

Improved Reduction from BDD to μ SVP

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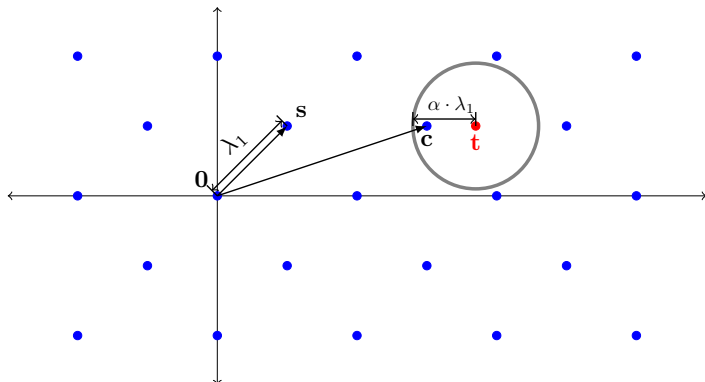


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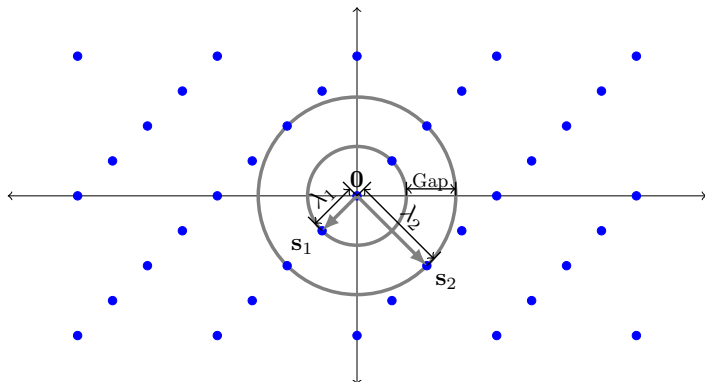
Bounded Distance Decoding (BDD) and unique Shortest Vector Problem (USVP)



Bounded Distance Decoding for $\alpha \geq 0$ (BDD_α)

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$, a vector $\mathbf{t} \in \mathbb{Q}^n$ such that $\text{dist}(\mathbf{t}, \mathcal{L}(\mathbf{B})) \leq \alpha \cdot \lambda_1(\mathbf{B})$.

Output: a lattice vector $\mathbf{c} \in \mathcal{L}(\mathbf{B})$ closest to \mathbf{t} .



Unique Shortest Vector Problem for $\gamma \geq 1$ (uSVP $_{\gamma}$)

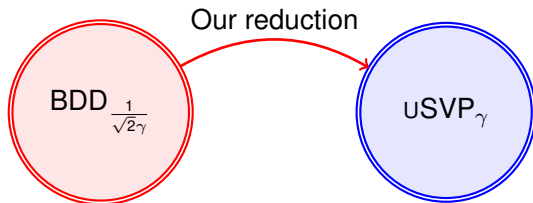
Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$ such that $\lambda_2(\mathcal{L}(\mathbf{B})) \geq \gamma \cdot \lambda_1(\mathcal{L}(\mathbf{B}))$.

Output: a non-zero vector $\mathbf{s}_1 \in \mathcal{L}(\mathbf{B})$ of norm $\lambda_1(\mathcal{L}(\mathbf{B}))$.

Improved reduction from BDD to uSVP

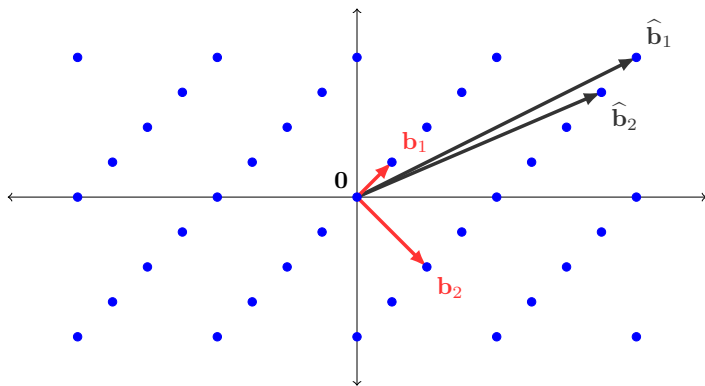
For $1 \leq \gamma \leq \text{poly}(n)$, we have

$$\text{BDD}_{1/(\sqrt{2}\gamma)} \leq \text{uSVP}_\gamma.$$



- Background
- The Lyubashevsky and Micciancio reduction and its limitation
- New reduction:
 - lattice sparsification.
 - reduction for $\gamma = 1$.
 - sphere packing.
- Open problems

