

Ideal-SVP is Hard for Small-Norm Uniform Prime Ideals

Joël Felderhoff, Alice Pellet-Mary, Damien Stehlé and Benjamin Wesolowski

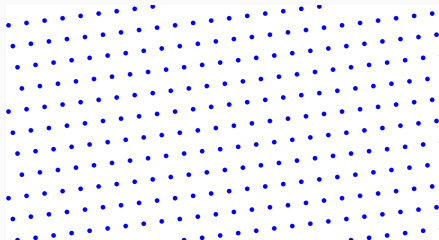
INRIA Lyon, ENS de Lyon

- New reduction: \mathcal{P}^{-1} -ideal-SVP to \mathcal{P} -ideal-SVP.
- Application: new distribution of NTRU instances with difficulty based on wc -ideal-SVP.

- New reduction: \mathcal{P}^{-1} -ideal-SVP to \mathcal{P} -ideal-SVP.
- Application: new distribution of NTRU instances with difficulty based on wc -ideal-SVP.

To appear in the proceedings of **TCC 2023**. Available at:
<https://eprint.iacr.org/2023/1370>

Definitions



A 2-dimensional lattice

Definition

For $\mathbf{b}_1, \dots, \mathbf{b}_n \in \mathbb{Z}^n$ linearly independent, the lattice spanned by the basis $\mathbf{b}_1, \dots, \mathbf{b}_n$ is $\mathcal{L} = \sum_i \mathbb{Z} \cdot \mathbf{b}_i \subset \mathbb{R}^n$.

It is discrete and has a shortest non-zero vector.

Finding any short non-zero vector in \mathcal{L} given $(\mathbf{b}_i)_i$ is hard in general.

Number fields and ideals

$K = \mathbb{Q}[X]/(X^n + 1)$, $\mathcal{O}_K = \mathbb{Z}[X]/(X^n + 1)$ for $n = 2^r$
(K a number field, \mathcal{O}_K its ring of integers).

The size of an element $a \in K$ is $\|a\| = \left(\sum_i |a_i|^2\right)^{1/2}$.

The size of an element is the ℓ_2 -norm of its Minkowski embedding.

$K = \mathbb{Q}[X]/(X^n + 1)$, $\mathcal{O}_K = \mathbb{Z}[X]/(X^n + 1)$ for $n = 2^r$
(K a number field, \mathcal{O}_K its ring of integers).

The size of an element $a \in K$ is $\|a\| = \left(\sum_i |a_i|^2\right)^{1/2}$.

The size of an element is the ℓ_2 -norm of its Minkowski embedding.

Definition (Ideal)

A set $\mathfrak{a} \subseteq K$ is an ideal if it is discrete, stable by addition and by multiplication by any element of \mathcal{O}_K . It is then a lattice.

Norm of an ideal: $\mathcal{N}(I) = \text{Vol}(I) / \text{Vol}(\mathcal{O}_K) \in \mathbb{Z}$.

Let $\mathfrak{a}, \mathfrak{b}$ ideals of K , and $a \in K$.

Principal ideal

$$(a) = \{x \cdot a, x \in \mathcal{O}_K\}.$$

Multiplication and inverse

$$\mathfrak{a} \cdot \mathfrak{b} = \{\sum_i a_i \cdot b_i\}, \mathfrak{a}^{-1} = \{x \in K, x \cdot \mathfrak{a} \subseteq \mathcal{O}_K\}.$$

We have that $\mathfrak{a} \cdot \mathfrak{a}^{-1} = \mathcal{O}_K$.

Prime ideals

An ideal \mathfrak{p} is prime ($\mathfrak{p} \in \mathcal{P}$) if

$$\mathfrak{p} = \mathfrak{a} \cdot \mathfrak{b} \Rightarrow \mathfrak{a} = \mathcal{O}_K \text{ or } \mathfrak{b} = \mathcal{O}_K$$

The problem ideal-HSVP

Definition (ideal-HSVP $_{\gamma}$)

Given an ideal $\mathfrak{a} \subseteq K$, find $x \in \mathfrak{a} \setminus \{0\}$ with $\|x\| \leq \gamma \cdot \mathcal{N}(\mathfrak{a})^{1/d}$.

Ideal lattices are **not typical lattices**. E.g., they verify $\lambda_1(I) \approx \lambda_d(I)$.

¹[CDPR16, CDW17, PHS19]

Definition (ideal-HSVP $_{\gamma}$)

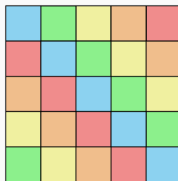
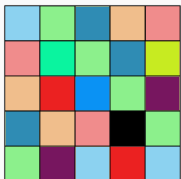
Given an ideal $\mathfrak{a} \subseteq K$, find $x \in \mathfrak{a} \setminus \{0\}$ with $\|x\| \leq \gamma \cdot \mathcal{N}(\mathfrak{a})^{1/d}$.

Ideal lattices are **not typical lattices**. E.g., they verify $\lambda_1(I) \approx \lambda_d(I)$.

- There are specific attacks on ideal lattices¹.
- Ideals are the simplest examples of module lattices (KYBER, DILITHIUM).
- ideal-HSVP is related to other structured lattice problems (Module-SVP, NTRU, RingLWE).

¹[CDPR16, CDW17, PHS19]

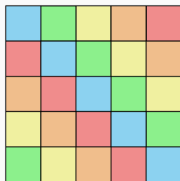
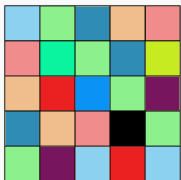
Why small ideal lattices?



Typical lattice basis: $O(d^2)$ integers vs ideal lattice basis: $O(d)$ integers.²

²Images from [Qua14]

Why small ideal lattices?



