

Lattice cryptography and cryptanalysis

Wessel van Woerden (Université de Bordeaux, IMB, Inria).

Part I

Lattice theory

- ▶ Lattices
- ▶ Hard problems

Cryptography

- ▶ Trapdoor bases
- ▶ Encryption, Signature

Cryptanalysis

- ▶ Lattice Sieving
- ▶ Basis Reduction

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Lattices used in cryptography

- ▶ SIS, LWE, declWE
- ▶ Security proofs

Hardness Reductions

- ▶ search to decision
- ▶ WC to AC reductions

Algebraic Lattices

- ▶ Ideal and module lattices
- ▶ NTRU, RLWE, mod-LWE

Plan

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acknowledgements: many slides adapted from Alice Pellet-Mary!

Lattice theory

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most important:

row vectors $(x\mathbf{G}) \rightarrow$ column vectors $(\mathbf{G}x)$

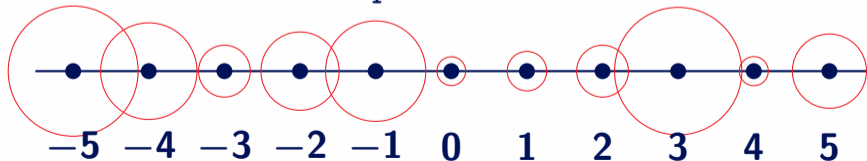
Lattice

A lattice $\mathcal{L} \subset \mathbb{R}^n$ is a discrete subgroup of \mathbb{R}^n .

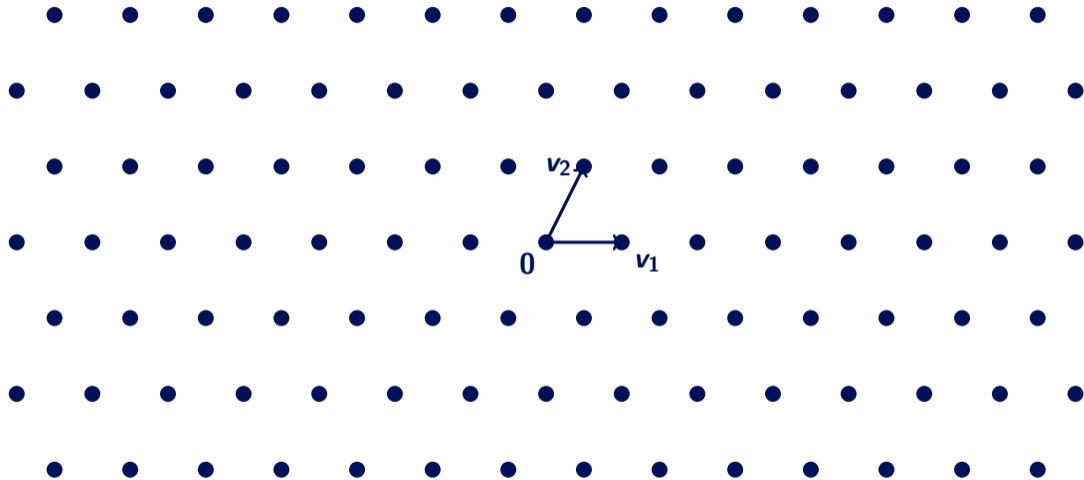
Discrete

For every $\mathbf{v} \in \mathcal{L}$ there exists an open ball around \mathbf{v} that contains no other elements from \mathcal{L} .

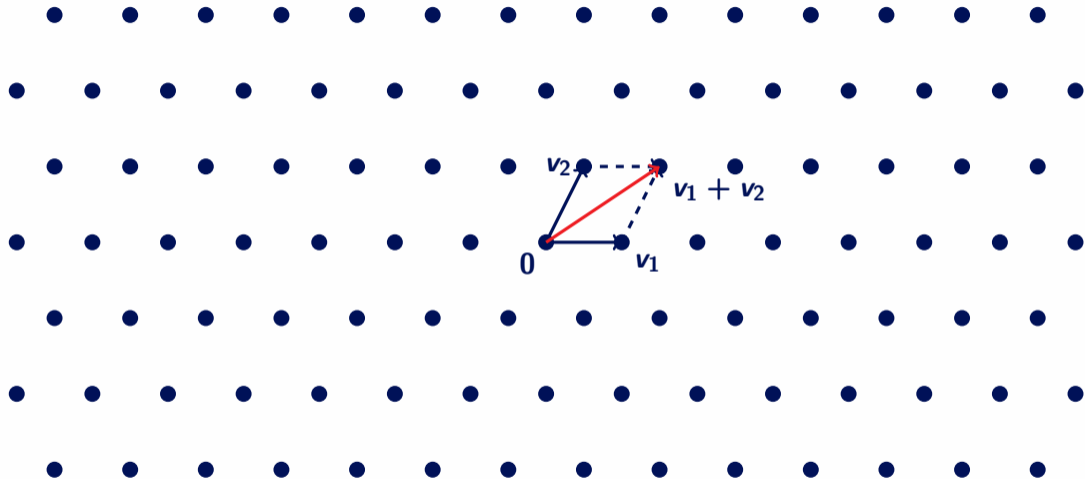
Example $\mathbb{Z} \subset \mathbb{R}$:



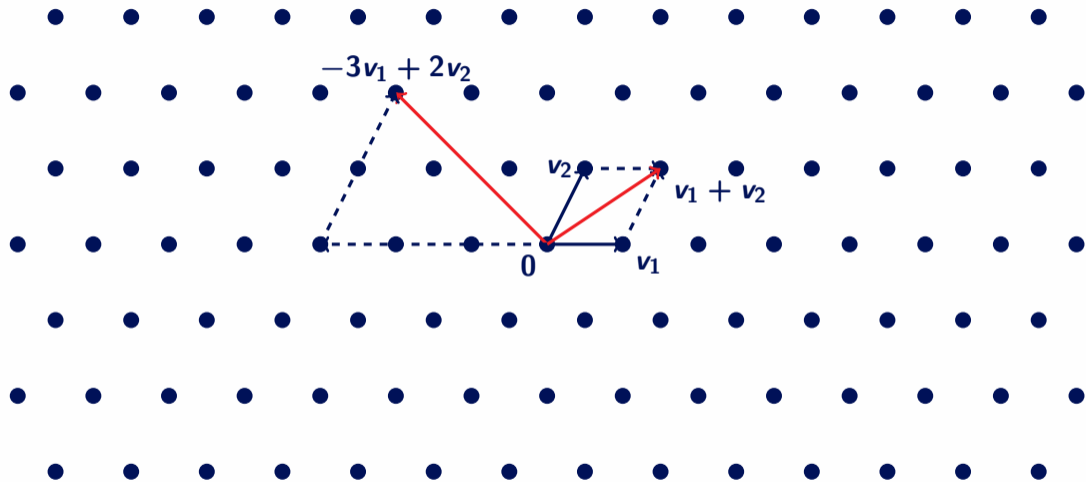
Additive



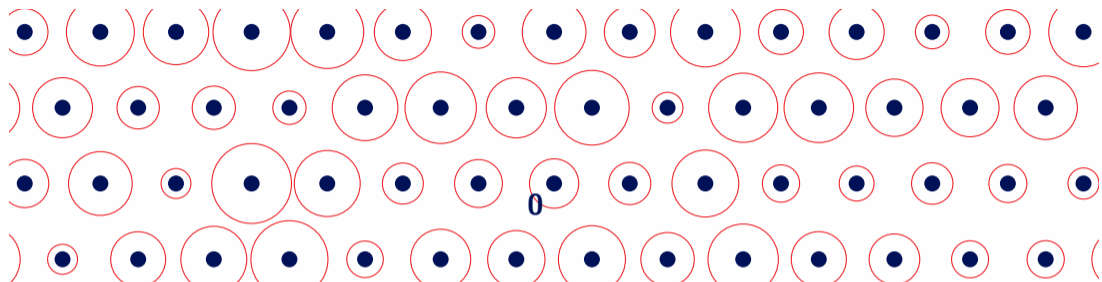
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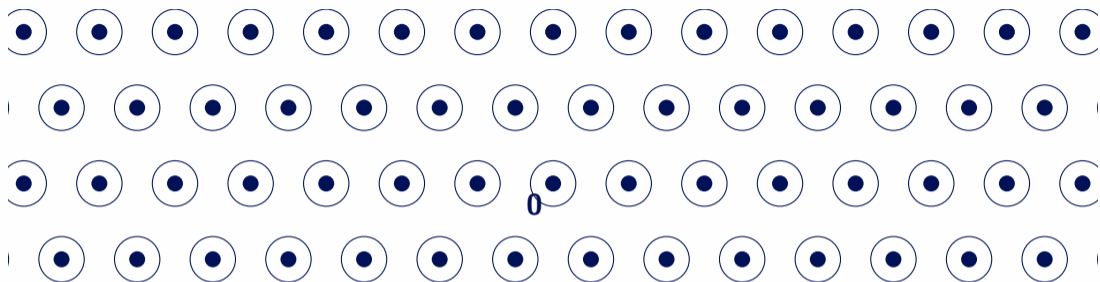
Additive



First minimum of a lattice

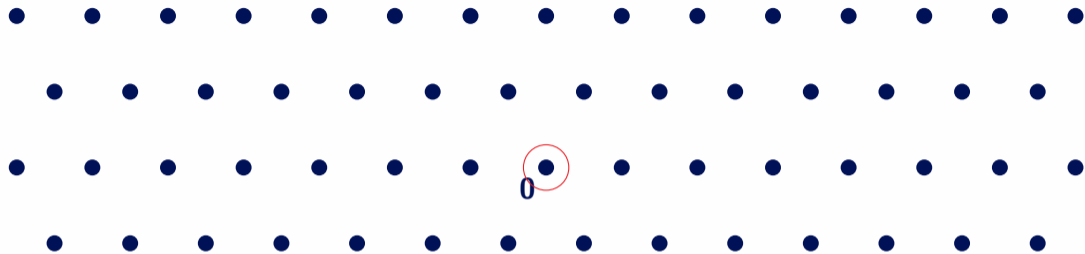


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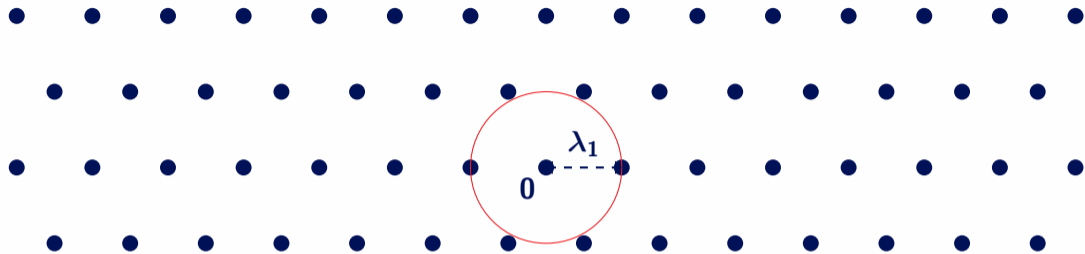
By the additivity the neighborhood of every lattice point looks the same.

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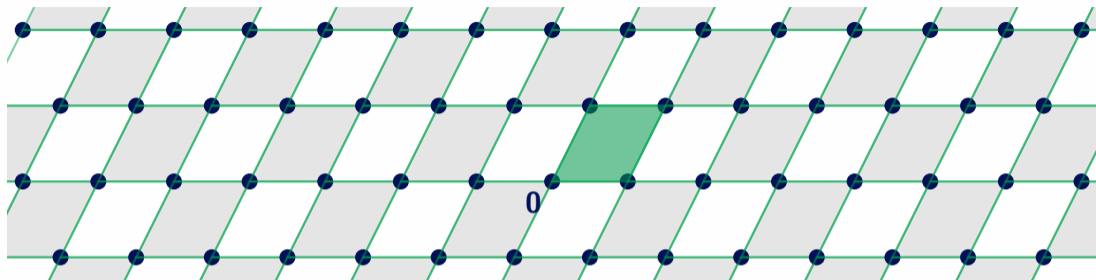
First minimum of a lattice



The **first minimum** $\lambda_1(\mathcal{L})$ of a lattice \mathcal{L} is the length of the shortest nonzero lattice vector:

$$\lambda_1(\mathcal{L}) = \min_{x \in \mathcal{L} \setminus \{0\}} \{\|x\|\} > 0.$$

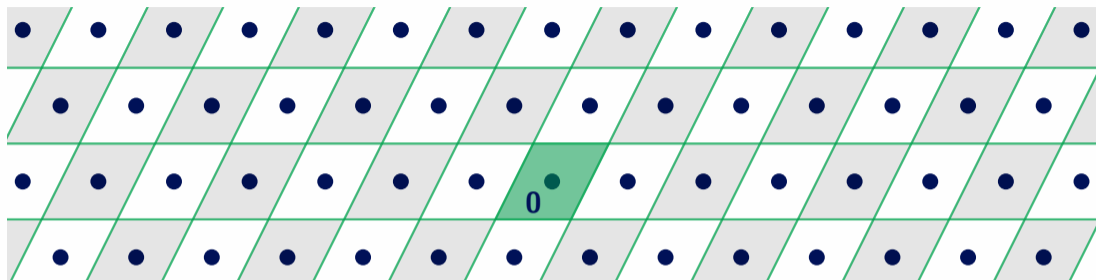
Volume of a lattice



The **volume** $\text{vol}(\mathcal{L})$ of a lattice \mathcal{L} is the (co-)volume of any fundamental area w.r.t. translation of the lattice:

$$\text{vol}(\mathcal{L}) = \text{vol}(\mathbb{R}^n / \mathcal{L}) \quad (\text{density}(\mathcal{L}) = 1 / \text{vol}(\mathcal{L}))$$