Lattices

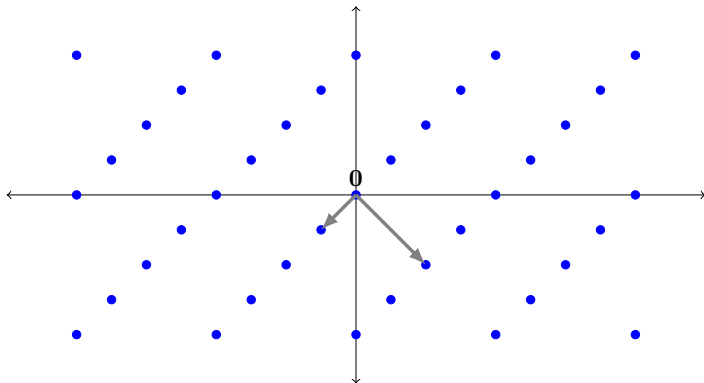


A definition of lattice

Given $\mathbf{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\} \subseteq \mathbb{Q}^m$ a set of linear independent vectors, the lattice \mathcal{L} spanned by the \mathbf{b}_i 's is

$$\mathcal{L}(\mathbf{B}) = \left\{ \sum_{i \in [n]} u_i \mathbf{b}_i : \mathbf{u} \in \mathbb{Z}^n \right\}.$$

Lattice Minima

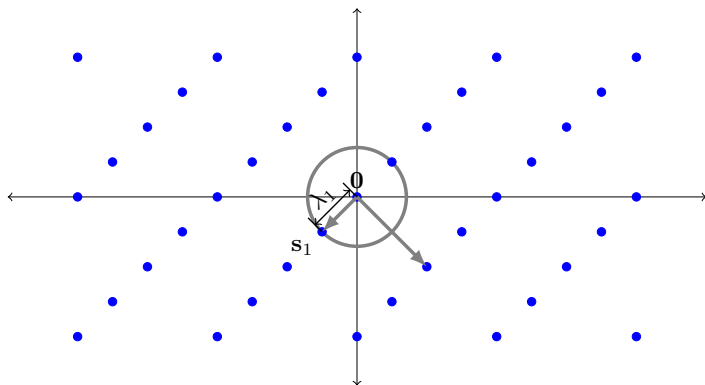


Lattice minimum

Given a lattice \mathcal{L} , the i -th minimum of \mathcal{L} is defined as:

$$\lambda_i(\mathcal{L}) = \inf\{r : \dim(\text{span}(\mathcal{L} \cap \mathcal{B}(\mathbf{0}, r))) \geq i\}.$$

Lattice Minima – first minimum

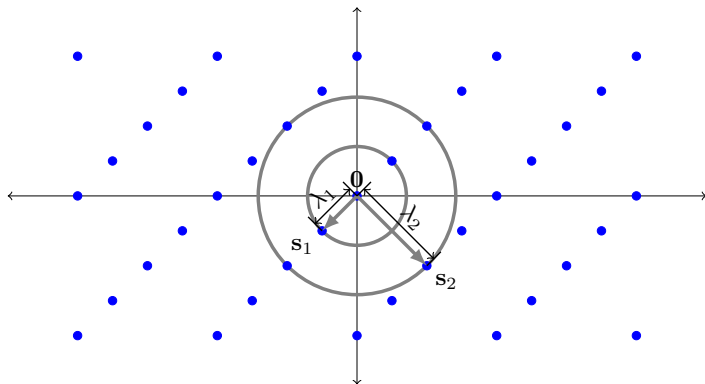


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Lattice Minima – second minimum