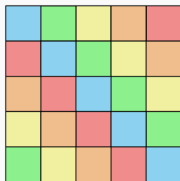
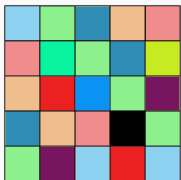
Typical lattice basis: $O(d^2)$ integers vs **ideal lattice basis:** $O(d)$ integers.²

Bitsize of a typical element of \mathfrak{a} is $\log(\mathcal{N}(\mathfrak{a}))$.

→ We want $\mathcal{N}(\mathfrak{a}) \approx \text{poly}(d)^d$ in order to have small keys.

²Images from [Qua14]

Why small ideal lattices?



Typical lattice basis: $O(d^2)$ integers vs **ideal lattice basis:** $O(d)$ integers.²

Bitsize of a typical element of \mathfrak{a} is $\log(\mathcal{N}(\mathfrak{a}))$.

→ We want $\mathcal{N}(\mathfrak{a}) \approx \text{poly}(d)^d$ in order to have small keys.

Also: faster algorithms.

²Images from [Qua14]

Average-case to average-case reduction

Worst-case: Solve \mathcal{P} for **all** instance of \mathcal{P} (for the worst instance).

Average-case for D : Solve \mathcal{P} for $I \leftarrow D$ **with non-negligible probability**.

For cryptography, we are interested in **Average-case** hardness.

Average-case to average-case reduction

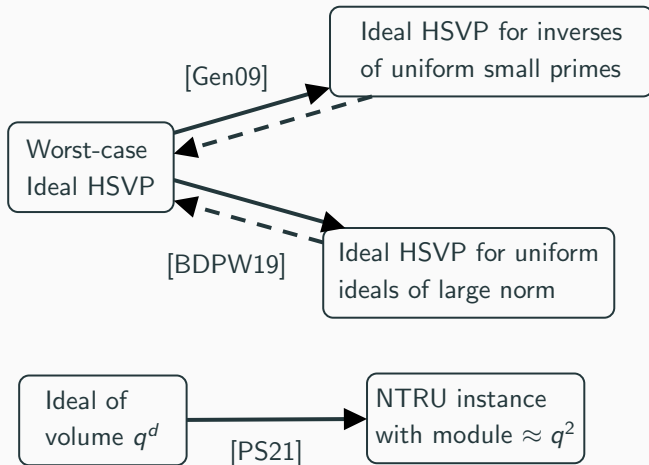
Worst-case: Solve \mathcal{P} for **all** instance of \mathcal{P} (for the worst instance).

Average-case for D : Solve \mathcal{P} for $I \leftarrow D$ **with non-negligible probability**.

For cryptography, we are interested in **Average-case** hardness.

Here we show an Average-case to Average-case reduction.

Prior works on ideal-HSVP



Random version of ideal-HSVP

\mathcal{W} -ideal-HSVP: solving ideal-HSVP for a uniform element of \mathcal{W} .

Note: there are sets \mathcal{W} such that \mathcal{W} -ideal-HSVP is easy [BGP22].

Random version of ideal-HSVP

\mathcal{W} -ideal-HSVP: solving ideal-HSVP for a uniform element of \mathcal{W} .

Note: there are sets \mathcal{W} such that \mathcal{W} -ideal-HSVP is easy [BGP22].

We show that \mathcal{P}^{-1} -ideal-HSVP reduces to \mathcal{P} -ideal-HSVP.

Two reasons

1. [Gen09]: ideal-HSVP (for all ideals) reduces to \mathcal{P}^{-1} -ideal-HSVP.
2. The NTRU reduction from [PS21] works for integral ideals.

Sampling ideals

Algorithm 2.1 ArakelovSampling algorithm

Output: An ideal \mathfrak{b}

- 1: Let \mathfrak{q} a uniform small prime ideal.
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q}$.
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_\infty(r) \cap I)$
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$
-

Algorithm 2.1 ArakelovSampling algorithm

Output: An ideal \mathfrak{b}

- 1: Let \mathfrak{q} a uniform small prime ideal.
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q}$.
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_\infty(r) \cap I)$
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$
-

Algorithm 2.1 ArakelovSampling algorithm

Output: An ideal \mathfrak{b}

- 1: Let \mathfrak{q} a uniform small prime ideal.
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q}$.
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_\infty(r) \cap I)$
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$
-

Algorithm 2.1 ArakelovSampling algorithm

Output: An ideal \mathfrak{b}

- 1: Let \mathfrak{q} a uniform small prime ideal.
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q}$.
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_\infty(r) \cap I)$
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$
-

Algorithm 2.1 ArakelovSampling algorithm

Output: An ideal \mathfrak{b}

- 1: Let \mathfrak{q} a uniform small prime ideal.
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q}$.
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_\infty(r) \cap I)$
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$
-

Algorithm 2.1 ArakelovSampling algorithm

Output: An ideal \mathfrak{b}

- 1: Let \mathfrak{q} a uniform small prime ideal.
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q}$.
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_\infty(r) \cap I)$
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$
-

Algorithm 2.1 ArakelovSampling algorithm

Output: An ideal \mathfrak{b}

- 1: Let \mathfrak{q} a uniform small prime ideal.
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q}$.
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_\infty(r) \cap I)$
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$
-

Outputs uniform integral ideals of norm $\approx r^d$ for $r = 2^{O(d)}$.
 \Rightarrow Too big for our use-cases!

What if we want to sample with a trapdoor inside?

We modify our algorithm to output some small element in \mathfrak{b}^{-1} .

Algorithm 2.2 ArakelovSampling' algorithm

Output: An ideal \mathfrak{b} and $y \in \mathfrak{b}^{-1}$.

- 1: Let \mathfrak{q} an uniform small prime ideal.
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q}$
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_\infty(r) \cap I)$.
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$
-

What if we want to sample with a trapdoor inside?

We modify our algorithm to output some small element in \mathfrak{b}^{-1} .

Algorithm 2.2 ArakelovSampling' algorithm

Output: An ideal \mathfrak{b} and $y \in \mathfrak{b}^{-1}$.

