n + 1)$)~~

Define $\bar{a} = (\overline{\sigma(a)}_1, \dots, \overline{\sigma(a)}_d)$ and assume $\bar{a} \in \mathbb{R}^d$ if $a \in R$
 (for example $R = \mathbb{Z}[x]/(x^n + 1)$)

$$\text{Then } \|a\|_2 = \sqrt{\sum_i |\sigma(a)_i|^2} = \sqrt{\text{Tr}(a\bar{a})}$$

$$\text{and } \|\vec{b}\|_2 = \sqrt{\sum_i \text{Tr}(b_i \bar{b}_i)} = \sqrt{\text{Tr}(\sum_i b_i \bar{b}_i)}$$

\hookrightarrow define a "scalar product"

$$\langle \vec{b}, \vec{c} \rangle = \sum_i b_i \bar{c}_i \in R(\text{or } K)$$

2) QR factorisation over \mathbb{R}^2

For \mathbb{R}^2

$$b_1^* = b_1$$

$$b_2^* = b_2 - \underbrace{\frac{\langle b_2, b_1^* \rangle}{\|b_1^*\|^2}}_{\in \mathbb{K}} b_1^* \in \mathbb{K}^2$$

$$\hookrightarrow \langle b_2^*, b_1^* \rangle = 0$$

$$Q = \begin{pmatrix} b_1^* & b_2^* \\ \|b_1^*\| & \|b_2^*\| \end{pmatrix}$$

$$R = \begin{pmatrix} \|b_1^*\| \frac{\langle b_2, b_1^* \rangle}{\|b_1^*\|^2} & 0 \\ 0 & \|b_2^*\| \end{pmatrix}$$

$$\|b_i^*\| = \sqrt{\langle b_i^*, b_i^* \rangle}$$

Qu: $\sqrt{\cdot}$ in \mathbb{K} ? \rightarrow can avoid $\sqrt{\cdot}$ by using "Gram-Schmidt"

\hookrightarrow div of $\frac{\langle b_2, b_1^* \rangle}{\|b_1^*\|^2}$ by $\|b_1^*\|$

\hookrightarrow div of $\langle b_2, b_1^* \rangle$ by $\|b_1^*\|^2$

\downarrow or, we can define K_R and $\sqrt{\cdot}$ is well defined

Properties: * $QR = \underbrace{(b_1 \ b_2)}_B$

* R triangular

$$* Q \bar{Q}^T = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

* $R \begin{pmatrix} u \\ v \end{pmatrix}$ short $\Leftrightarrow B \begin{pmatrix} u \\ v \end{pmatrix}$ short

(preserves geometry)

$$\hookrightarrow \langle B \begin{pmatrix} u \\ v \end{pmatrix}, B \begin{pmatrix} u \\ v \end{pmatrix} \rangle$$

$$= (\bar{u} \ \bar{v}) \bar{B}^T B \begin{pmatrix} u \\ v \end{pmatrix}$$

$$= (\bar{u} \ \bar{v}) \bar{R}^T \underbrace{\bar{Q}^T Q}_I R \begin{pmatrix} u \\ v \end{pmatrix} = \langle R \begin{pmatrix} u \\ v \end{pmatrix}, R \begin{pmatrix} u \\ v \end{pmatrix} \rangle$$

and $\| \cdot \|$ follows because $\sqrt{\text{Tr}(\langle \cdot, \cdot \rangle)}$

↳ QR factorisation is OK, we can assume that our basis is triangular

3) Euclidean division

$a, b \in R$ (or K)

we would like $r \in R$ s.t. $\|a + rb\| \leq \frac{1}{2} \|b\|$

Pb: Most of the number fields we are interested in are not euclidean \rightarrow no such r .

But: There should exist (counting argument) $u, v \in R$

s.t. * $\|au + bv\| \leq \frac{1}{2} \|b\|$

* $\|bu\| \leq \text{poly}(d)$

(here, R is "nice")
small basis (small volume)
 Δ^d in general
or even $\lambda_n(R)$

↳ we can use it to reduce a small multiple of \vec{b}_2 by \vec{b}_1 .