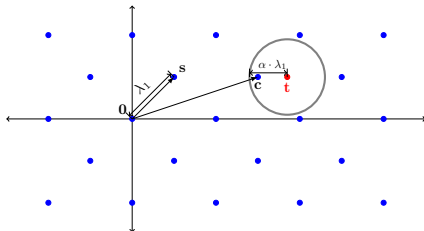


Lattice minimum

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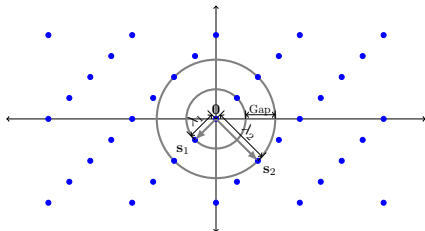
$$\lambda_i(\mathcal{L}) = \inf\{r : \dim(\text{span}(\mathcal{L} \cap \mathcal{B}(\mathbf{0}, r))) \geq i\}.$$

Why is BDD interesting?



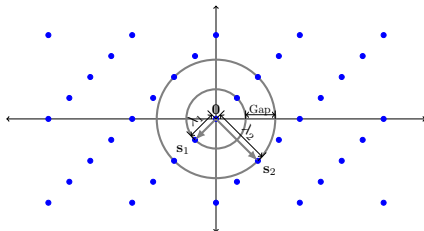
- ▶ In **cryptography**:
 - ▶ Learning With Error (LWE) problem serves as a security foundation.
 - ▶ LWE is an average-case variant of BDD.
- ▶ In **communication theory** – white Gaussian noise channel:
 - ▶ Wifi, mobile phone *etc*;
 - ▶ View message as a lattice point, Gaussian noise is added in channel transmission, decoding is solving BDD.

Why is uSVP interesting?



- ▶ Best known algorithm (especially in practice) for solving BDD is via solving uSVP:
 - ▶ First, reduce BDD to uSVP.
 - ▶ Second, solve **uSVP** by lattice reduction, *e.g.*, LLL and BKZ.

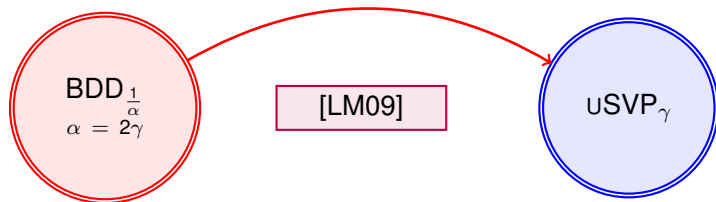
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BDD $\frac{1}{\text{poly}(n)}$ and uSVP $\text{poly}(n)$ are hard;
Best known algorithm takes **exponential** time in dimension n .

Prior works on BDD to uSVP



- Slightly improved for some α , Liu *et al*, 2014; Galbraith; Micciancio, 2015.

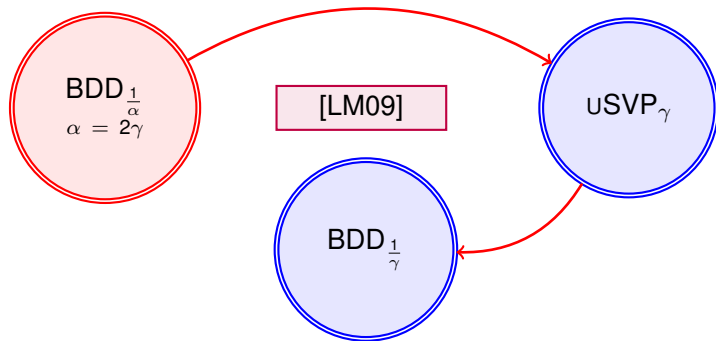
[LM09]: V. Lyubashevsky and D. Micciancio. On bounded distance decoding, unique shortest vectors, and the minimum distance problem, CRYPTO, 2009.

[LWXZ14]: M. Liu, X. Wang, G. Xu and X. Zheng. A note on BDD problems with λ_2 -gap. Inf. Process. Lett., 2014.

[Ga15]: Private communication, 2015.

[Mi15]: Private communication, 2015.

Prior works on BDD to uSVP



There is a factor **2** to be improved.