- 1: Let \mathfrak{q} an uniform small prime ideal.
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q}$
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_\infty(r) \cap I)$.
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$
-

What if we want to sample with a trapdoor inside?

We modify our algorithm to output some small element in \mathfrak{b}^{-1} .

Algorithm 2.2 ArakelovSampling' algorithm

Output: An ideal \mathfrak{b} and $y \in \mathfrak{b}^{-1}$.

- 1: Let $(\mathfrak{q}, \nu_{\mathfrak{q}}) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q}$
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_{\infty}(r) \cap I)$.
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$
-

What if we want to sample with a trapdoor inside?

We modify our algorithm to output some small element in \mathfrak{b}^{-1} .

Algorithm 2.2 ArakelovSampling' algorithm

Output: An ideal \mathfrak{b} and $y \in \mathfrak{b}^{-1}$.

- 1: Let $(\mathfrak{q}, v_{\mathfrak{q}}) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q}$
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_{\infty}(r) \cap I)$.
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$
-

What if we want to sample with a trapdoor inside?

We modify our algorithm to output some small element in \mathfrak{b}^{-1} .

Algorithm 2.2 ArakelovSampling' algorithm

Output: An ideal \mathfrak{b} and $y \in \mathfrak{b}^{-1}$.

- 1: Let $(q, v_q) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot q$
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_\infty(r) \cap I)$.
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$
-

What if we want to sample with a trapdoor inside?

We modify our algorithm to output some small element in \mathfrak{b}^{-1} .

Algorithm 2.2 ArakelovSampling' algorithm

Output: An ideal \mathfrak{b} and $y \in \mathfrak{b}^{-1}$.

- 1: Let $(\mathfrak{q}, \nu_{\mathfrak{q}}) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q}$ and $s_I = \exp(\zeta) \cdot u \cdot s_{\mathfrak{q}} \in I$.
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_{\infty}(r) \cap I)$.
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$
-

What if we want to sample with a trapdoor inside?

We modify our algorithm to output some small element in \mathfrak{b}^{-1} .

Algorithm 2.2 ArakelovSampling' algorithm

Output: An ideal \mathfrak{b} and $y \in \mathfrak{b}^{-1}$.

- 1: Let $(\mathfrak{q}, v_{\mathfrak{q}}) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q}$ and $s_I = \exp(\zeta) \cdot u \cdot s_{\mathfrak{q}} \in I$.
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_{\infty}(r) \cap I)$.
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$
-

What if we want to sample with a trapdoor inside?

We modify our algorithm to output some small element in \mathfrak{b}^{-1} .

Algorithm 2.2 ArakelovSampling' algorithm

Output: An ideal \mathfrak{b} and $y \in \mathfrak{b}^{-1}$.

- 1: Let $(\mathfrak{q}, v_{\mathfrak{q}}) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q}$ and $s_I = \exp(\zeta) \cdot u \cdot s_{\mathfrak{q}} \in I$.
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_{\infty}(r) \cap I)$.
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$
-

What if we want to sample with a trapdoor inside?

We modify our algorithm to output some small element in \mathfrak{b}^{-1} .

Algorithm 2.2 ArakelovSampling' algorithm

Output: An ideal \mathfrak{b} and $y \in \mathfrak{b}^{-1}$.

- 1: Let $(\mathfrak{q}, v_{\mathfrak{q}}) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q}$ and $s_I = \exp(\zeta) \cdot u \cdot s_{\mathfrak{q}} \in I$.
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_{\infty}(r) \cap I)$.
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$ and $y = x^{-1} \cdot s_I$.
-

What if we want to sample with a trapdoor inside?

We modify our algorithm to output some small element in \mathfrak{b}^{-1} .

Algorithm 2.2 ArakelovSampling' algorithm

Output: An ideal \mathfrak{b} and $y \in \mathfrak{b}^{-1}$.

- 1: Let $(q, v_q) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $I = \exp(\zeta) \cdot u \cdot q$ and $s_I = \exp(\zeta) \cdot u \cdot s_q \in I$.
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_\infty(r) \cap I)$.
 - 5: **Return** $\mathfrak{b} = x \cdot I^{-1}$ and $y = x^{-1} \cdot s_I$.
-

Drawback

The element $y = x^{-1} \cdot s_I$ can be very large compared to $\mathcal{N}(\mathfrak{b}^{-1})^{1/d}$.

What if we want to sample with a trapdoor inside?

We modify our algorithm to output some small element in \mathfrak{b}^{-1} .

Algorithm 2.2 ArakelovSampling' algorithm

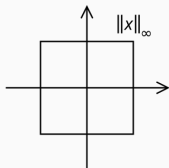
Output: An ideal \mathfrak{b} and $y \in \mathfrak{b}^{-1}$.

- 1: Let $(q, v_q) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)
 - 2: Sample a small continuous Gaussian ζ and a uniform rotation u .
 - 3: Let $l = \exp(\zeta) \cdot u \cdot q$ and $s_l = \exp(\zeta) \cdot u \cdot s_q \in l$.
 - 4: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_\infty(r) \cap l)$.
 - 5: **Return** $\mathfrak{b} = x \cdot l^{-1}$ and $y = x^{-1} \cdot s_l$.
-

Drawback

The element $y = x^{-1} \cdot s_l$ can be very large compared to $\mathcal{N}(\mathfrak{b}^{-1})^{1/d}$.
→ This happens if x is **unbalanced**

Some details on ArakelovSampling



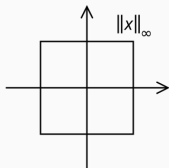
The set $\mathcal{B}_\infty(r)$

1. We pick \mathfrak{q} uniform prime.
2. We sample $x \leftarrow \mathcal{U}(\mathcal{B}_\infty(r) \cap \mathfrak{q})$.
3. We return $\mathfrak{b} = x \cdot \mathfrak{q}^{-1}$.