Rq: over \mathbb{Z} : when we have $\begin{pmatrix} r_{11} & r_{12} \\ 0 & r_{22} \end{pmatrix}$
and $r_{11} \gg r_{22}$, we are roughly done
because any vector is $\geq \min(r_{11}, r_{22}) \approx r_{11}$

↳ so $\begin{pmatrix} r_{11} \\ 0 \end{pmatrix}$ is ~~not~~ a small vector.

→ Same is true over R .

The case where we make progress is

$r_{22} \ll r_{11}$. And then, even if we compute

$$\|u\left(\frac{r_{12}}{r_{22}}\right) + v\left(\frac{r_{11}}{2}\right)\|^2 \text{ with } u \text{ not too large}$$

$$L_2 = \underbrace{u^2 r_{22}^2}_{\ll r_{11}^2} + \underbrace{(ur_{12} + vr_{11})^2}_{\leq \left(\frac{r_{11}}{2}\right)^2} \leq \frac{r_{11}^2}{4}$$

\hookrightarrow we make progress

\hookrightarrow again, same is true over R

3.2) How to compute it?

[input]: $a, b \in \mathbb{R}$

[output]: $u, v \in \mathbb{R}$ s.t. $\|au + bv\| \leq \frac{1}{2} \|b\|$

$$\|u\| \leq \text{poly}(d)$$

1. $\|au + bv\| \leq \frac{1}{2} \|b\|$ means au should be close to $-bv$
(or to bv , if we change the sign of v)

2. au and bv : products \rightarrow take the Log to have sums

def Log: $\mathbb{R} \rightarrow \mathbb{R}^d$

$$x \mapsto (\log(|\sigma(x)|);$$

typical in
number theory

pb: $\text{Log}(1+i) = \text{Log}(11)$ but i not close to 1



so $\|\text{Log}(au) - \text{Log}(bv)\|$ small

$\Rightarrow \|au - bv\|$ small

\rightarrow use $\overline{\text{Log}}: \mathbb{R} \xrightarrow{\text{ed}} \mathbb{R}^d \times (\mathbb{R}/2\pi)^d$

$$x \mapsto (\log|\sigma_i(x)|, \Theta(\sigma_i(x))) \iff$$

$$\Theta(re^{it}) = t$$

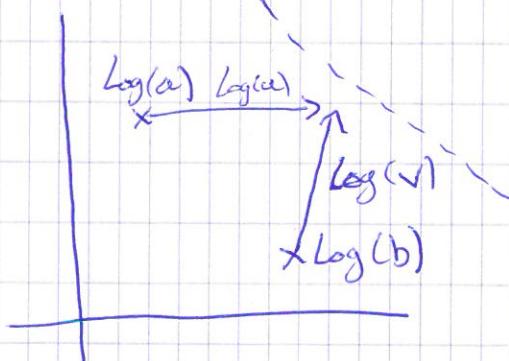
Now: if $\|\overline{\text{Log}}(x) - \overline{\text{Log}}(y)\| \leq \varepsilon$ ~~then~~ $+ \varepsilon \leq \frac{1}{2}$

then $\|x - y\|_\infty \leq 4\varepsilon \min(\|x\|_\infty, \|y\|_\infty)$

$$(\|x - y\| \leq \varepsilon \min(\|x\|, \|y\|))$$

\hookrightarrow obj: $\|\overline{\text{Log}}(ua) - \overline{\text{Log}}(vb)\| \leq \frac{1}{\log(d)} + \|\overline{\text{Log}}(u)\|, \|\overline{\text{Log}}(v)\| \leq O(\log(d))$

Picture (with only Log) just for the idea
(maths after)