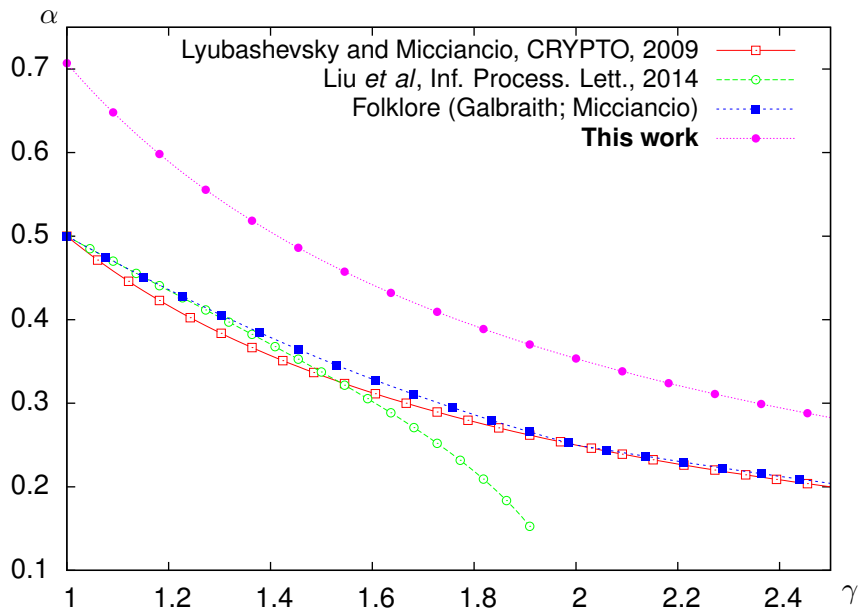
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Comparison with prior works

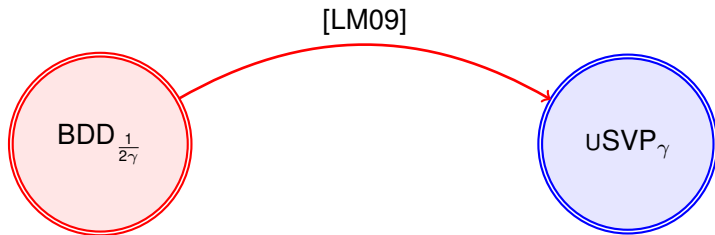


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The Lyubashevsky and Micciancio reduction

For any $\gamma \geq 1$, we have

$$\text{BDD}_{1/(2\gamma)} \leq \text{USVP}_{\gamma}.$$

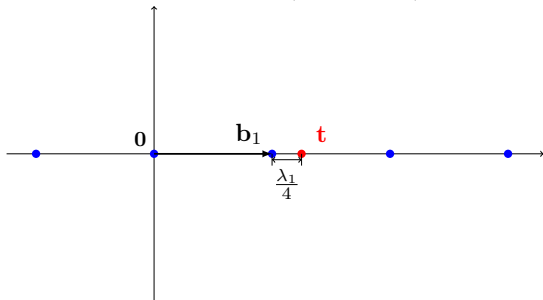


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A simple case

- ▶ $\text{BDD}_{1/4}$ instance: $(\mathcal{L}(\mathbf{b}_1), \mathbf{t})$.

$$\text{BDD}_{\frac{1}{2\gamma}}$$
$$\left(\boxed{\mathbf{B}}, \boxed{\mathbf{t}} \right)$$
$$(n = 1, \gamma = 2)$$



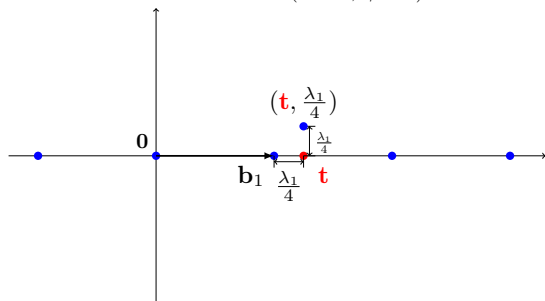
A simple case

- ▶ Lift vector \mathbf{t} into a higher dimension space by $\lambda_1(\mathcal{L}(\mathbf{b}_1))/4$.