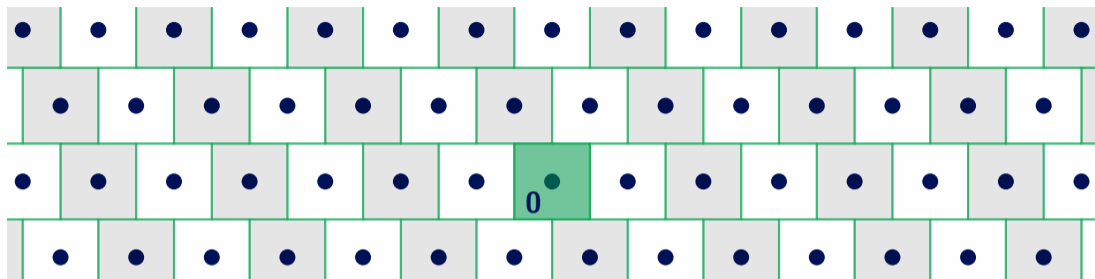
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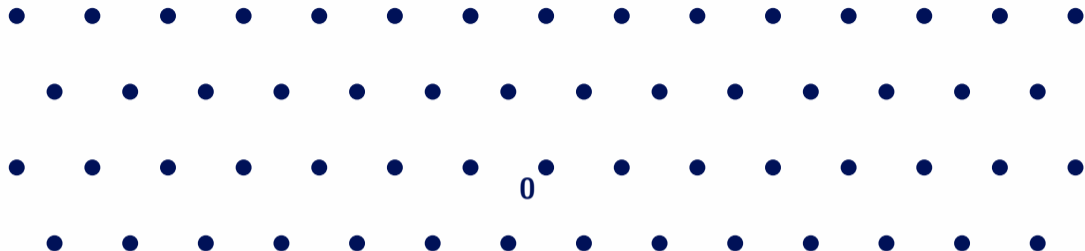
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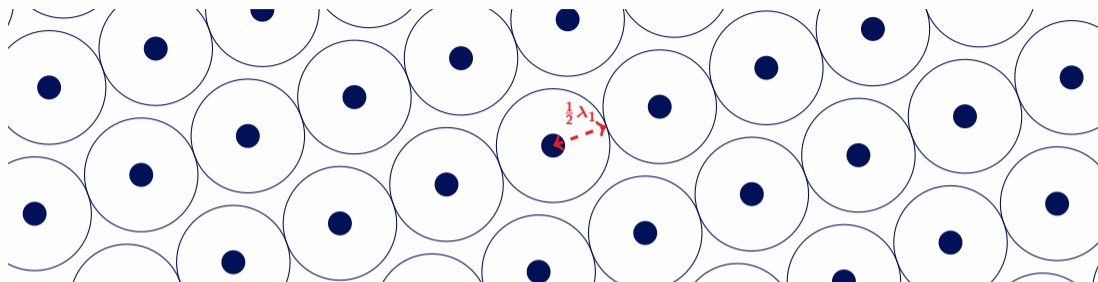
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Minkowski's Theorem

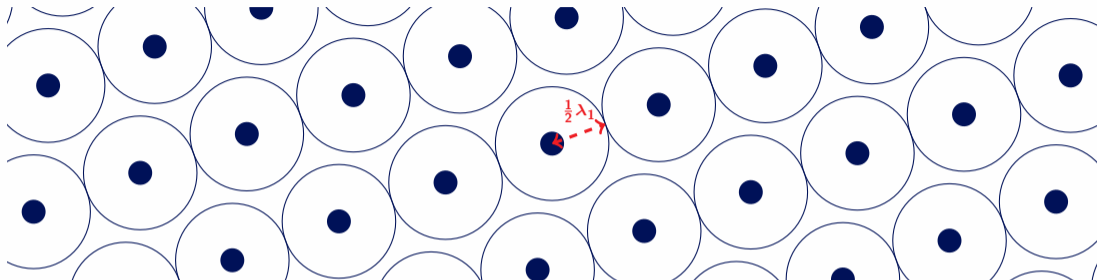


Minkowski's Theorem

For a full-rank lattice $\mathcal{L} \subset \mathbb{R}^n$ we have

$$\text{vol}\left(\frac{1}{2}\lambda_1(\mathcal{L}) \cdot \mathcal{B}^n\right) \leq \text{vol}(\mathcal{L})$$

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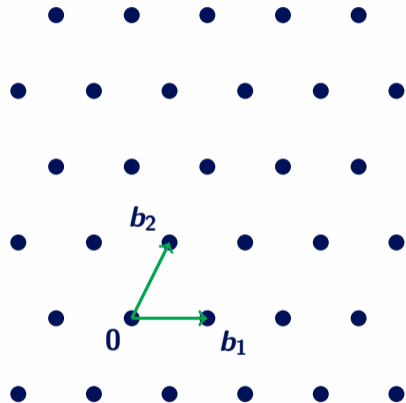
$$\lambda_1(\mathcal{L}) \leq 2 \underbrace{\frac{\text{vol}(\mathcal{L})^{1/n}}{\text{vol}(\mathcal{B}^n)^{1/n}}}_{\text{Mk}(\mathcal{L})} \approx 2 \cdot \sqrt{n/2\pi e} \cdot \text{vol}(\mathcal{L})^{1/n}$$

Basis Representation

Lattice basis

\mathbb{R} -linearly independent $\mathbf{b}_1, \dots, \mathbf{b}_n$

$$\mathcal{L}(B) := \left\{ \sum_i x_i \mathbf{b}_i : \mathbf{x} \in \mathbb{Z}^n \right\} \subset \mathbb{R}^n.$$



Basis Representation

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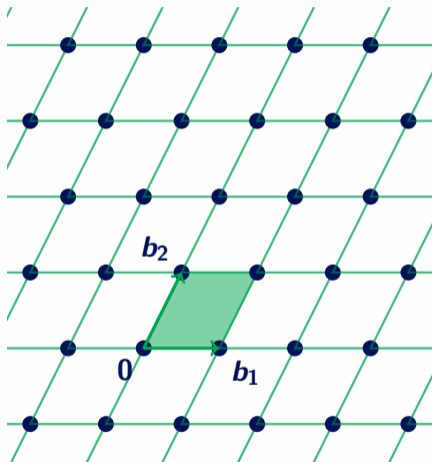
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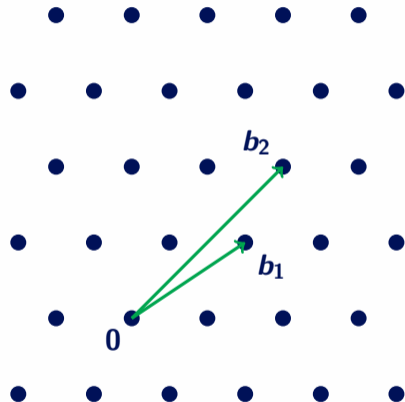
Fundamental Parallelepiped

$$\mathcal{P}(B) = B \cdot [0, 1]^n$$

$$\text{vol}(\mathcal{L}) = \text{vol}(\mathcal{P}(B)) = |\det(B)|$$



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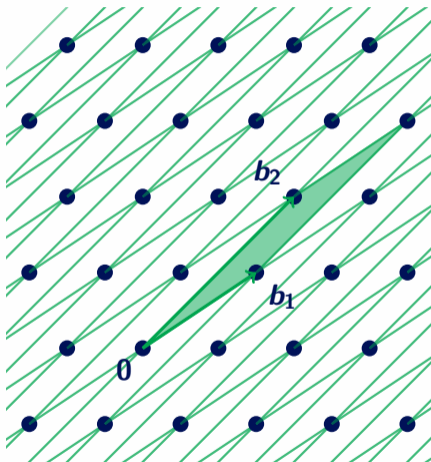
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Infinitely many distinct bases

$$\mathbf{B}' = \mathbf{B} \cdot \mathbf{U} \text{ for } \mathbf{U} \in \mathcal{GL}_n(\mathbb{Z}).$$

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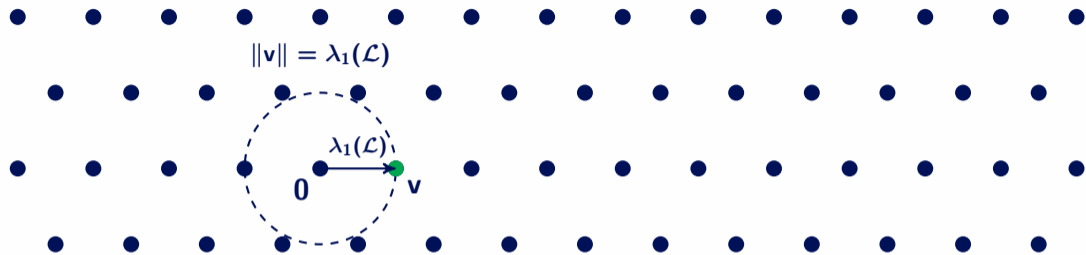
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Shortest Vector Problem (SVP)

Find a **shortest nonzero** vector

$v \in \mathcal{L}$ of length $\lambda_1(\mathcal{L})$.

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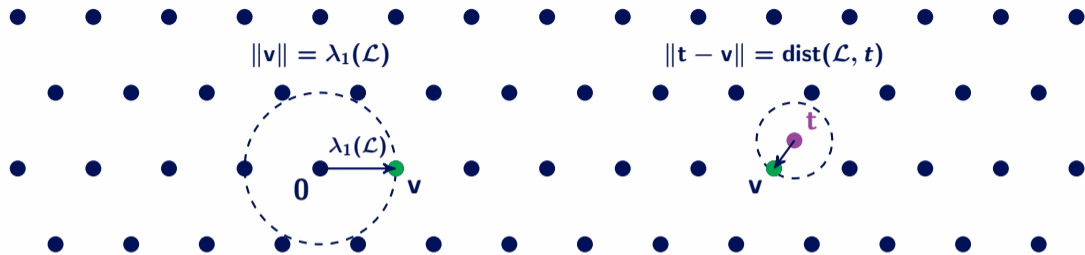
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Supposedly **hard** to solve when n is large
(even with a quantum computer)

How hard is SVP/CVP?

In theory: best algorithm has asymptotic complexity $2^{c \cdot n + o(n)}$

classical: $c \approx 0.292$, or quantum: $c \approx 0.265$)

\Rightarrow not polynomial

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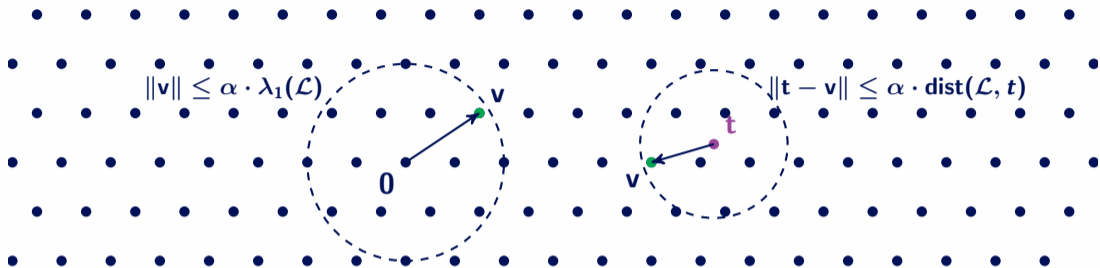
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- ▶ from $n = 400$ to $n = 1000$ \rightsquigarrow cryptography

Approximate versions



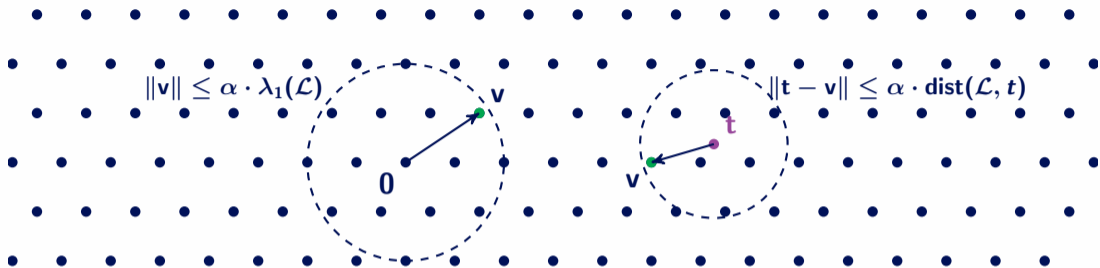
α -approx-SVP

Find a **short nonzero** vector $v \in \mathcal{L}$ of length $\leq \alpha \cdot \lambda_1(\mathcal{L})$.

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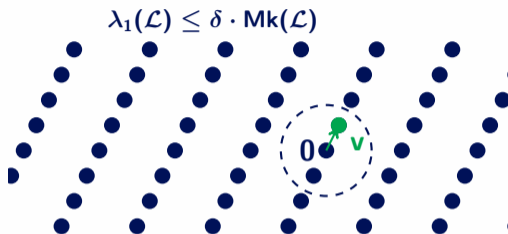
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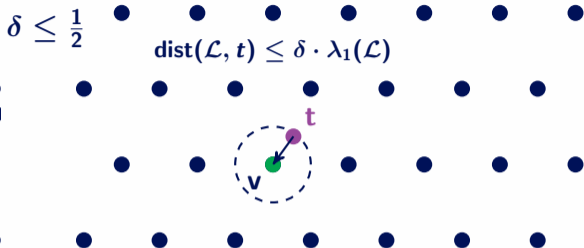
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Promise versions



δ -uSVP

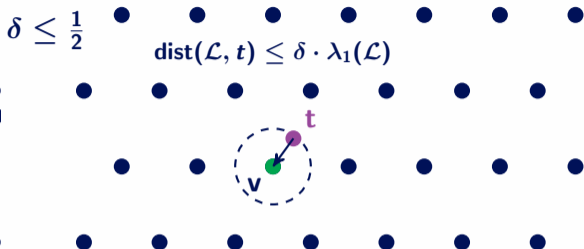
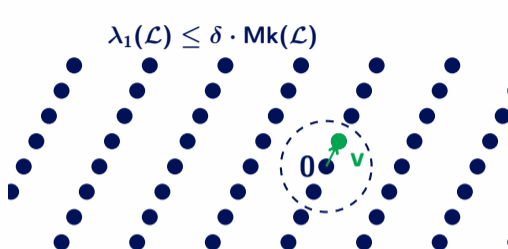
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CVP with a target unusually
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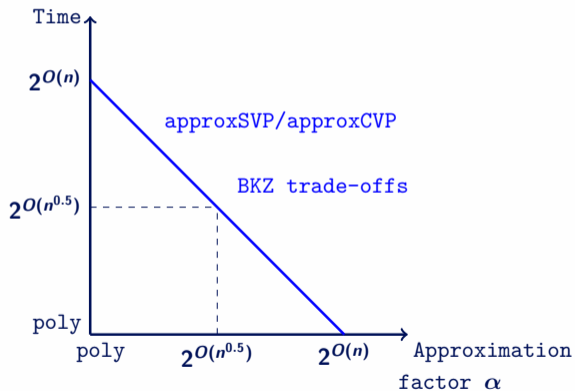
Bounded Distance Decoding (δ -BDD)