Sufficient conditions for uniform \mathfrak{b}

1. $|\mathcal{B}_\infty(r) \cap \mathfrak{q}|$ does not depend on \mathfrak{q} (too much).
2. $\text{Vol}(\text{Log}(\mathcal{B}_\infty(r)) \cap \{\sum x_i = t\})$ is \approx constant for $t \in [A, B]$.

Some details on ArakelovSampling



The set $\mathcal{B}_\infty(r)$

1. We pick \mathfrak{q} uniform prime.
2. We sample $x \leftarrow \mathcal{U}(\mathcal{B}_\infty(r) \cap \mathfrak{q})$.
3. We return $\mathfrak{b} = x \cdot \mathfrak{q}^{-1}$.

Sufficient conditions for uniform \mathfrak{b}

1. $|\mathcal{B}_\infty(r) \cap \mathfrak{q}|$ does not depend on \mathfrak{q} (too much).
2. $\text{Vol}(\text{Log}(\mathcal{B}_\infty(r)) \cap \{\sum x_i = t\})$ is \approx constant for $t \in [A, B]$.

Drawback

There are $x \in \mathcal{B}_\infty(r)$ with $\|x^{-1}\|$ very large.

Main contribution:

\mathcal{P}^{-1} -ideal-SVP **to** \mathcal{P} -ideal-SVP

First contribution: Generalized Arakelov ideal sampling

We generalize the approach of [BDPW20, Boe22]:

Algorithm 3.1 $\text{SampleIdeal}_{\mathcal{B}_{A,B}}$ algorithm

Input: \mathfrak{a} an ideal, $s_{\mathfrak{a}} \in \mathfrak{a}$ small, $\mathcal{B}_{A,B} \subset K_{\mathbb{R}}$ a **well chosen** set.

Output: (\mathfrak{b}, y) such that $y \in (\mathfrak{b} \cdot \mathfrak{a})^{-1}$.

1: Let $(q, v_q) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)

2: Sample ζ and u .

3: Let $I = \exp(\zeta) \cdot u \cdot q \cdot \mathfrak{a}$

4: Let $s_I = \exp(\zeta) \cdot u \cdot s_q \cdot s_{\mathfrak{a}} \in I$.

5: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_{A,B} \cap I)$ using s_I .

6: **Return** $(\mathfrak{b} = x \cdot I^{-1}, y = x^{-1} \cdot s_I \cdot v_q)$

$\triangleright y \in (\mathfrak{b} \cdot \mathfrak{a})^{-1}$

First contribution: Generalized Arakelov ideal sampling

We generalize the approach of [BDPW20, Boe22]:

Algorithm 3.1 $\text{SampleIdeal}_{\mathcal{B}_{A,B}}$ algorithm

Input: \mathfrak{a} an ideal, $s_{\mathfrak{a}} \in \mathfrak{a}$ small, $\mathcal{B}_{A,B} \subset K_{\mathbb{R}}$ a **well chosen** set.

Output: (\mathfrak{b}, y) such that $y \in (\mathfrak{b} \cdot \mathfrak{a})^{-1}$.

1: **Let** $(\mathfrak{q}, v_{\mathfrak{q}}) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)

2: Sample ζ and u .

3: Let $I = \exp(\zeta) \cdot u \cdot \mathfrak{q} \cdot \mathfrak{a}$

4: Let $s_I = \exp(\zeta) \cdot u \cdot s_{\mathfrak{q}} \cdot s_{\mathfrak{a}} \in I$.

5: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_{A,B} \cap I)$ using s_I .

6: **Return** $(\mathfrak{b} = x \cdot I^{-1}, y = x^{-1} \cdot s_I \cdot v_{\mathfrak{q}})$

$\triangleright y \in (\mathfrak{b} \cdot \mathfrak{a})^{-1}$

First contribution: Generalized Arakelov ideal sampling

We generalize the approach of [BDPW20, Boe22]:

Algorithm 3.1 $\text{SampleIdeal}_{\mathcal{B}_{A,B}}$ algorithm

Input: \mathfrak{a} an ideal, $s_{\mathfrak{a}} \in \mathfrak{a}$ small, $\mathcal{B}_{A,B} \subset K_{\mathbb{R}}$ a **well chosen** set.

Output: (\mathfrak{b}, y) such that $y \in (\mathfrak{b} \cdot \mathfrak{a})^{-1}$.

1: Let $(q, v_q) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)

2: **Sample** ζ and u .

3: Let $I = \exp(\zeta) \cdot u \cdot q \cdot \mathfrak{a}$

4: Let $s_I = \exp(\zeta) \cdot u \cdot s_q \cdot s_{\mathfrak{a}} \in I$.

5: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_{A,B} \cap I)$ using s_I .

6: **Return** $(\mathfrak{b} = x \cdot I^{-1}, y = x^{-1} \cdot s_I \cdot v_q)$

$\triangleright y \in (\mathfrak{b} \cdot \mathfrak{a})^{-1}$

First contribution: Generalized Arakelov ideal sampling

We generalize the approach of [BDPW20, Boe22]:

Algorithm 3.1 $\text{SampleIdeal}_{\mathcal{B}_{A,B}}$ algorithm

Input: \mathfrak{a} an ideal, $s_{\mathfrak{a}} \in \mathfrak{a}$ small, $\mathcal{B}_{A,B} \subset K_{\mathbb{R}}$ a **well chosen** set.

Output: (\mathfrak{b}, y) such that $y \in (\mathfrak{b} \cdot \mathfrak{a})^{-1}$.

1: Let $(q, v_q) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)

2: Sample ζ and u .

3: Let $I = \exp(\zeta) \cdot u \cdot q \cdot \mathfrak{a}$

4: Let $s_I = \exp(\zeta) \cdot u \cdot s_q \cdot s_{\mathfrak{a}} \in I$.

5: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_{A,B} \cap I)$ using s_I .