$$\text{BDD}_{\frac{1}{2\gamma}}$$

$$\left(\boxed{\mathbf{B}}, \boxed{\mathbf{t}} \right)$$

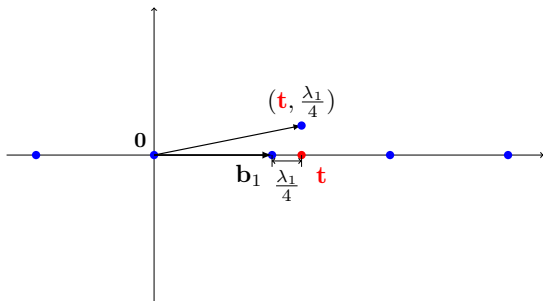
$$(n = 1, \gamma = 2)$$



A simple case

- ▶ Kannan embedding.

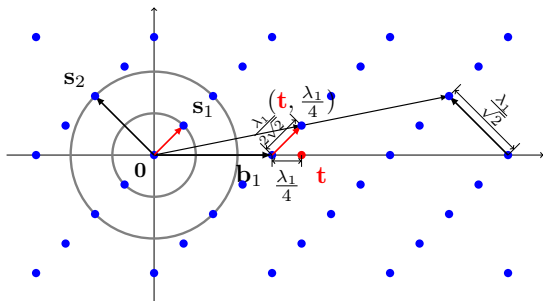
$$\begin{array}{ccc} \text{BDD}_{\frac{1}{2\gamma}} & & \text{USVP}_{\gamma'} \\ \left(\begin{array}{c} \boxed{\mathbf{B}} \\ \boxed{\mathbf{t}} \end{array} \right) & \xrightarrow[\text{embedding}]{\text{Kannan}} & \mathbf{B}' = \begin{bmatrix} \boxed{\mathbf{B}} & \boxed{\mathbf{t}} \\ \boxed{0} & \boxed{\frac{\lambda_1}{2\gamma}} \end{bmatrix} \\ (n=1, \gamma=2) & & (n'=2, \gamma'=2) \end{array}$$



A simple case

- ▶ Finally, we obtain a uSVP instance with $\lambda'_2 = 2\lambda'_1$.

$$\begin{array}{ccc}
 \text{BDD}_{\frac{1}{2\gamma}} & & \text{uSVP}_{\gamma'} \\
 \left(\begin{array}{c} \boxed{\mathbf{B}} \\ \boxed{\mathbf{t}} \end{array} \right) & \xrightarrow[\text{embedding}]{\text{Kannan}} & \mathbf{B}' = \begin{bmatrix} \boxed{\mathbf{B}} & \boxed{\mathbf{t}} \\ \boxed{0} & \boxed{\frac{\lambda_1}{2\gamma}} \end{bmatrix} \\
 (n = 1, \gamma = 2) & & (n' = 2, \gamma' = 2)
 \end{array}$$



Algorithm for solving BDD

Version 1. The $\text{BDD}_{1/(2\gamma)}$ to USVP_γ reduction.

Input: a basis $\mathbf{B} = \{\mathbf{b}_i\}_{i \in [n]}$, and a target point \mathbf{t} .

Output: a lattice point \mathbf{c} such that $\|\mathbf{c} - \mathbf{t}\| = \text{dist}(\mathbf{t}, \mathcal{L})$.

0. Define

$$\mathbf{B}' = \begin{pmatrix} \mathbf{B} & \mathbf{t} \\ \mathbf{0} & \frac{\lambda_1(\mathcal{L}(\mathbf{B}))}{2\gamma} \end{pmatrix}.$$

1. Run the USVP_γ solver on input \mathbf{B}' . Let $\mathbf{s}' = \begin{pmatrix} \mathbf{s}'_1 \\ \mathbf{s}'_2 \end{pmatrix}$ be its output.