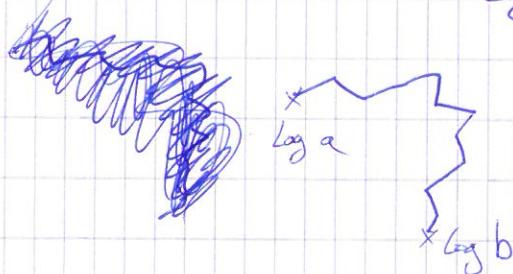


$$\overline{\text{Log}}(av) = \overline{\text{Log}}(a) + \overline{\text{Log}}(v)$$

obj: $\overline{\text{Log}}(u/v)$ close to $\overline{\text{Log}}(b/a)$

\hookrightarrow This is a CVP over a set which is not a lattice

\hookrightarrow make it a lattice: consider a factor basis g_1, \dots, g_k and look for u, v as a product of g_i 's



$$B = \begin{bmatrix} \overline{\text{Log}}(g_1) & \dots & \overline{\text{Log}}(g_k) \\ \vdots & & \vdots \\ 1 & & 1 \end{bmatrix}$$

$$L = L(B)$$

$$r = \begin{bmatrix} \overline{\text{Log}}(b/a) \\ \vdots \\ 0 \end{bmatrix}$$

\rightarrow add a block of 2π
+ a block of units

Algo: * compute h and t

* solve CVP in L with $t \rightarrow$ output s

* write $s = \sum \log(u_i v_i)$

* output u, v

why does it work? How close is s to t ?

(correctness of Algo) (i.e. How close $\log(u)$ to $\log(v)$)

\hookrightarrow depends on the density of L .

$\text{Vol}(L)$ is fixed \rightarrow increasing dim shorten vectors

\rightarrow with a heuristic counting argument, we believe that $k = O(d^2)$ is enough for L to be dense enough.

Run time of Algo: everything poly except "solve CVP in L "

\hookrightarrow assume oracle and everything becomes poly-time

Now, we have the 2 key ingredients: QR and Euclidean div \rightarrow we can do LLL in \mathbb{R}^2 .

\hookrightarrow to prove that LLL terminates in poly time + output is small we need an extra ingredient: $N(\cdot)$

III)

LLL in dim r

Very similar to dim 2:

QR

$$\begin{bmatrix} \text{row } i \\ \vdots \\ \text{row } r \\ \text{row } r+1 \\ \vdots \\ \text{row } m \end{bmatrix}$$

→ choose i s.t.

$$\begin{bmatrix} \text{row } i \\ \vdots \\ \text{row } m \\ \text{row } r+1 \\ \vdots \\ \text{row } r \end{bmatrix}$$

can be improved

and do the reduce
and SWAP.

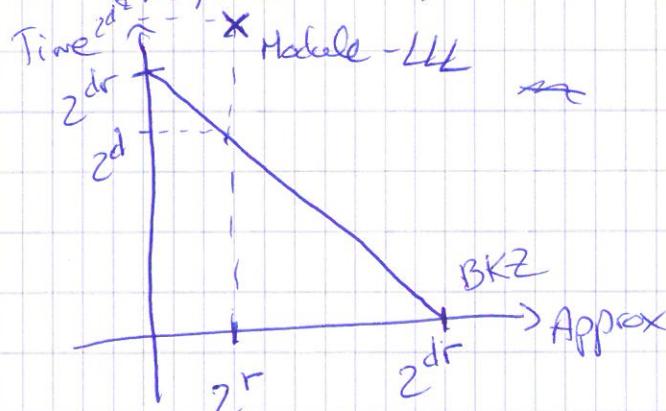
Result: Approx factor $\text{quasi-poly}(d) O(\frac{r}{k})$

Time $\text{poly}(d, k)$ if oracle

What if we actually want to run it?

need to instantiate the oracle → with generic algorithms

$\dim(L) = d^2 \rightarrow \text{CVP in } L \text{ can be done in } 2^{d^2}$



Don't use it in practice

Open problems:

* improve CVP in L?

↳ decrease its dim?

↳ use its structure?

* Generalize LLL to BKZ?

LLL: Leverage SVP in dim 2 $\rightarrow 2^n$ -SVP in dim n

BKZ: SVP in dim β $\rightarrow 2^{n/\beta}$ -SVP in dim n

The reduction should be easy

The hard part is how to solve SVP in dim β

(That was the hard part in LLL too)