6: **Return** $(\mathfrak{b} = x \cdot I^{-1}, y = x^{-1} \cdot s_I \cdot v_q)$

$\triangleright y \in (\mathfrak{b} \cdot \mathfrak{a})^{-1}$

First contribution: Generalized Arakelov ideal sampling

We generalize the approach of [BDPW20, Boe22]:

Algorithm 3.1 $\text{SampleIdeal}_{\mathcal{B}_{A,B}}$ algorithm

Input: \mathfrak{a} an ideal, $s_{\mathfrak{a}} \in \mathfrak{a}$ small, $\mathcal{B}_{A,B} \subset K_{\mathbb{R}}$ a **well chosen** set.

Output: (\mathfrak{b}, y) such that $y \in (\mathfrak{b} \cdot \mathfrak{a})^{-1}$.

1: Let $(q, v_q) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)

2: Sample ζ and u .

3: Let $I = \exp(\zeta) \cdot u \cdot q \cdot \mathfrak{a}$

4: Let $s_I = \exp(\zeta) \cdot u \cdot s_q \cdot s_{\mathfrak{a}} \in I$.

5: **Sample** $x \leftarrow \mathcal{U}(\mathcal{B}_{A,B} \cap I)$ using s_I .

6: **Return** $(\mathfrak{b} = x \cdot I^{-1}, y = x^{-1} \cdot s_I \cdot v_q)$

$\triangleright y \in (\mathfrak{b} \cdot \mathfrak{a})^{-1}$

First contribution: Generalized Arakelov ideal sampling

We generalize the approach of [BDPW20, Boe22]:

Algorithm 3.1 $\text{SampleIdeal}_{\mathcal{B}_{A,B}}$ algorithm

Input: \mathfrak{a} an ideal, $s_{\mathfrak{a}} \in \mathfrak{a}$ small, $\mathcal{B}_{A,B} \subset K_{\mathbb{R}}$ a **well chosen** set.

Output: (\mathfrak{b}, y) such that $y \in (\mathfrak{b} \cdot \mathfrak{a})^{-1}$.

1: Let $(q, v_q) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)

2: Sample ζ and u .

3: Let $I = \exp(\zeta) \cdot u \cdot q \cdot \mathfrak{a}$

4: Let $s_I = \exp(\zeta) \cdot u \cdot s_q \cdot s_{\mathfrak{a}} \in I$.

5: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_{A,B} \cap I)$ using s_I .

6: **Return** $(\mathfrak{b} = x \cdot I^{-1}, y = x^{-1} \cdot s_I \cdot v_q)$

$\triangleright y \in (\mathfrak{b} \cdot \mathfrak{a})^{-1}$

First contribution: Generalized Arakelov ideal sampling

We generalize the approach of [BDPW20, Boe22]:

Algorithm 3.1 $\text{SampleIdeal}_{\mathcal{B}_{A,B}}$ algorithm

Input: \mathfrak{a} an ideal, $s_{\mathfrak{a}} \in \mathfrak{a}$ small, $\mathcal{B}_{A,B} \subset K_{\mathbb{R}}$ a **well chosen** set.

Output: (\mathfrak{b}, y) such that $y \in (\mathfrak{b} \cdot \mathfrak{a})^{-1}$.

1: Let $(q, v_q) \leftarrow \text{SampleWithTrap}(\cdot)$. (Quantum)

2: Sample ζ and u .

3: Let $I = \exp(\zeta) \cdot u \cdot q \cdot \mathfrak{a}$

4: Let $s_I = \exp(\zeta) \cdot u \cdot s_q \cdot s_{\mathfrak{a}} \in I$.

5: Sample $x \leftarrow \mathcal{U}(\mathcal{B}_{A,B} \cap I)$ using s_I .

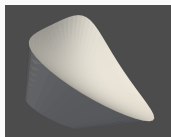
6: **Return** $(\mathfrak{b} = x \cdot I^{-1}, y = x^{-1} \cdot s_I \cdot v_q)$ $\triangleright y \in (\mathfrak{b} \cdot \mathfrak{a})^{-1}$

Theorem

Let $(\mathfrak{b}, y) = \text{SampleIdeal}_{\mathcal{B}_{A,B}}(\mathfrak{a}, s_{\mathfrak{a}}, A, B)$.

If $\mathcal{B}_{A,B}$ is **well chosen** then \mathfrak{b} is almost uniform in $\mathcal{I}_{A,B}$ and y is small.

What does “well chosen” mean?



1. $|\mathcal{B}_{A,B} \cap \mathfrak{a}|$ does not depend on \mathfrak{a} (too much).
2. $\text{Vol}(\text{Log}(\mathcal{B}_{A,B}) \cap \{\sum x_i = t\})$ is constant for $t \in [A, B]$.
3. Its elements must be balanced.

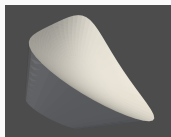
Balanced elements (for Minkowski embedding)

$x \in K$ is balanced if for all i ,

$$\frac{1}{\eta} \leq \frac{x_i}{\prod_j x_j^{1/d}} \leq \eta.$$

This is the same as $x \approx \mathcal{N}(x)^{1/d} \cdot (1, \dots, 1)$.

What does “well chosen” mean?



1. $|\mathcal{B}_{A,B} \cap \mathfrak{a}|$ does not depend on \mathfrak{a} (too much).
2. $\text{Vol}(\text{Log}(\mathcal{B}_{A,B}) \cap \{\sum x_i = t\})$ is constant for $t \in [A, B]$.
3. Its elements must be balanced.

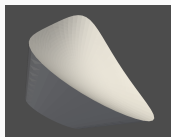
Balanced elements (for Minkowski embedding)

$x \in K$ is balanced if for all i ,

$$\frac{1}{\eta} \leq \frac{x_i}{\prod_j x_j^{1/d}} \leq \eta.$$

This is the same as $x \approx \mathcal{N}(x)^{1/d} \cdot (1, \dots, 1)$.