CVP with a target **unusually close**
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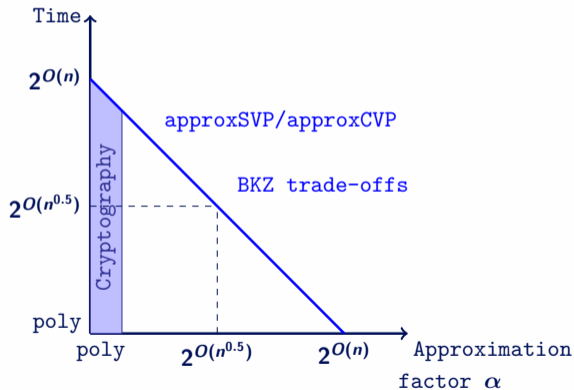
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BKZ algorithm



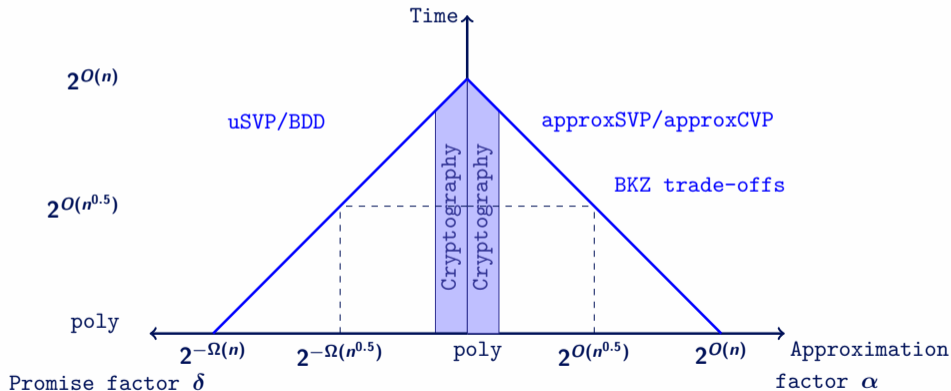
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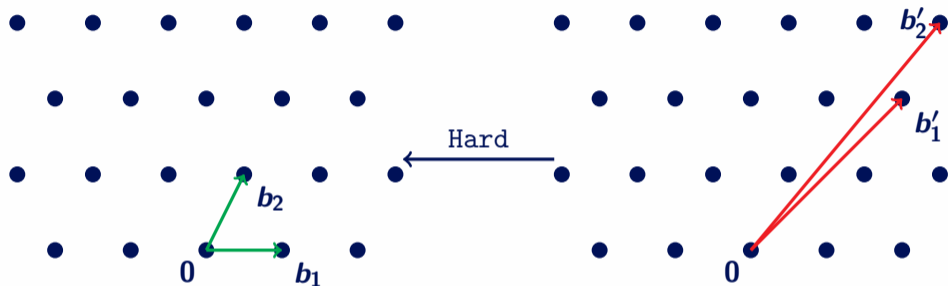
How to build cryptography from this?

Lattice-based cryptography

Good vs bad basis

Good basis (Secret key)

Bad basis (Public key)

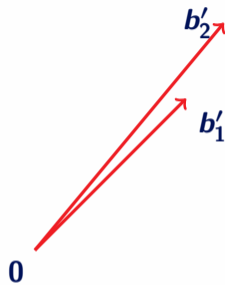
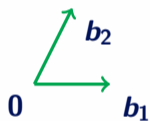


Short and close to orthogonal

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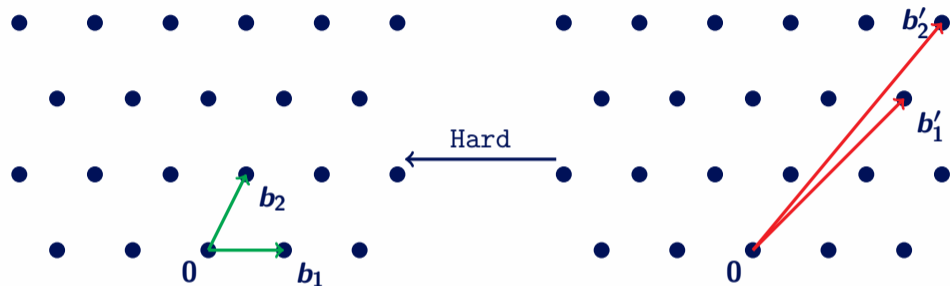


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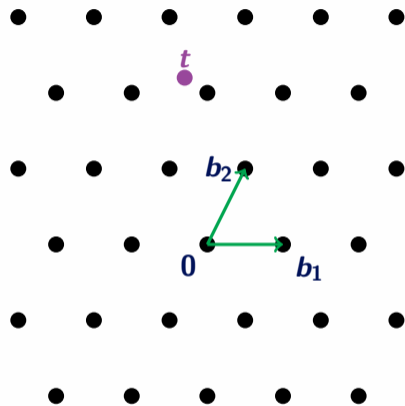
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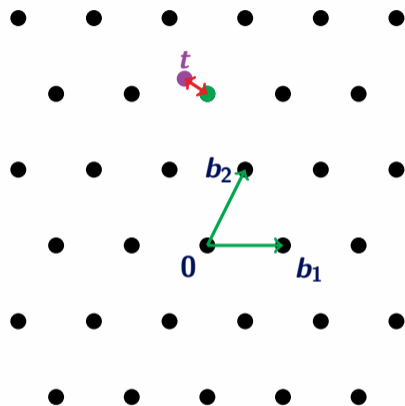
Keygen: Generate a random lattice along with a good basis
(NTRU, LWE, SIS, ...)

Solving CVP with a short basis



Input: $t = -1.4 \cdot b_1 + 2.2 \cdot b_2$

Solving CVP with a short basis

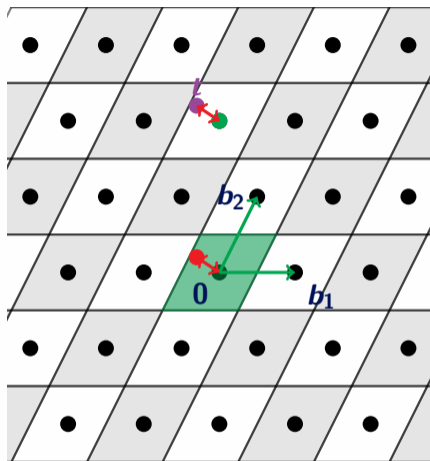


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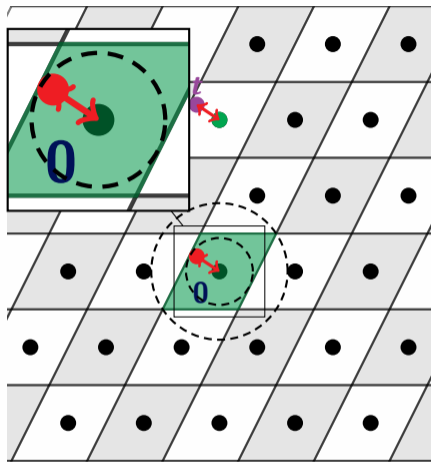
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$$\text{Output: } v = -1 \cdot b_1 + 2 \cdot b_2$$

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$$e \in B \cdot \left[-\frac{1}{2}, \frac{1}{2}\right)^n$$

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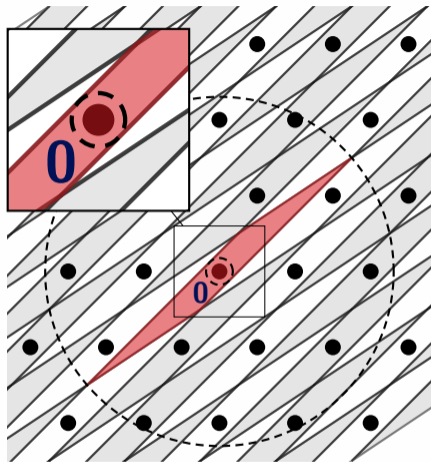
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BDD: inner-radius

approxCVP: outer-radius

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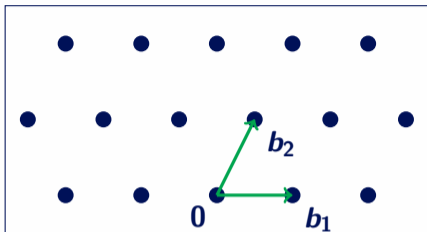
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The better the basis,
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BDD: inner-radius approxCVP: outer-radius

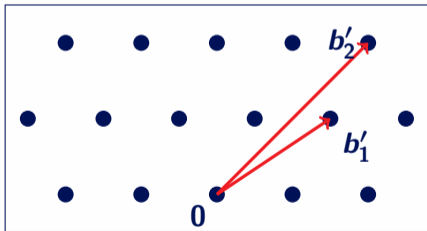
Encryption via BDD



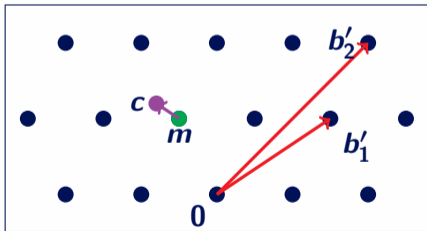
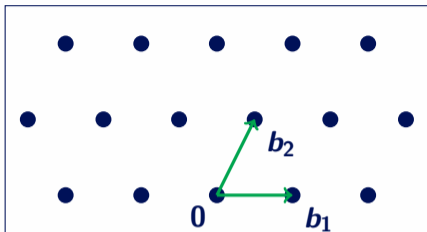
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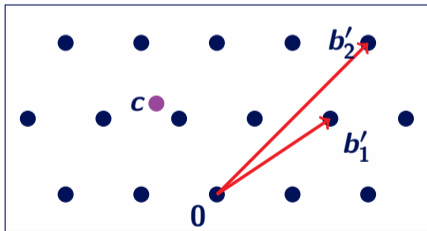
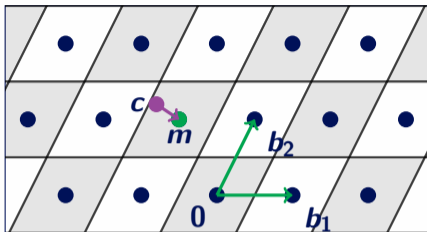
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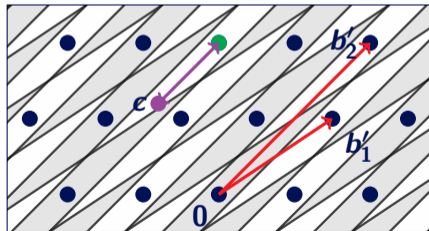
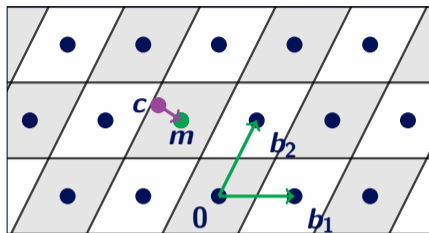
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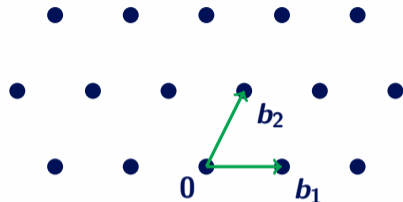
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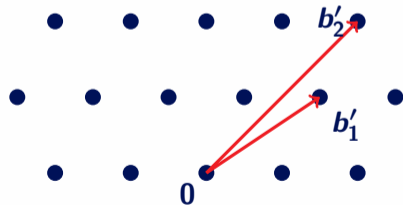
Hash-and-sign signature scheme via approxCVP



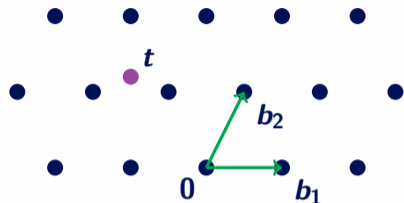
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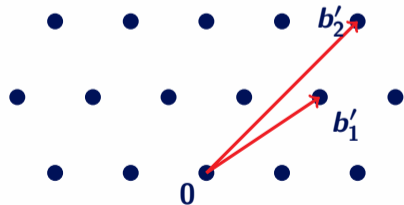
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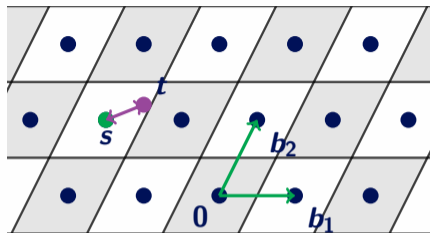
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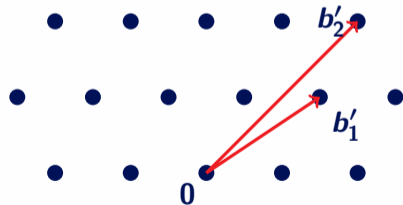
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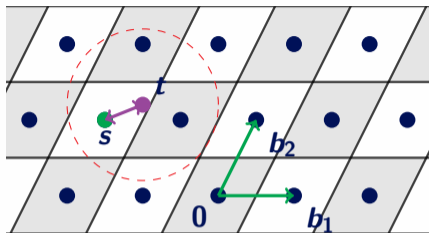
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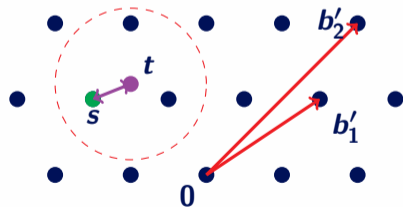
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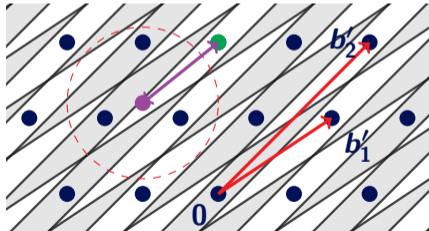
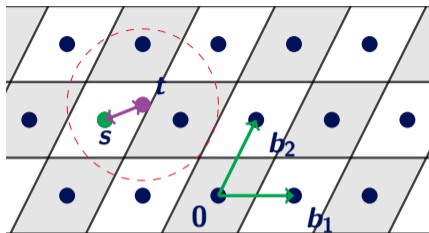
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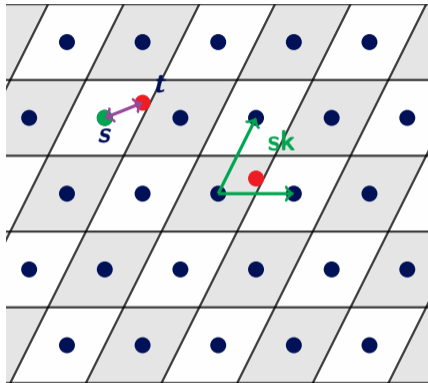
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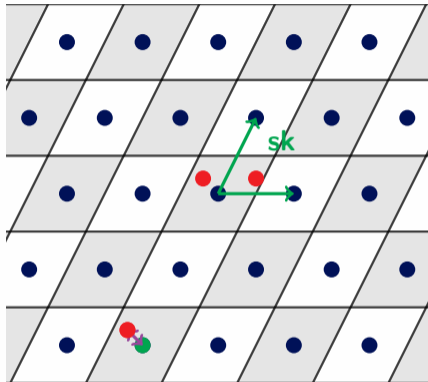
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- ▶ ask for a signature s on m
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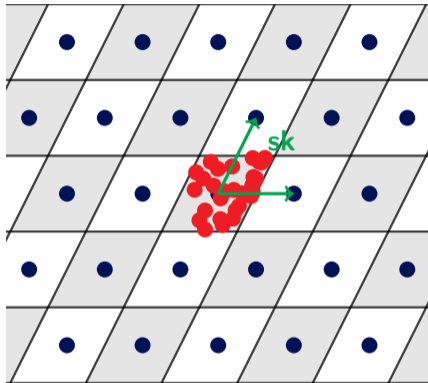
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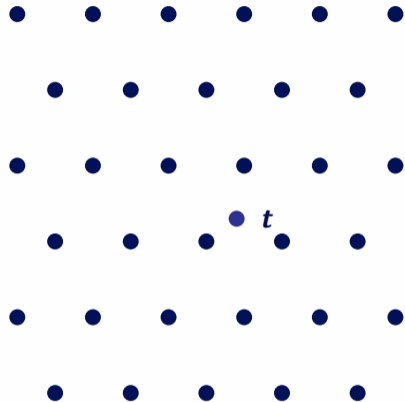