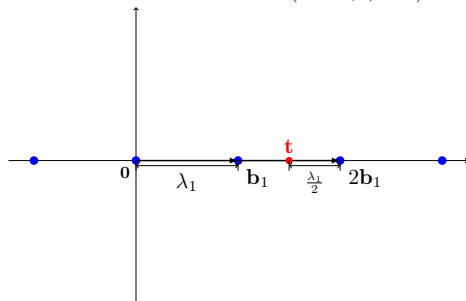
2. Output $\mathbf{t} - \mathbf{s}'_1$.

$$\begin{array}{ccc} \text{BDD}_{\frac{1}{2\gamma}} & & \text{USVP}_{\gamma'} \\ \left(\begin{array}{c} \boxed{\mathbf{B}} \\ \boxed{\mathbf{t}} \end{array} \right) & \xrightarrow[\text{embedding}]{\text{Kannan}} & \mathbf{B}' = \begin{bmatrix} \boxed{\mathbf{B}} & \boxed{\mathbf{t}} \\ \boxed{\mathbf{0}} & \boxed{\frac{\lambda_1}{2\gamma}} \end{bmatrix} \\ (n=1, \gamma=2) & & (n'=2, \gamma'=2) \end{array}$$

Limiting case of the reduction

- ▶ BDD_{1/2} instance: $(\mathcal{L}(\mathbf{b}_1), \mathbf{t})$.

$$\text{BDD}_{\frac{1}{2\gamma}}$$
$$\left(\boxed{\mathbf{B}}, \boxed{\mathbf{t}} \right)$$
$$(n = 1, \gamma = 1)$$



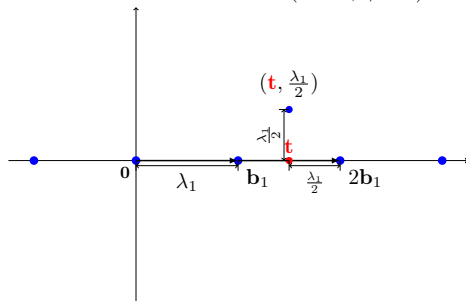
Limiting case of the reduction

- ▶ Lift vector \mathbf{t} into a higher dimension space by $\lambda_1(\mathcal{L}(\mathbf{b}_1))/2$.

$$\text{BDD}_{\frac{1}{2\gamma}}$$

$$\left(\begin{array}{c} \boxed{\mathbf{B}} \\ \boxed{\mathbf{t}} \end{array} \right)$$

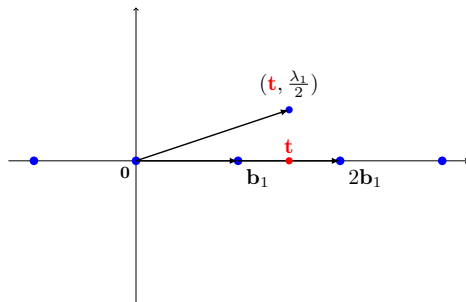
$$(n = 1, \gamma = 1)$$



Limiting case of the reduction

- ▶ Kannan embedding.

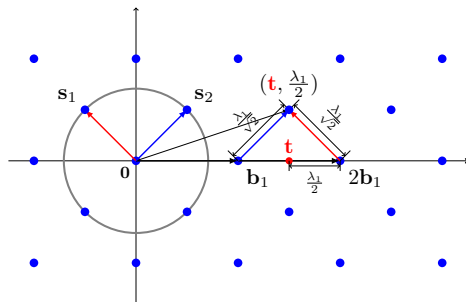
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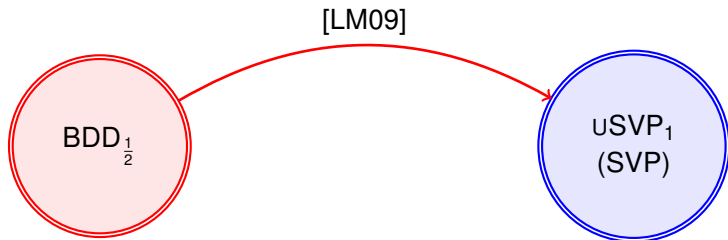
Limiting case of the reduction

- ▶ We are at the limit: $\lambda'_1 = \lambda'_2$.