What does “well chosen” mean?



1. $|\mathcal{B}_{A,B} \cap \mathfrak{a}|$ does not depend on \mathfrak{a} (too much).
2. $\text{Vol}(\text{Log}(\mathcal{B}_{A,B}) \cap \{\sum x_i = t\})$ is constant for $t \in [A, B]$.
3. Its elements must be balanced.

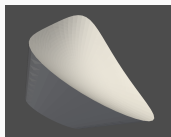
Balanced elements (for Minkowski embedding)

$x \in K$ is balanced if for all i ,

$$\frac{1}{\eta} \leq \frac{x_i}{\prod_j x_j^{1/d}} \leq \eta.$$

This is the same as $x \approx \mathcal{N}(x)^{1/d} \cdot (1, \dots, 1)$.

What does “well chosen” mean?



1. $|\mathcal{B}_{A,B} \cap \mathfrak{a}|$ does not depend on \mathfrak{a} (too much).
2. $\text{Vol}(\text{Log}(\mathcal{B}_{A,B}) \cap \{\sum x_i = t\})$ is constant for $t \in [A, B]$.
3. Its elements must be balanced.

Balanced elements (for Minkowski embedding)

$x \in K$ is balanced if for all i ,

$$\frac{1}{\eta} \leq \frac{x_i}{\prod_j x_j^{1/d}} \leq \eta.$$

This is the same as $x \approx \mathcal{N}(x)^{1/d} \cdot (1, \dots, 1)$.

In [BDPW20]: $\mathcal{B}_\infty(r)$: verifies items 1 and 2 but not 3!

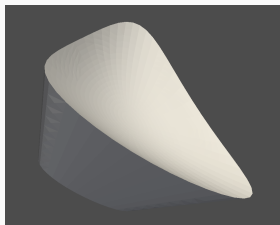
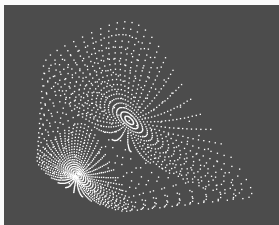
Reminder: conditions for being **well chosen**:

1. $|\mathcal{B}_{A,B} \cap \mathfrak{a}|$ does not depend on \mathfrak{a} (too much).
2. $\text{Vol}(\text{Log}(\mathcal{B}_{A,B}) \cap \{\sum x_i = t\})$ is constant for $t \in [A, B]$.
3. Its elements are balanced.

Our shape

Reminder: conditions for being **well chosen**:

1. $|\mathcal{B}_{A,B} \cap \mathfrak{a}|$ does not depend on \mathfrak{a} (too much).
2. $\text{Vol}(\text{Log}(\mathcal{B}_{A,B}) \cap \{\sum x_i = t\})$ is constant for $t \in [A, B]$.
3. Its elements are balanced.

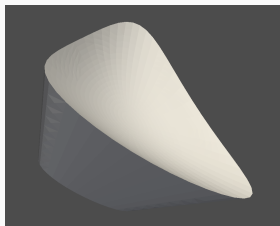
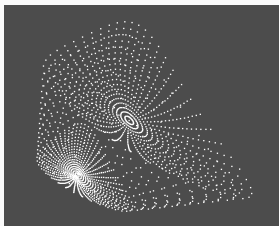


$$\mathcal{B}_{A,B}^\eta = \left\{ x \in K_{\mathbb{R}}, |\mathcal{N}(x)| \in [A, B], \left\| \text{Log} \left(\frac{x}{\mathcal{N}(x)^{1/d}} \right) \right\|_2 \leq \log(\eta) \right\}$$

Our shape

Reminder: conditions for being **well chosen**:

1. $|\mathcal{B}_{A,B} \cap \mathfrak{a}|$ does not depend on \mathfrak{a} (too much).
2. $\text{Vol}(\text{Log}(\mathcal{B}_{A,B}) \cap \{\sum x_i = t\})$ is constant for $t \in [A, B]$.
3. Its elements are balanced.

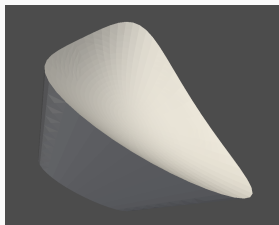
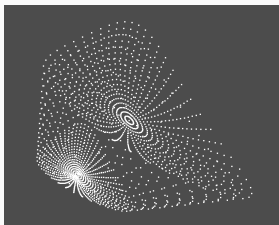


$$\mathcal{B}_{A,B}^\eta = \left\{ x \in K_{\mathbb{R}}, \quad |\mathcal{N}(x)| \in [A, B], \quad \left\| \text{Log} \left(\frac{x}{\mathcal{N}(x)^{1/d}} \right) \right\|_2 \leq \log(\eta) \right\}$$

Our shape

Reminder: conditions for being **well chosen**:

1. $|\mathcal{B}_{A,B} \cap \mathfrak{a}|$ does not depend on \mathfrak{a} (too much).
2. $\text{Vol}(\text{Log}(\mathcal{B}_{A,B}) \cap \{\sum x_i = t\})$ is constant for $t \in [A, B]$.
3. Its elements are **balanced**.