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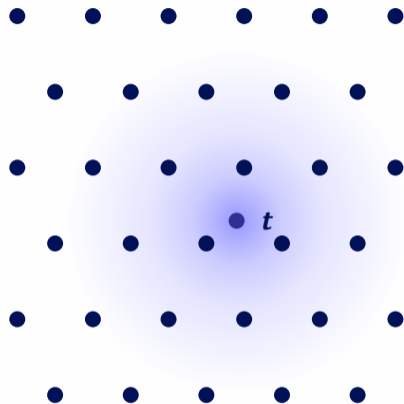
From the shape of the parallelepiped, one can recover the short basis

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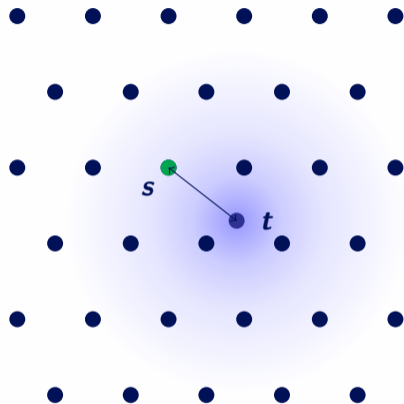
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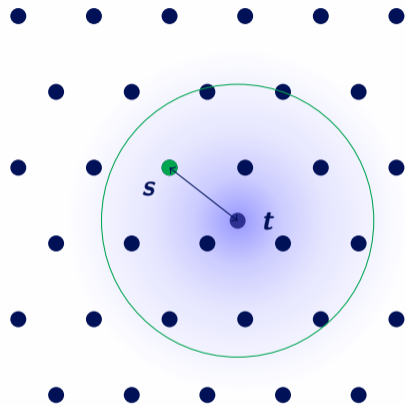
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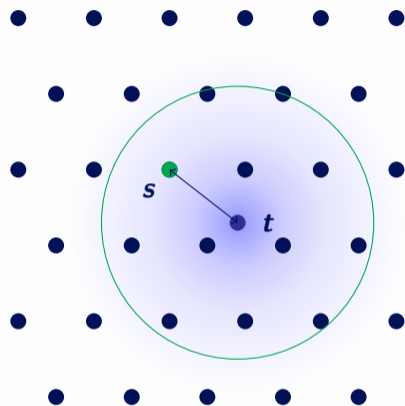
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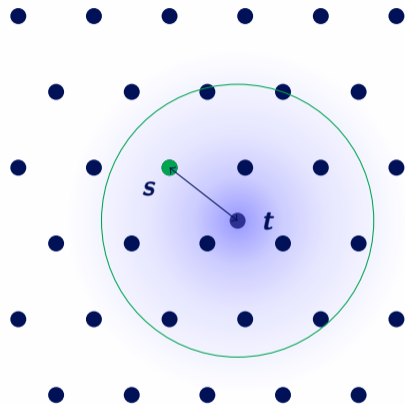
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FALCON = the above + NTRU lattices.

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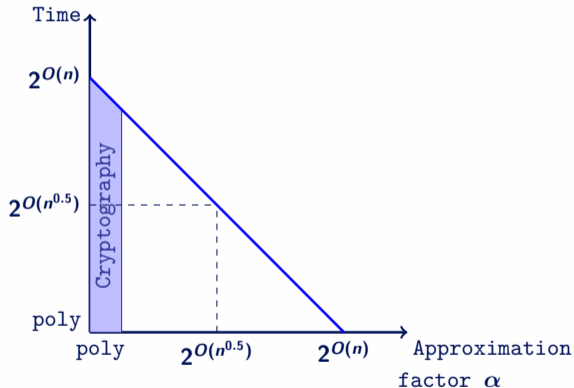
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Cryptanalysis - Algorithms to solve (approx)SVP

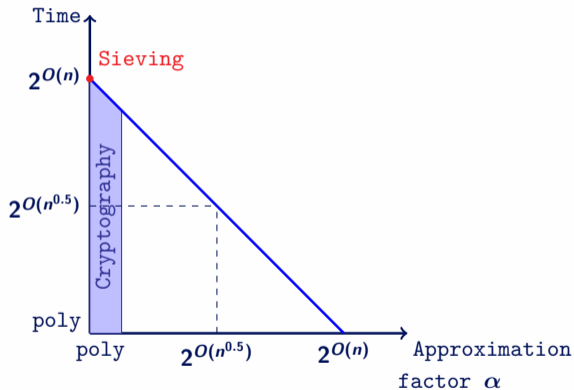
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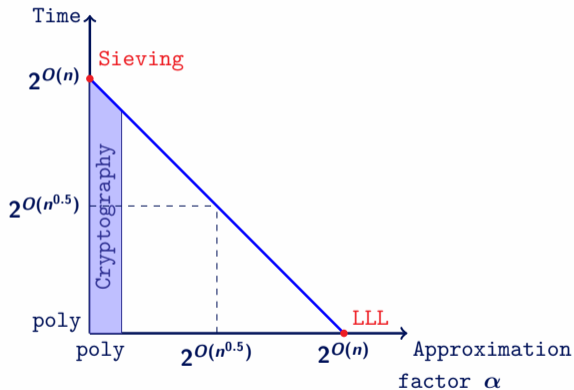
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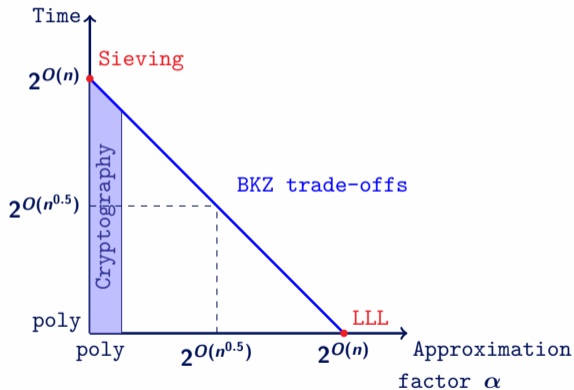
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Heuristically solving SVP with lattice sieving

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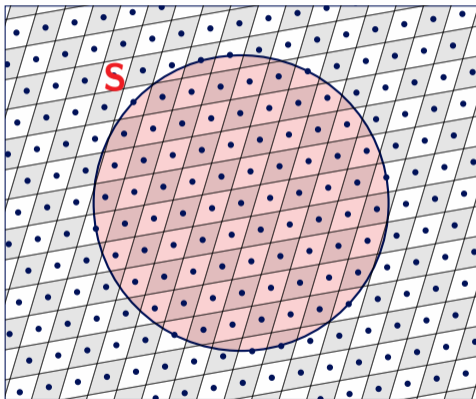
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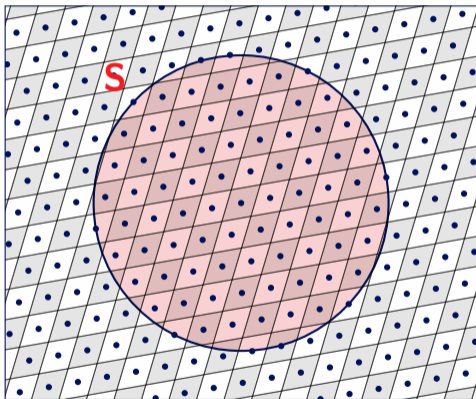
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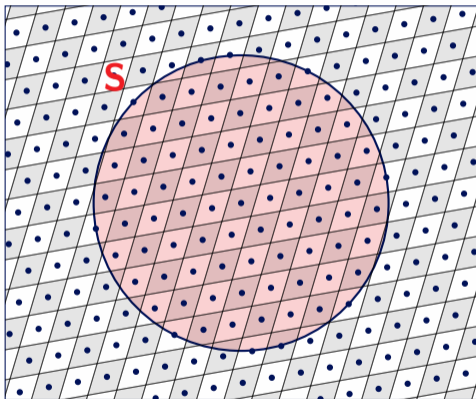


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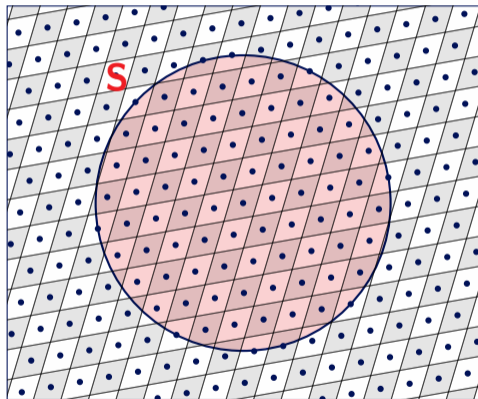
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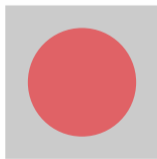
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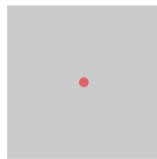
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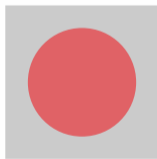
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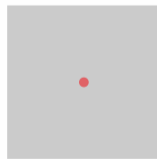
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n -dimensional balls with a fixed radius 'disappear' for large n .

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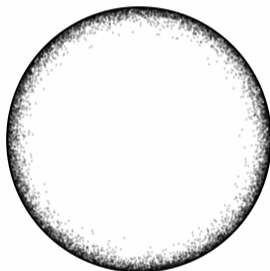
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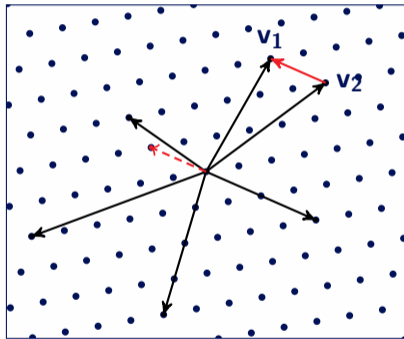
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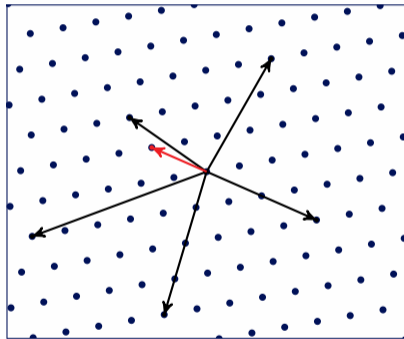
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1. Sample a list $L \subset \mathcal{L}$ of (long) lattice vectors.
2. Repeat:



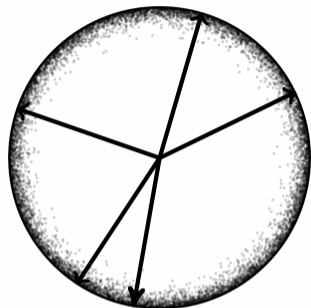
Find close vectors $v_1, v_2 \in L$.



Replace $v_2 \leftarrow v_1 - v_2$.

Heuristic complexity analysis

Start with a list L of
 N vectors of length $\leq \gamma$.

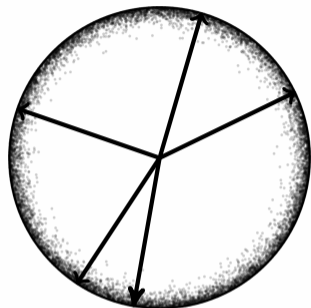


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Heuristic assumption

vectors in list L have uniform directions.



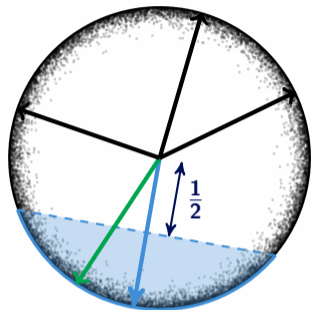
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relative volume spherical cap $\approx (3/4 + \epsilon)^{n/2+o(n)}$



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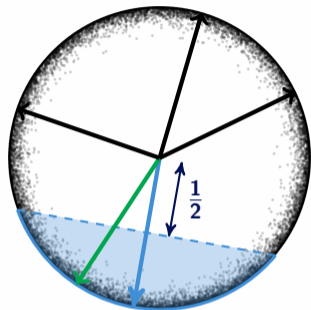
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N^2 pairs, new list size N , so need $N^2 \cdot (3/4)^{n/2} \geq N$.



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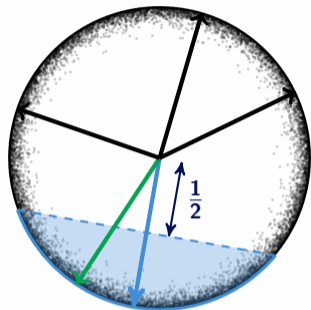
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Space: $N \cdot \text{poly}(n) = (4/3)^{n/2+o(n)} = 2^{0.2075+o(n)}$

Time: $N^2 \cdot \text{poly}(n) = (4/3)^{n+o(n)} = 2^{0.415n+o(n)}$.



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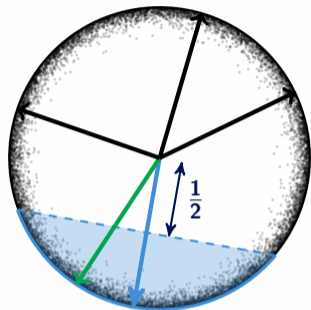
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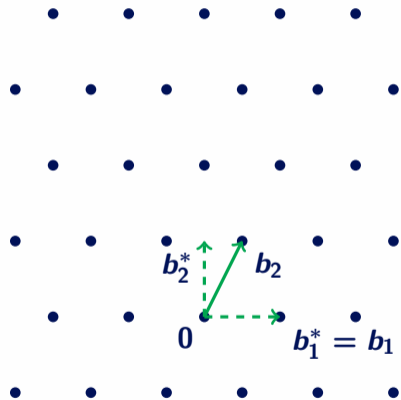


Can be improved to
 $2^{0.292n+o(n)}$!

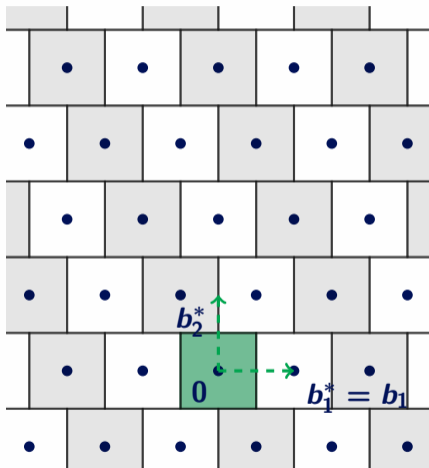
Solving approxSVP/CVP via basis reduction

Gram-Schmidt Orthogonalisation

$$\text{GSO: } \mathbf{b}_i^* := \underbrace{\pi_{(b_1, \dots, b_{i-1})^\perp}}_{\pi_i}(\mathbf{b}_i)$$



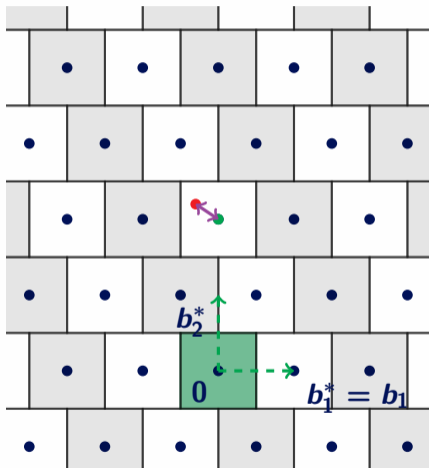
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Nearest plane algorithm

Input: target $t = e$

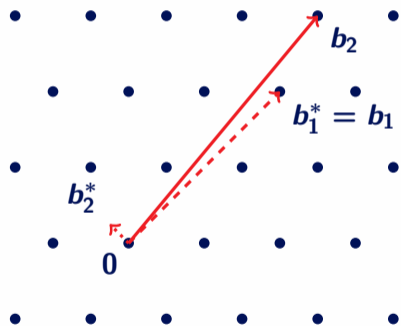
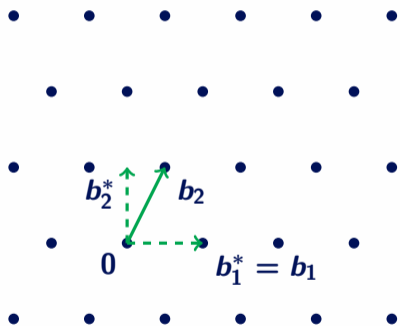
For $j = n, \dots, 1$:

$$e \leftarrow e - \left\lfloor \frac{\langle e, b_j^* \rangle}{\langle b_j^*, b_j^* \rangle} \right\rfloor b_j.$$

Output: $e \in \mathcal{F}_{B^*}$

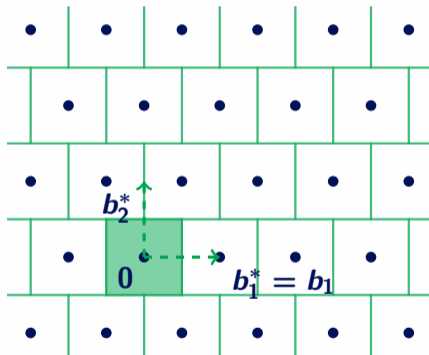
Good vs Bad basis

$$b_i^* := \underbrace{\pi(b_1, \dots, b_{i-1})^\perp}_{\pi_i}(b_i)$$

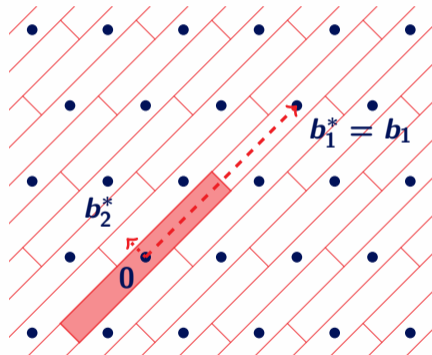


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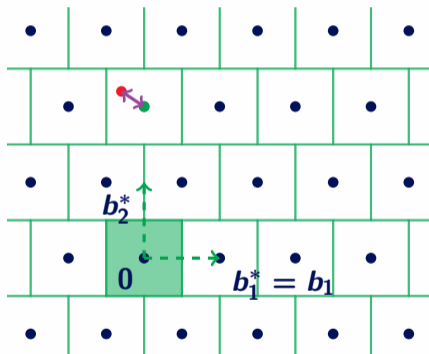


$$\text{vol}(\mathcal{L}) = \text{vol}(\mathcal{F}_{B^*}) = \prod_{i=1}^k \|b_i^*\|$$



Good vs Bad basis

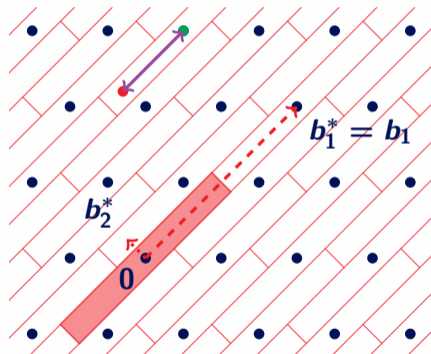
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BDD: $\|e\| < \frac{1}{2} \min_i \|b_i^*\|,$

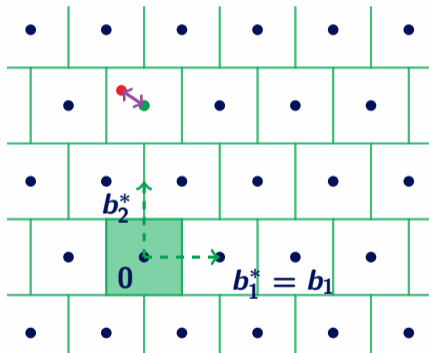
approxCVP: $\|e\|^2 \leq \frac{1}{4} \sum_i \|b_i^*\|^2.$

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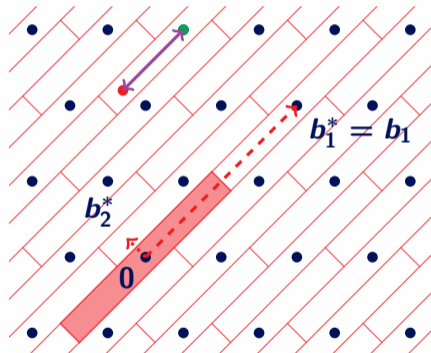
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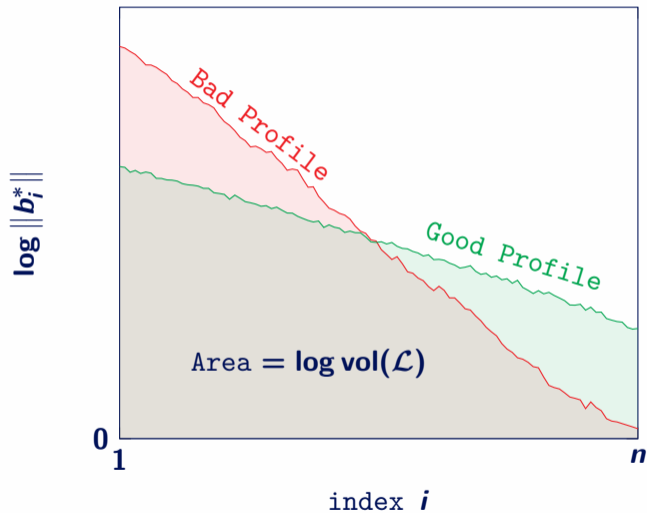
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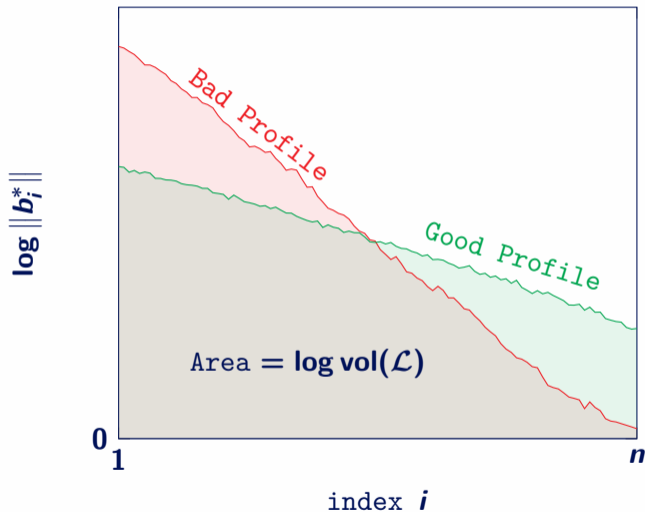
Good Basis:

$$\|b_1^*\| \approx \dots \approx \|b_k^*\|$$

Basis Profile



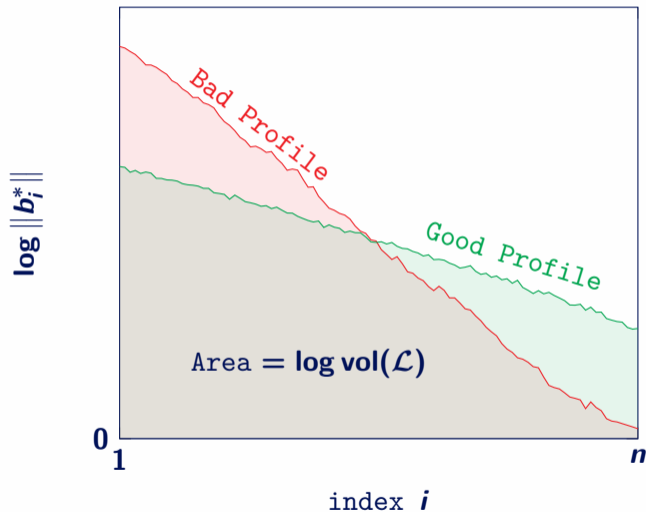
Basis Profile



Basis profile

Measures the **length** and **orthogonality** of a basis

Basis Profile



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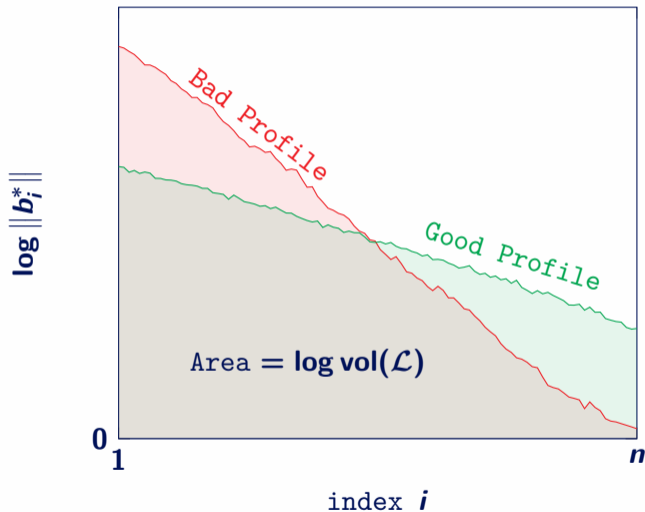
Measures the **length** and **orthogonality** of a basis

Basis reduction

Input: **Bad** basis

Output: **good** basis

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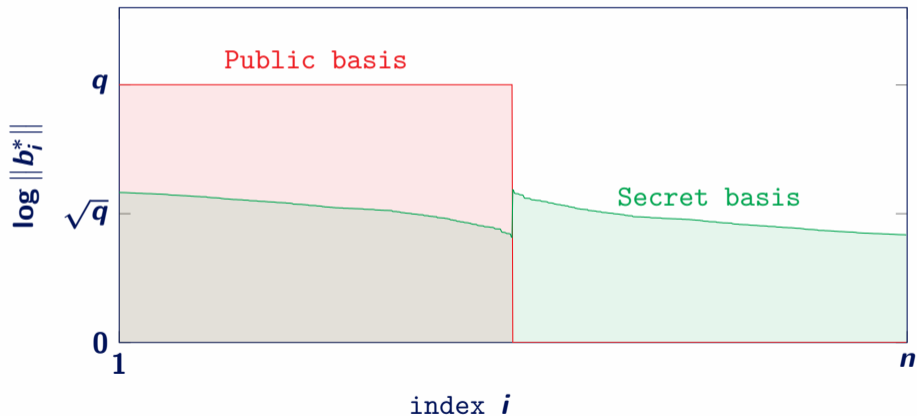
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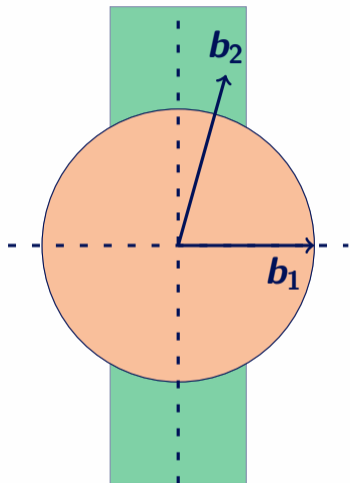
Flatten the profile!

Example: NTRU public vs secret basis

public and secret bases generated from the NTRU problem



Lagrange Reduction (n=2)



Wristwatch Lemma

For any lattice \mathcal{L} of rank 2
there exists a basis (b_1, b_2) s.t.

$$\|b_1\| \leq \|b_2\|$$

$$|\langle b_1, b_2 \rangle| \leq \frac{1}{2} \|b_1\|$$



$$\|b_1^*\| \leq \sqrt{\frac{4}{3}} \cdot \|b_2^*\|$$

Definition

A basis \mathbf{B} of \mathcal{L} is LLL-reduced if $(\pi_i(\mathbf{b}_i), \pi_i(\mathbf{b}_{i+1}))$ is Lagrange Reduced for all $i < n$.