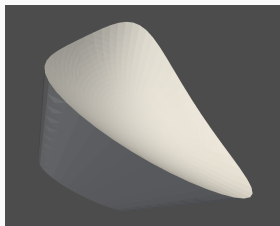
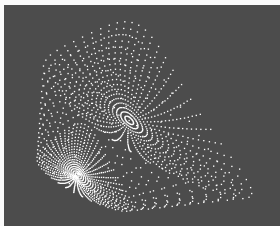


$$\mathcal{B}_{A,B}^\eta = \left\{ x \in K_{\mathbb{R}}, |\mathcal{N}(x)| \in [A, B], \left\| \text{Log} \left(\frac{x}{\mathcal{N}(x)^{1/d}} \right) \right\|_2 \leq \log(\eta) \right\}$$

Our shape

Reminder: conditions for being **well chosen**:

1. $|\mathcal{B}_{A,B} \cap \mathfrak{a}|$ does not depend on \mathfrak{a} (too much).
2. $\text{Vol}(\text{Log}(\mathcal{B}_{A,B}) \cap \{\sum x_i = t\})$ is constant for $t \in [A, B]$.
3. Its elements are balanced.



$$\mathcal{B}_{A,B}^\eta = \left\{ x \in K_{\mathbb{R}}, |\mathcal{N}(x)| \in [A, B], \left\| \text{Log} \left(\frac{x}{\mathcal{N}(x)^{1/d}} \right) \right\|_2 \leq \log(\eta) \right\}$$

Our \mathcal{P}^{-1} -ideal-SVP to \mathcal{P} -ideal-SVP reduction

The oracle \mathcal{O} solves ideal-HSVP for \mathfrak{p} uniform prime of norm in $[A, B]$.

Input: An ideal $I = \mathfrak{p}^{-1}$ with \mathfrak{p} uniform prime of norm in $[A, B]$.

Output: $x \in \mathfrak{p}^{-1} \setminus \{0\}$ small.

- 1: Let $s_{\mathfrak{p}} = \mathcal{O}(\mathfrak{p})$.
 - 2: Let $(\mathfrak{b}, y) = \text{SampleIdeal}_{A,B}(\mathfrak{p}, s_{\mathfrak{p}})$. $\triangleright \|y\|$ *small*
 - 3: **if** \mathfrak{b} is not prime. **then**
 - 4: Fail.
 - 5: Let $s_{\mathfrak{b}} = \mathcal{O}(\mathfrak{b})$. $\triangleright \|s_{\mathfrak{b}}\|$ *small*
 - 6: **Return** $\underbrace{s_{\mathfrak{b}}}_{\in \mathfrak{b}} \cdot \underbrace{y}_{\in (\mathfrak{b} \cdot \mathfrak{p})^{-1}} \in \mathfrak{p}^{-1}$. $\triangleright \|y \cdot s_{\mathfrak{b}}\|$ *small*
-

Our \mathcal{P}^{-1} -ideal-SVP to \mathcal{P} -ideal-SVP reduction

The oracle \mathcal{O} solves ideal-HSVP for \mathfrak{p} uniform prime of norm in $[A, B]$.

Input: An ideal $I = \mathfrak{p}^{-1}$ with \mathfrak{p} **uniform** prime of norm in $[A, B]$.

Output: $x \in \mathfrak{p}^{-1} \setminus \{0\}$ small.

- 1: Let $s_{\mathfrak{p}} = \mathcal{O}(\mathfrak{p})$.
 - 2: Let $(\mathfrak{b}, y) = \text{SampleIdeal}_{A,B}(\mathfrak{p}, s_{\mathfrak{p}})$. $\triangleright \|y\|$ *small*
 - 3: **if** \mathfrak{b} is not prime. **then**
 - 4: Fail.
 - 5: Let $s_{\mathfrak{b}} = \mathcal{O}(\mathfrak{b})$. $\triangleright \|s_{\mathfrak{b}}\|$ *small*
 - 6: **Return** $\underbrace{s_{\mathfrak{b}}}_{\in \mathfrak{b}} \cdot \underbrace{y}_{\in (\mathfrak{b} \cdot \mathfrak{p})^{-1}} \in \mathfrak{p}^{-1}$. $\triangleright \|y \cdot s_{\mathfrak{b}}\|$ *small*
-

Our \mathcal{P}^{-1} -ideal-SVP to \mathcal{P} -ideal-SVP reduction

The oracle \mathcal{O} solves ideal-HSVP for \mathfrak{p} uniform prime of norm in $[A, B]$.

Input: An ideal $I = \mathfrak{p}^{-1}$ with \mathfrak{p} uniform prime of norm in $[A, B]$.

Output: $x \in \mathfrak{p}^{-1} \setminus \{0\}$ small.

1: Let $s_{\mathfrak{p}} = \mathcal{O}(\mathfrak{p})$.

2: Let $(\mathfrak{b}, y) = \text{SampleIdeal}_{A,B}(\mathfrak{p}, s_{\mathfrak{p}})$.

▷ $\|y\|$ small

3: **if** \mathfrak{b} is not prime. **then**

4: Fail.

5: Let $s_{\mathfrak{b}} = \mathcal{O}(\mathfrak{b})$.

▷ $\|s_{\mathfrak{b}}\|$ small

6: **Return** $\underbrace{s_{\mathfrak{b}}}_{\in \mathfrak{b}} \cdot \underbrace{y}_{\in (\mathfrak{b} \cdot \mathfrak{p})^{-1}} \in \mathfrak{p}^{-1}$.

▷ $\|y \cdot s_{\mathfrak{b}}\|$ small

Our \mathcal{P}^{-1} -ideal-SVP to \mathcal{P} -ideal-SVP reduction

The oracle \mathcal{O} solves ideal-HSVP for \mathfrak{p} uniform prime of norm in $[A, B]$.

Input: An ideal $I = \mathfrak{p}^{-1}$ with \mathfrak{p} uniform prime of norm in $[A, B]$.

Output: $x \in \mathfrak{p}^{-1} \setminus \{0\}$ small.

1: Let $s_{\mathfrak{p}} = \mathcal{O}(\mathfrak{p})$.

2: Let $(\mathfrak{b}, y) = \text{SampleIdeal}_{A,B}(\mathfrak{p}, s_{\mathfrak{p}})$.

▷ $\|y\|$ small

3: **if \mathfrak{b} is not prime. then**

4: **Fail.**

5: Let $s_{\mathfrak{b}} = \mathcal{O}(\mathfrak{b})$.