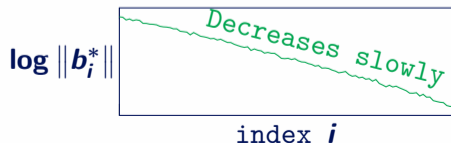
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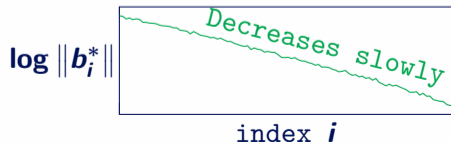
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Algorithm

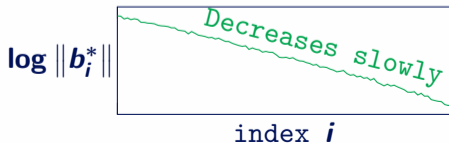
While $\exists i$ s.t. $(\pi_i(\mathbf{b}_i), \pi_i(\mathbf{b}_{i+1}))$ is **not** Lagrange Reduced, Lagrange Reduce it.



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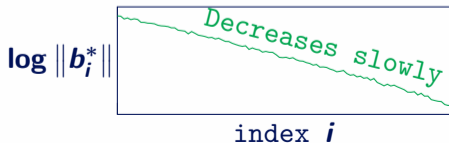
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Termination in poly-time:

Requires a slight relaxation.
(ϵ -Lagrange Reduced)

Proof argument:

$$P = \sum_{i \leq n} (n + 1 - i) \cdot \log \|\mathbf{b}_i^*\|$$

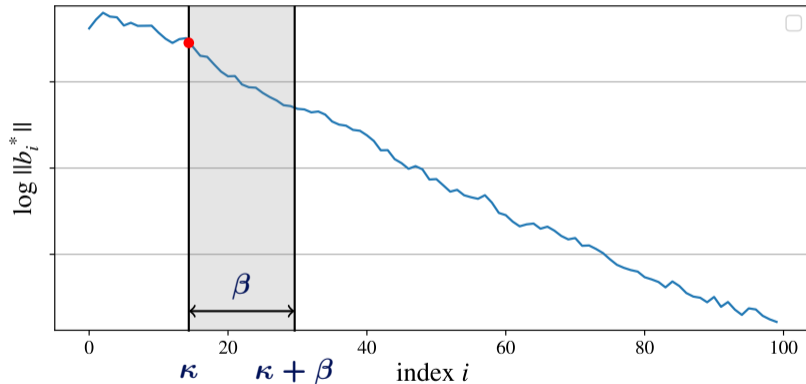
Decreases by ϵ at each step
and is lower-bounded.

BKZ algorithm

- ▶ Define the projected sublattice basis $B_{l:r} := (\pi_l(b_l), \dots, \pi_l(b_{r-1}))$.

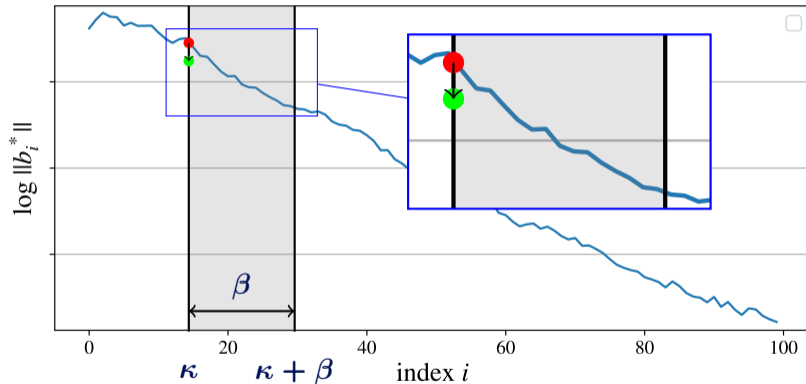
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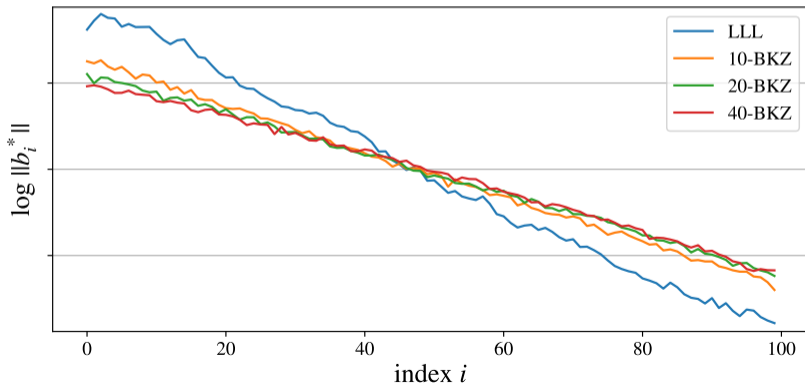
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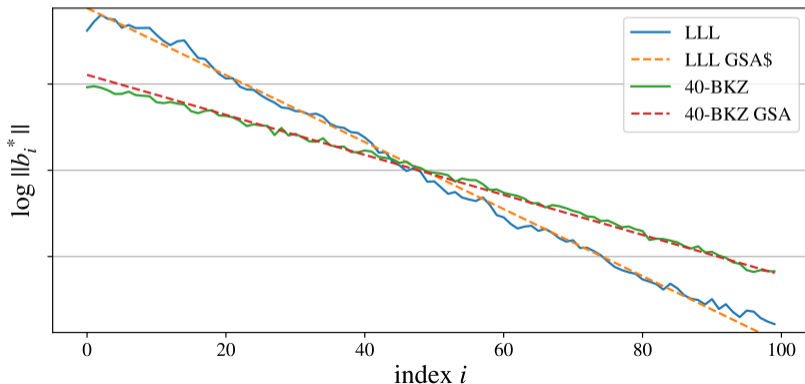
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- ▶ Behaviour well understood for ‘random’ lattices. [GSA]



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We have seen:

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- ▶ Same algorithms also solve promise variants uSVP and BDD

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- ▶ Keygen: what families of lattices to use? (SIS, LWE, NTRU, ...)
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- ▶ More efficiency: algebraic lattices (ideal and module lattices)

Part II

Part I

Lattice theory

- ▶ Lattices
- ▶ Hard problems

Cryptography

- ▶ Trapdoor bases
- ▶ Encryption, Signature

Cryptanalysis

- ▶ Lattice Sieving
- ▶ Basis Reduction

Part II

Lattices used in cryptography

- ▶ SIS, LWE, decLWE
- ▶ Security proofs

Hardness Reductions

- ▶ search to decision
- ▶ WC to AC reductions

Algebraic Lattices

- ▶ Ideal and module lattices
- ▶ NTRU, RLWE, mod-LWE

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For crypto, we need problems that are hard *on average*

(i.e., for a random instance, the problem is hard with overwhelming probability)

random q -ary lattices

q-ary lattices

Notations: q, n, m integers, $1 \leq n \ll m$, $\mathbb{Z}_q := \mathbb{Z}/q\mathbb{Z}$

- ▶ A lattice $\mathcal{L} \subset \mathbb{R}^m$ of dimension m is called **q-ary** if

$$q\mathbb{Z}^m \subset \mathcal{L} \subset \mathbb{Z}^m.$$

- ▶ Let $\mathbf{A} \in \mathbb{Z}_q^{m \times n}$, then we define the **row-generated q-ary** lattice

$$\Lambda_q(\mathbf{A}) := \{y \in \mathbb{Z}^m : y \equiv \mathbf{A}x \pmod{q} \text{ for some } x \in \mathbb{Z}_q^n\} = \mathbf{A}\mathbb{Z}^n + q\mathbb{Z}^m$$

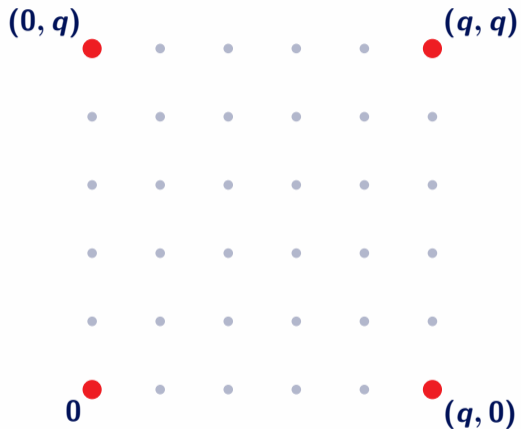
- ▶ and the **parity-check q-ary** lattice

$$\Lambda_q^\perp(\mathbf{A}) := \{x \in \mathbb{Z}^m : x^\top \mathbf{A} \equiv 0 \pmod{q}\} = \ker(\mathbf{A}^\top : \mathbb{Z}^m \rightarrow \mathbb{Z}_q^n)$$

- ▶ **Exercise:** if q prime and \mathbf{A} has full column-rank, then

$$\text{vol}(\Lambda_q(\mathbf{A})) = q^{m-n}, \quad \text{vol}(\Lambda_q^\perp(\mathbf{A})) = q^n$$

Example

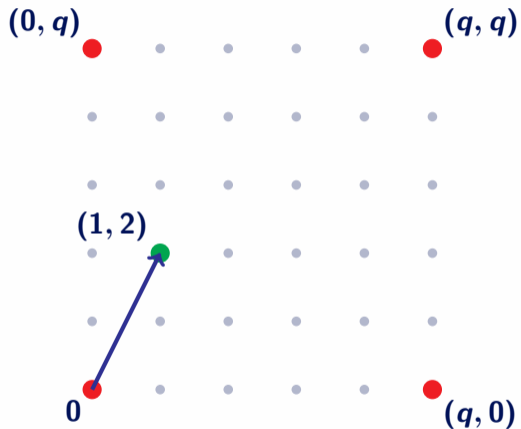


Suppose $q = 5, n = 1, m = 2,$

$$A = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

$$\Lambda_q(A) = A\mathbb{Z}^n + q\mathbb{Z}^m = \begin{pmatrix} 1 \\ 2 \end{pmatrix} \cdot \mathbb{Z} + 5\mathbb{Z}^2$$

Example

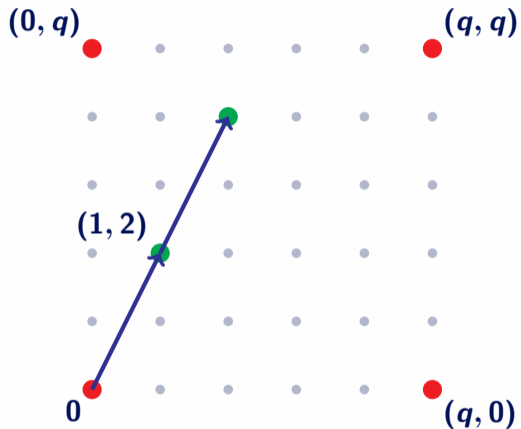


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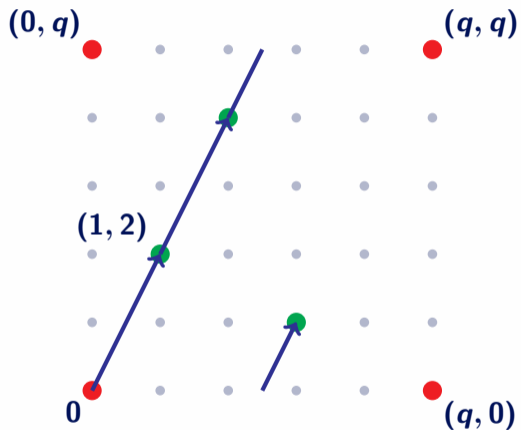


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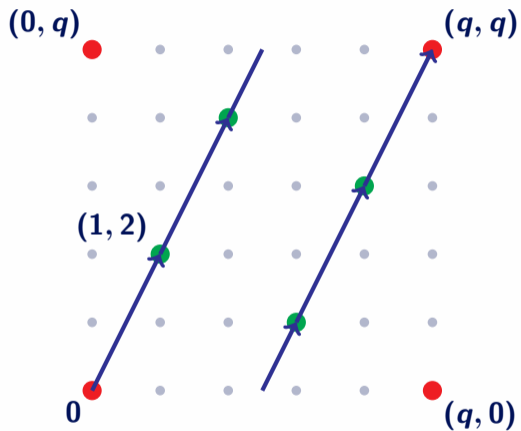


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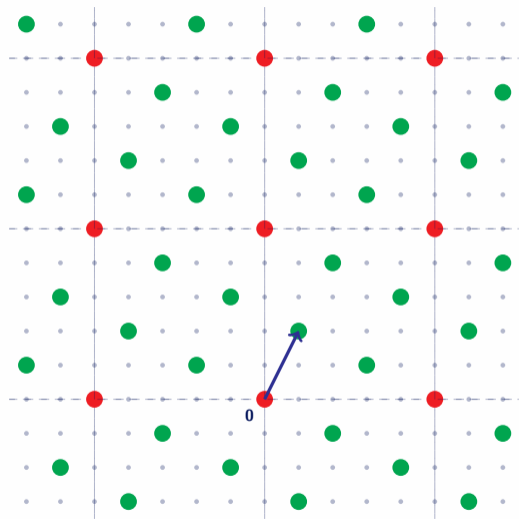


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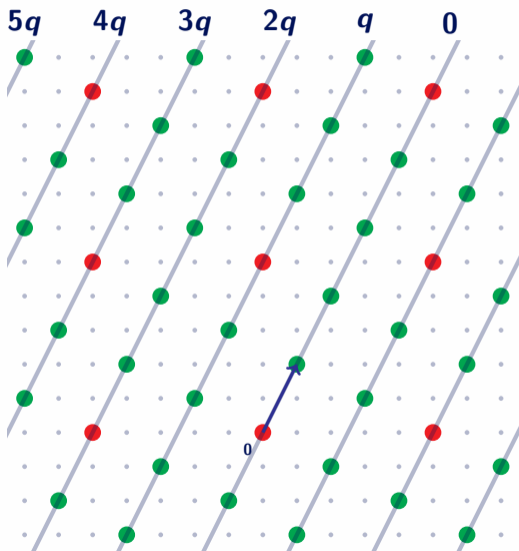


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Parity-check representation:

$$\begin{aligned} \Lambda_q \left(\begin{pmatrix} 1 \\ 2 \end{pmatrix} \right) &= \Lambda_q^\perp \left(\begin{pmatrix} -2 \\ 1 \end{pmatrix} \right) \\ &= \{(x, y) \in \mathbb{Z}^2 : -2x + y \equiv 0 \pmod{q}\} \end{aligned}$$

Family of random q -ary lattices

- ▶ Random q -ary lattice: sample $\mathbf{A} \in \mathcal{U}(\mathbb{Z}_q^{m \times n})$, and consider $\Lambda_q(\mathbf{A})$

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Solving \mathbf{X} in any lattice of rank m	\geq	Solving \mathbf{X} with non-negligible prob. in a random q -ary lattice
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Worst-case to average-case reduction

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Worst-case to average-case reduction

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The SIS problem

Notations: q, B integers, $1 \leq B \ll q$, $\mathbb{Z}_q := \mathbb{Z}/q\mathbb{Z}$

SIS (Short Integer Solution) [Ajt96]

Parameters: B and q

Problem: Given $A \leftarrow \mathcal{U}(\mathbb{Z}_q^{m \times n})$ (with $n \log q < m$)

Find $x \in \mathbb{Z}^m$ s.t. $x^T A = 0 \pmod q$ with $\|x\| \leq B$ and $x \neq 0$.