▷ $\|s_{\mathfrak{b}}\|$ small

6: **Return** $\underbrace{s_{\mathfrak{b}}}_{\in \mathfrak{b}} \cdot \underbrace{y}_{\in (\mathfrak{b} \cdot \mathfrak{p})^{-1}} \in \mathfrak{p}^{-1}$.

▷ $\|y \cdot s_{\mathfrak{b}}\|$ small

Our \mathcal{P}^{-1} -ideal-SVP to \mathcal{P} -ideal-SVP reduction

The oracle \mathcal{O} solves ideal-HSVP for \mathfrak{p} uniform prime of norm in $[A, B]$.

Input: An ideal $I = \mathfrak{p}^{-1}$ with \mathfrak{p} uniform prime of norm in $[A, B]$.

Output: $x \in \mathfrak{p}^{-1} \setminus \{0\}$ small.

1: Let $s_{\mathfrak{p}} = \mathcal{O}(\mathfrak{p})$.

2: Let $(\mathfrak{b}, y) = \text{SampleIdeal}_{A,B}(\mathfrak{p}, s_{\mathfrak{p}})$.

▷ $\|y\|$ small

3: **if** \mathfrak{b} is not prime. **then**

4: Fail.

5: **Let** $s_{\mathfrak{b}} = \mathcal{O}(\mathfrak{b})$.

▷ $\|s_{\mathfrak{b}}\|$ small

6: **Return** $\underbrace{s_{\mathfrak{b}}}_{\in \mathfrak{b}} \cdot \underbrace{y}_{\in (\mathfrak{b} \cdot \mathfrak{p})^{-1}} \in \mathfrak{p}^{-1}$.

▷ $\|y \cdot s_{\mathfrak{b}}\|$ small

Our \mathcal{P}^{-1} -ideal-SVP to \mathcal{P} -ideal-SVP reduction

The oracle \mathcal{O} solves ideal-HSVP for \mathfrak{p} uniform prime of norm in $[A, B]$.

Input: An ideal $I = \mathfrak{p}^{-1}$ with \mathfrak{p} uniform prime of norm in $[A, B]$.

Output: $x \in \mathfrak{p}^{-1} \setminus \{0\}$ small.

1: Let $s_{\mathfrak{p}} = \mathcal{O}(\mathfrak{p})$.

2: Let $(\mathfrak{b}, y) = \text{SampleIdeal}_{A,B}(\mathfrak{p}, s_{\mathfrak{p}})$.

▷ $\|y\|$ small

3: **if** \mathfrak{b} is not prime. **then**

4: Fail.

5: Let $s_{\mathfrak{b}} = \mathcal{O}(\mathfrak{b})$.

▷ $\|s_{\mathfrak{b}}\|$ small

6: **Return** $\underbrace{s_{\mathfrak{b}}}_{\in \mathfrak{b}} \cdot \underbrace{y}_{\in (\mathfrak{b} \cdot \mathfrak{p})^{-1}} \in \mathfrak{p}^{-1}$.

▷ $\|y \cdot s_{\mathfrak{b}}\|$ small

Wrapping up

Contributions:

- Solving ideal-HSVP on average over inverses of primes is at least as hard as solving ideal-HSVP on average over primes.
- This gives an NTRU instance distribution with hardness based on ideal-HSVP for all ideals.

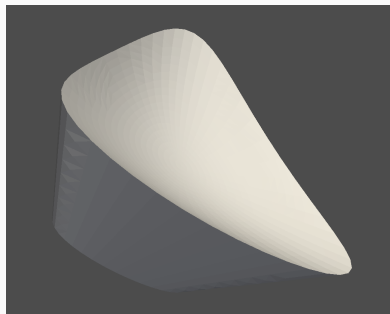
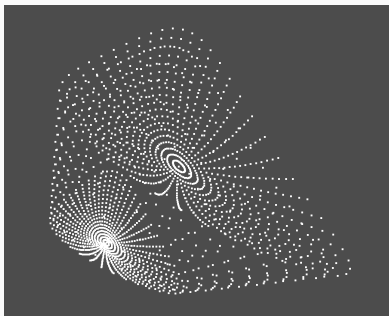
Contributions:






- Solving ideal-HSVP on average over inverses of primes is at least as hard as solving ideal-HSVP on average over primes.
- This gives an NTRU instance distribution with hardness based on ideal-HSVP for all ideals.





Open problems:

- Can we have such reduction without factoring?
- Can we get rid of the cost dependency in ρ_K ?
- Can we have more precise approximates for the running time?

Any question?



-  K. de Boer, L. Ducas, A. Pellet-Mary, and B. Wesolowski, *Random self-reducibility of Ideal-SVP via Arakelov random walks*, CRYPTO, 2020.
-  K. Boudgoust, E. Gachon, and A. Pellet-Mary, *Some easy instances of Ideal-SVP and implications on the partial Vandermonde knapsack problem*, CRYPTO, 2022.
-  K. de Boer, *Random walks on arakelov class groups.*, Ph.D. thesis, Leiden University, 2022, Available on request from the author.
-  R. Cramer, L. Ducas, C. Peikert, and O. Regev, *Recovering short generators of principal ideals in cyclotomic rings*, EUROCRYPT 2016, 2016.
-  R. Cramer, L. Ducas, and B. Wesolowski, *Short Stickelberger class relations and application to Ideal-SVP*, EUROCRYPT, 2017.

-  C. Gentry, *A fully homomorphic encryption scheme*, Ph.D. thesis, Stanford University, 2009.
-  A. Pellet-Mary, G. Hanrot, and D. Stehlé, *Approx-SVP in ideal lattices with pre-processing*, EUROCRYPT, 2019.
-  A. Pellet-Mary and D. Stehlé, *On the hardness of the NTRU problem*, ASIACRYPT, 2021.
-  Quartl, *Matrix pattern qt13*, 2014, File: Matrix pattern qt13.svg.