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ISIS (Inhomogeneous Short Integer Solution) [Ajt96]

Parameters: B and q

Problem: Given $A \leftarrow \mathcal{U}(\mathbb{Z}_q^{m \times n})$, $y \leftarrow \mathcal{U}(\mathbb{Z}_q^n)$

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Solving approx-CVP
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Solving ISIS
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(I)SIS is as hard as worst-case lattice problems

Theorem [Ajt96]

For any $m = \text{poly}(n)$ and $B > 0$ and sufficiently large $q \geq B \cdot \text{poly}(n)$, it holds that solving SIS is at least as hard as solving γ -SIVP on arbitrary n -dimensional lattice, for some approximation factor $\gamma = B \cdot \text{poly}(n)$.

(SIVP = shortest independent vectors problems.

Objective: find n short linearly independent vectors in the lattice)

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- ▶ the **poly** quantities have been improved in more recent works
- ▶ for typical parameters: $\text{SIS} \cong \text{ISIS}$
- ▶ see [Pei16] for a survey

SIS is a lattice problem

SIS (Short Integer Solution)

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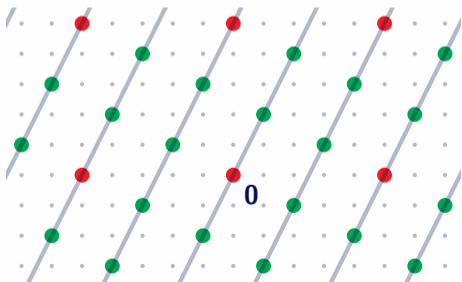
Find $\mathbf{x} \in \mathbb{Z}^m$ with $\|\mathbf{x}\| \leq B$ and $\mathbf{x} \neq \mathbf{0}$ s.t. $\mathbf{x}^T \mathbf{A} = \mathbf{0} \pmod{q}$.

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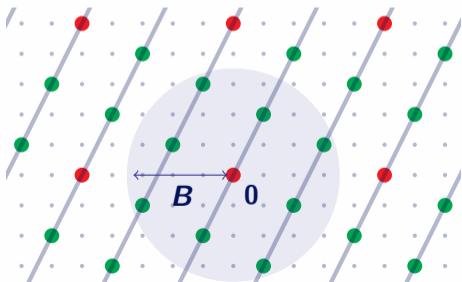
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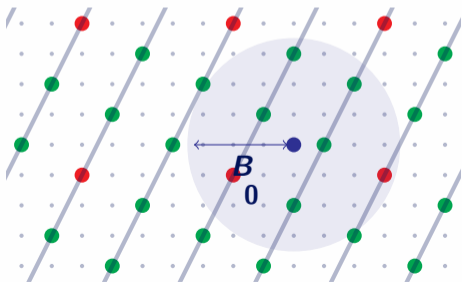
Average-case approx-SVP problem

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Average-case approx-CVP problem

Trapdoor basis

Lemma [Ajt99]

One can efficiently create a uniform SIS lattice $\Lambda_q^\perp(\mathbf{A})$ together with a short basis of it.

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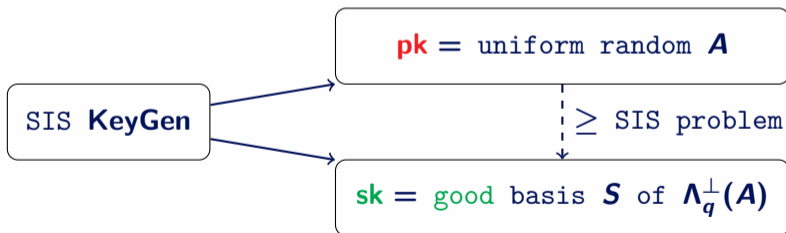
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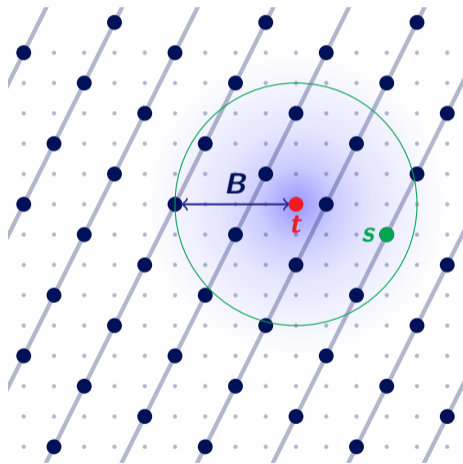
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Hash-and-sign signature scheme from SIS

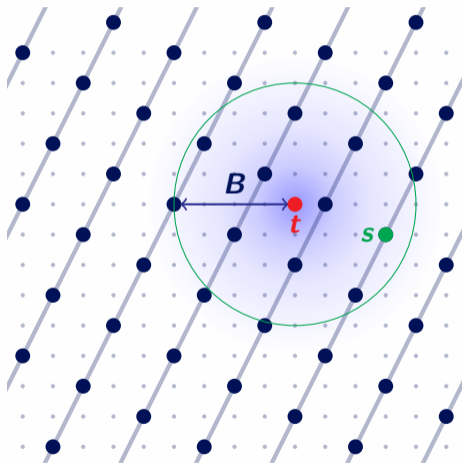


Sign: hash message to $t \in \mathbb{Z}_q^m$,

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Verify: $s \in \Lambda_q^\perp(\mathbf{A}) \wedge \|t - s\| \leq B$

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Security proof

key-recovery \geq SIS problem

signature forgery \geq ISIS problem

(assuming no leakage from sampling
can be proven in Random Oracle Model) .

Signature scheme based on hard
average-case lattice problem

The LWE problem

LWE (Learning With Errors) [Reg05]

Sample $\mathbf{A} \leftarrow \mathcal{U}(\mathbb{Z}_q^{m \times n})$, $\mathbf{s} \leftarrow \mathcal{U}(\mathbb{Z}_q^n)$ and $\mathbf{e} \leftarrow \mathcal{U}(\{-B, \dots, B\}^m)$

Given \mathbf{A} and \mathbf{b} , where $\mathbf{b} := \mathbf{A}\mathbf{s} + \mathbf{e} \bmod q$

Recover \mathbf{s} or \mathbf{e}

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Remark. Sometimes \mathbf{s} is small in \mathbb{Z}_q (not uniform)

- ▶ this is (almost) equivalent
- ▶ prove it (*hint*: you are allowed to change m)

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Solving BDD in **any** lattice of rank m \gtrsim Solving LWE with non-negligible probability \gtrsim Solving approx-SVP in **any** lattice of rank n quantumly!

LWE is quantumly as hard as worst-case lattice problems

Theorem [Reg05]

For any $m = \text{poly}(n)$, modulus $q \leq 2^{\text{poly}(n)}$ and $B \geq 2\sqrt{n}$, solving LWE is at least as hard as quantumly solving γ -SIVP on arbitrary n -dimensional lattice, for some approximation factor $\gamma = \tilde{O}(n \cdot q/B)$.

🧠 the reduction is for a variant of LWE where s and e are sampled from a discrete Gaussian distribution of parameter B 🧠

[Reg05] Regev. On lattices, learning with errors, random linear codes, and cryptography. STOC.

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Remark: the reduction can be made fully classical [Pei09, BLPRS13]

[Pei09] Peikert. Public-key cryptosystems from the worst-case shortest vector problem. STOC.

[BLPRS13] Brakerski, Langlois, Peikert, Regev, and Stehlé. Classical hardness of learning with errors. STOC

LWE is a lattice problem

$$\text{LWE instance } \left(\boxed{A}, \boxed{b} = \boxed{A} \boxed{s} + \boxed{e} \pmod{q} \right), \text{ } e \text{ small}$$

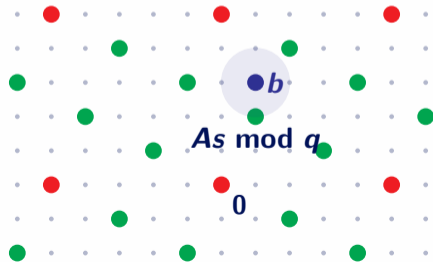
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BDD

target $b = v + e$

lattice $\Lambda_q(A)$

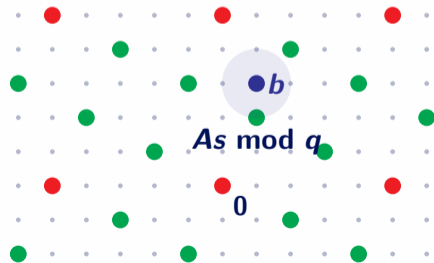


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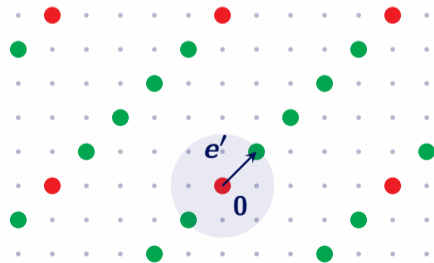
lattice $\Lambda_q(A)$



uSVP

lattice $\Lambda_q \left(\begin{pmatrix} A & b \\ 0_n & 1 \end{pmatrix} \right)$

contains short $e' := (e^\perp, 1)^\perp$



Decision variant of LWE (1)

decision-LWE (1)

Sample $A \leftarrow \mathcal{U}(\mathbb{Z}_q^{m \times n})$, $s \leftarrow \mathcal{U}(\mathbb{Z}_q^n)$ and $e \leftarrow \mathcal{U}(\{-B, \dots, B\}^m)$

Given A and b , where $b := A s + e \bmod q$ or $b \leftarrow \mathcal{U}(\mathbb{Z}_q^m)$

Guess whether b is uniform or not.

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\Rightarrow decision problems can be easier to use for crypto

Decision variant of LWE (2)

if dec-LWE is hard: $\left(\boxed{A}, \boxed{b} = \boxed{A} \boxed{s} + \boxed{e} \bmod q \right) \approx \left(\boxed{A}, \boxed{b} \right)$

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For a random q -ary lattice:

BDD target $b \approx$ uniform random target

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random q -ary lattice with planted short vector
 \approx
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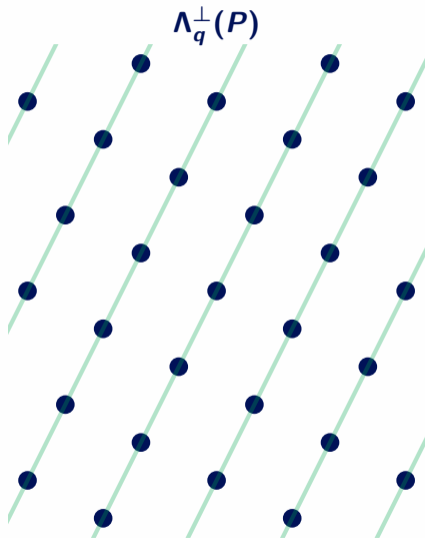
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useful in security proofs!

1 bit encryption scheme from LWE

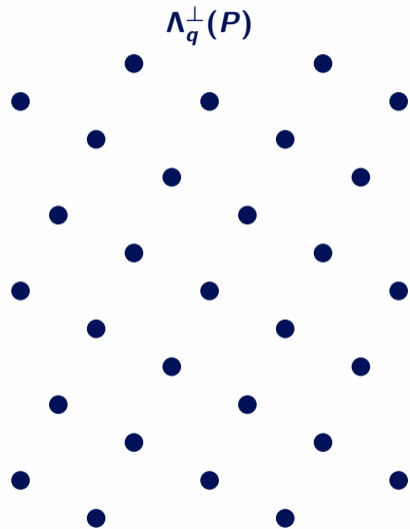


KeyGen:

$$\mathbf{pk} = (A, b = As + e), P = \begin{pmatrix} A & b \\ 0 & 1 \end{pmatrix}.$$

$$\mathbf{sk} = e, \text{ short vector } \begin{pmatrix} e \\ 1 \end{pmatrix} \in \Lambda_q(P).$$

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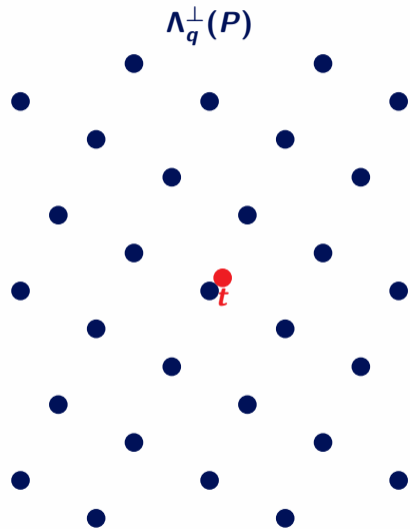
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Encrypt(m, \mathbf{pk}):

Generate: BDD instance $t = v + e'$ in $\Lambda_q^\perp(P)$

$$\text{Output: } c = t + \left\lfloor \frac{q}{2} \right\rfloor \cdot m \cdot (0, \dots, 0, 1)^\top.$$

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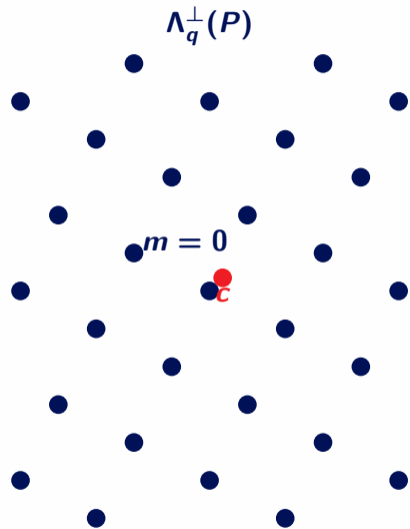
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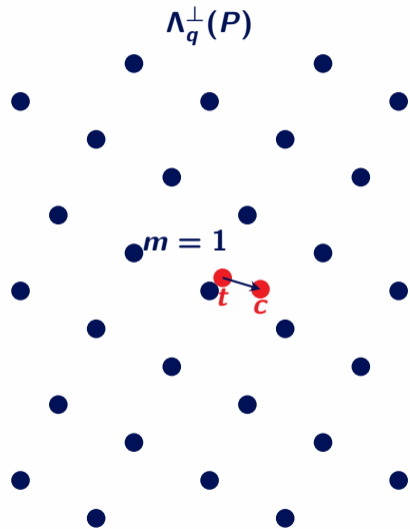
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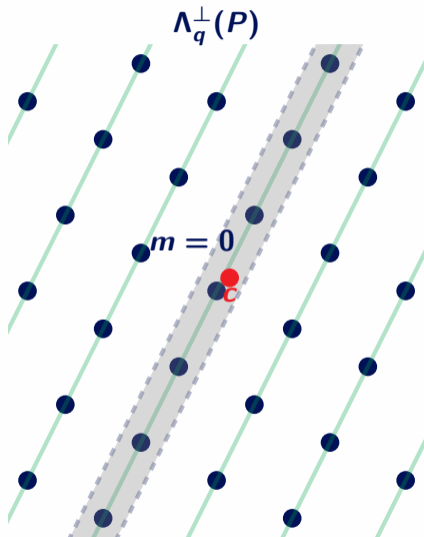
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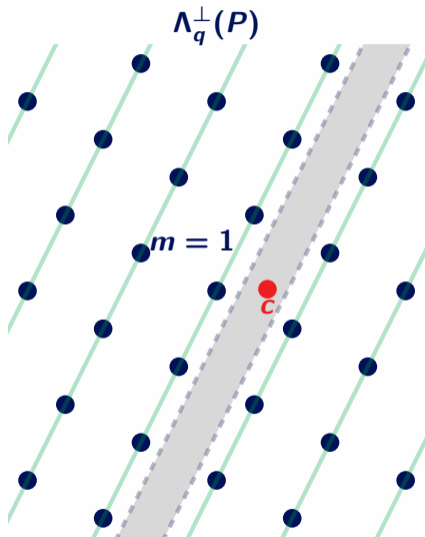
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1 bit encryption scheme from LWE

security proof

dec-LWE \Rightarrow

$\Lambda_q^\perp(P) \approx$ random q -lattice
(no information about sk)

dec-LWE \Rightarrow

$t \approx$ uniform random target
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(no information about m)

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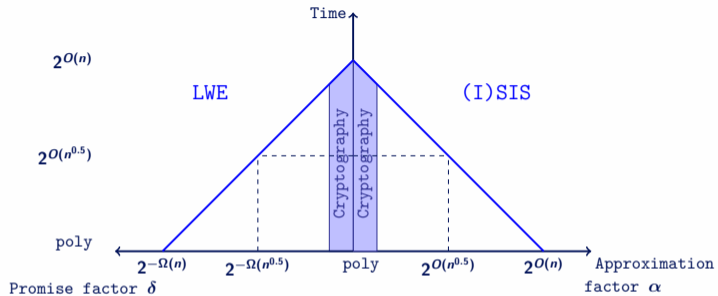
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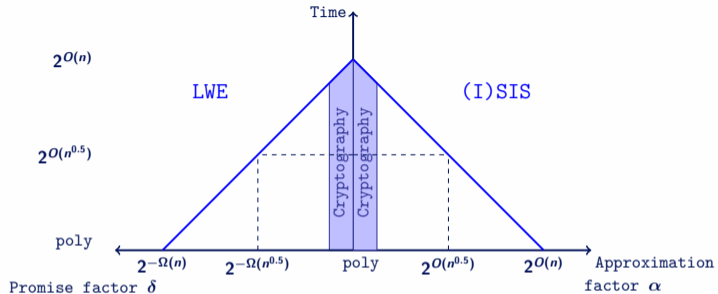
(I)SIS $\overset{\sim}{\longleftrightarrow}$ average-case SVP/CVP

LWE $\overset{\sim}{\longleftrightarrow}$ average case BDD/uSVP

LWE vs SIS

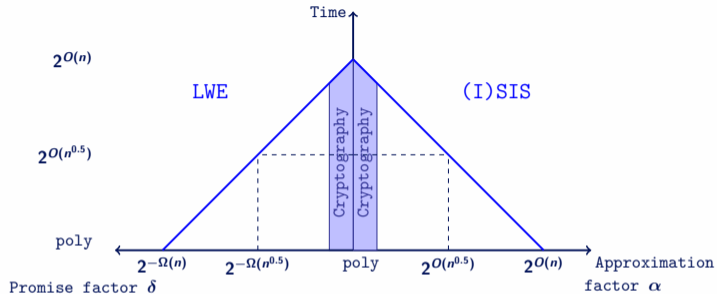


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Exercise

Prove that decision-LWE \leq SIS

Hint: See decryption of LWE encryption scheme

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- ▶ useful survey [Pei16]

Algebraic lattices

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Idea: add (algebraic) structure

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

Coefficient embedding: $\Sigma :$

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(this is also true for non principal ideals)

(we can replace Σ by σ and \mathbb{Z}^d by \mathbb{C}^d)

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Storage: n^2 coefficients $\rightarrow n$

Time: $O(n^2) \rightarrow O(n \log(n))$

(fast polynomial multiplication via FFT)

Module lattices

(Free) module:

$$M = \{B \cdot x \mid x \in \mathcal{O}_K^k\} \text{ for some matrix } B \in \mathcal{O}_K^{k \times k} \text{ with } \det_K(B) \neq 0$$

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Modules vs ideals

Ideal = Module of rank 1
(principal ideal = free module of rank 1)

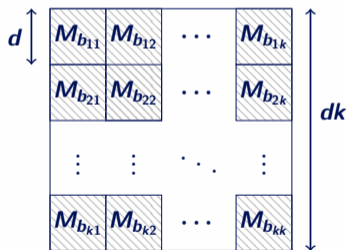
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(principal ideal = free module of rank 1)

In $K = \mathbb{Q}[X]/(X^d + 1)$:

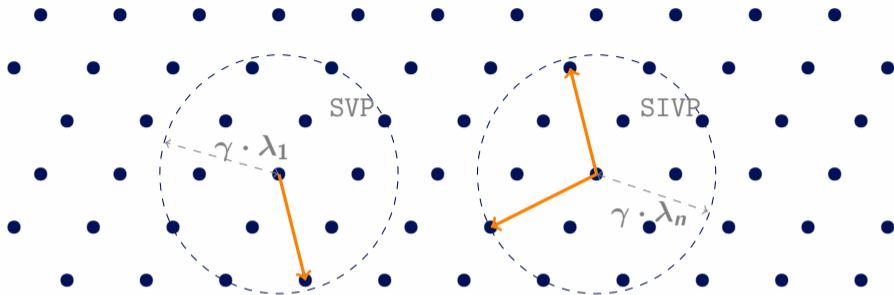
$$M_a = \begin{pmatrix} a_1 & -a_d & \cdots & -a_2 \\ a_2 & a_1 & \cdots & -a_3 \\ \vdots & \ddots & \ddots & \vdots \\ a_d & a_{d-1} & \cdots & a_1 \end{pmatrix}$$

basis of a
principal ideal lattice



basis of a free module lattice
of rank k

Algorithmic problems



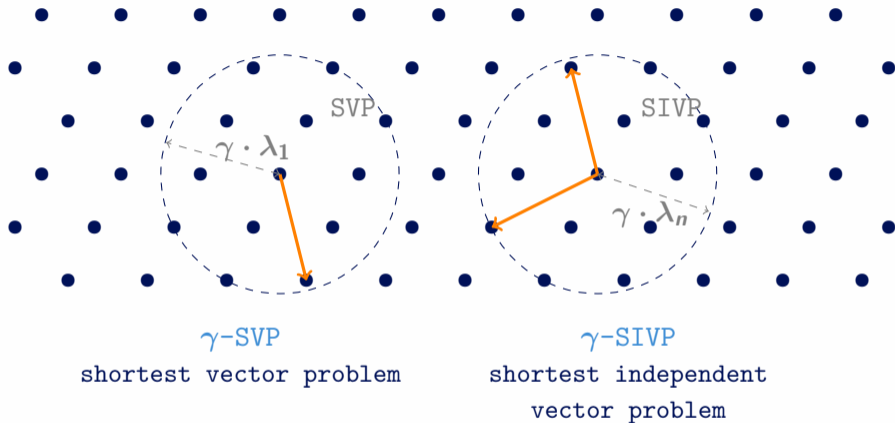
γ -SVP

shortest vector problem

γ -SIVP

shortest independent
vector problem

Algorithmic problems

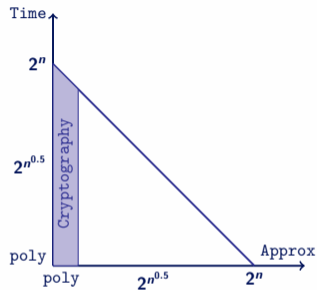


Notations:

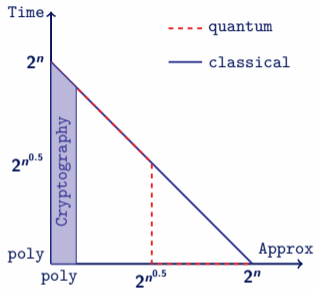
- ▶ id-X = problem X restricted to ideal lattices
 - ▶ mod-X_k = problem X restricted to module lattices of rank k
- (worst-case: we want algorithms for all ideal/module lattices)

Hardness of module SVP

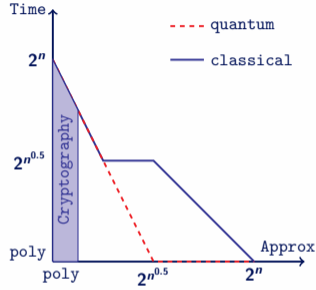
Asymptotics:



SVP and mod-SVP_k
($k \geq 2$)



id-SVP [CDW17]
(in cyclotomic fields)



id-SVP [PHS19, BR20]
(with $2^{O(n)}$ pre-processing)

[CDW17] Cramer, Ducas, Wesolowski. Short stickelberger class relations and application to ideal-SVP. Eurocrypt.

[PHS19] Pellet-Mary, Hanrot, Stehlé. Approx-SVP in ideal lattices with pre-processing. Eurocrypt.

[BR20] Bernard, Roux-Langlois. Twisted-PHS: using the product formula to solve approx-SVP in ideal lattices. AC.

(search) mod-LWE_k

Parameters: q and B

Problem: Sample

- ▶ $A \leftarrow \mathcal{U}((\mathcal{O}_K/q\mathcal{O}_K)^{m \times k})$
- ▶ secret $s \in (\mathcal{O}_K/q\mathcal{O}_K)^k$
- ▶ error $e \in \mathcal{O}_K^m$ with coefficients in $\{-B, \dots, B\}$

Given A and $b = A \cdot s + e \bmod q$, recover s

(size of s and e can be defined using Σ or σ)

Ring and Module-LWE

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$$\text{RLWE} = \text{mod-LWE}_1$$

mod-LWE vs mod-SIVP

$$\text{mod-uSVP}_{m+1} \geq \text{mod-BDD}_m \geq \text{mod-LWE}_k \underset{\text{quantumly!}}{\geq} \text{mod-SIVP}_k$$

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- ▶ as small as possible
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- ▶ but so that the closest point to \mathbf{b} is \mathbf{As}
- ▶ $m = k$ is not sufficient
- ▶ $m = k + 1$ might be sufficient depending on B and q
 - ▶ we need roughly $m = k \cdot \frac{\log(q)}{\log(q/B)}$
 - ▶ for $k = 1$, $m = 2$ is possible if $B \lesssim \sqrt{q}$

(search) NTRU

Parameters: $q \geq B > 1$

Objective: Sample $f, g \in \mathcal{O}_K$ with coefficients in $\{-B, \dots, B\}$.

Given $h = f \cdot g^{-1} \bmod q$, recover (f, g)

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dec-NTRU

Parameters: q, B Objective: distinguish between h as above and h uniform in $\mathcal{O}_K / (q\mathcal{O}_K)$

NTRU as a lattice

Recall: $h = f \cdot g^{-1} \bmod q$

Definition (NTRU Lattice)

$$\mathcal{L}^{h,q} := \{(a, b) \in R^2 : h \cdot b = a \bmod q\}$$

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$$\text{bad basis} = \begin{pmatrix} q & h \\ 0 & 1 \end{pmatrix}, \quad \text{good basis} = \begin{pmatrix} f & F \\ g & G \end{pmatrix}$$

Two regimes of NTRU

If $\|(f, g)\| \geq \text{poly}(\log n) \cdot \text{gh}(\mathcal{L}^{h,q})$

If $\|(f, g)\| \leq \text{gh}(\mathcal{L}^{h,q})$

Two regimes of NTRU

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[SS11] Stehlé and Steinfeld. Making NTRU as secure as worst-case problems over ideal lattices. Eurocrypt.

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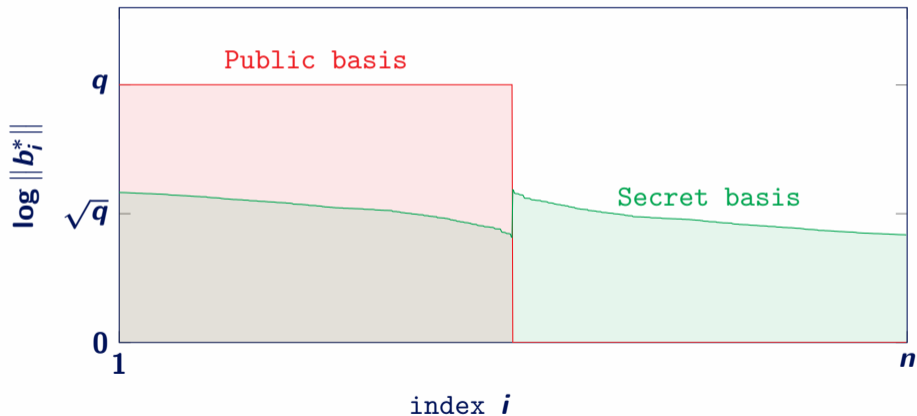
uSVP regime \Rightarrow short structured basis
 \Rightarrow efficient encryption/signature scheme
(e.g. NTRUEncrypt, NTRUSign, FALCON)

[SS11] Stehlé and Steinfeld. Making NTRU as secure as worst-case problems over ideal lattices. Eurocrypt.

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NTRU public vs secret basis

public and secret bases generated from the NTRU problem



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- ▶ Most worst-case to average-case reductions still apply
- ▶ Ideal lattices = rank 1 modules can be vulnerable
- ▶ NIST candidates (e.g. Kyber, Dilithium, Falcon) use rank ≥ 2
(seems safe so far, but arguably their weakest point)

Conclusion on lattice-based crypto

Advantages:

- ▶ many reductions (worst-case to average-case, search to decision, ...)
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Thank you