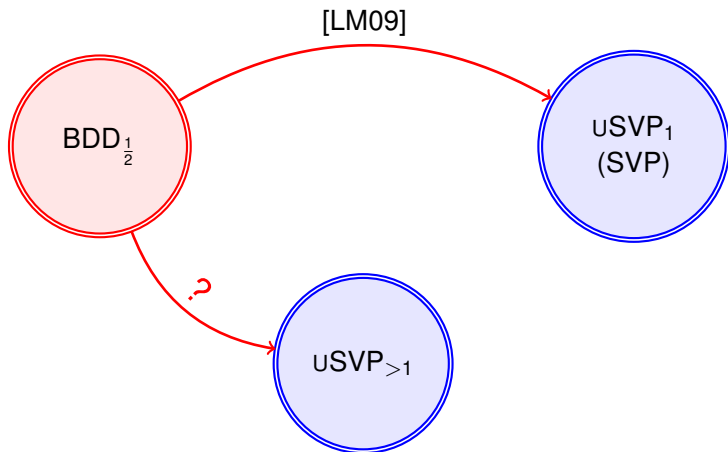
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 (n = 1, \gamma = 1) & & (n' = 2, \gamma' = 1)
 \end{array}$$



This is the best this reduction can achieve

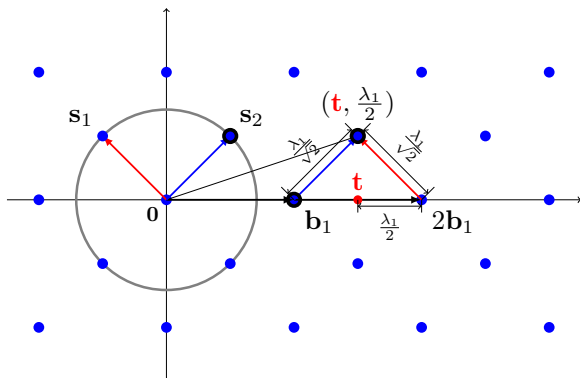


Can we improve it?



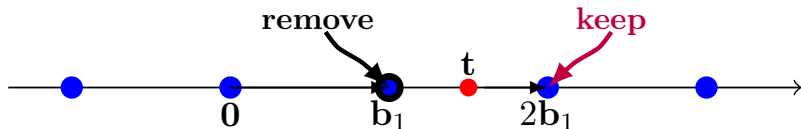
An attempt to circumvent the limitation

- ▶ Limitation in the Lyubushevsky and Micciancio reduction.



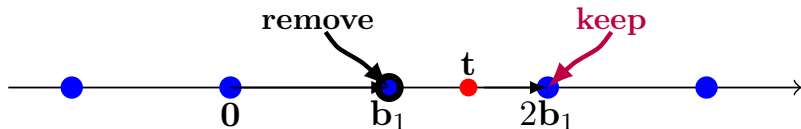
An attempt to circumvent the limitation

- ▶ A simple deterministic sparsification.
- ▶ Lattice $\mathcal{L}(\mathbf{B})$ with $\mathbf{B} = [\mathbf{b}_1]$.

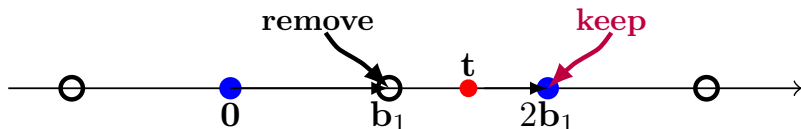


An attempt to circumvent the limitation

- ▶ A simple deterministic sparsification.
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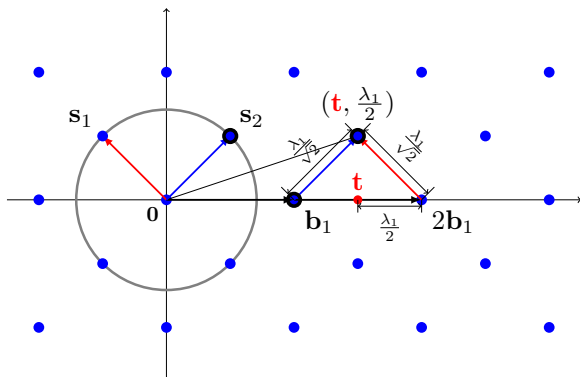


- ▶ Lattice $\mathcal{L}(\tilde{\mathbf{B}})$ with $\tilde{\mathbf{B}} = [2\mathbf{b}_1]$.



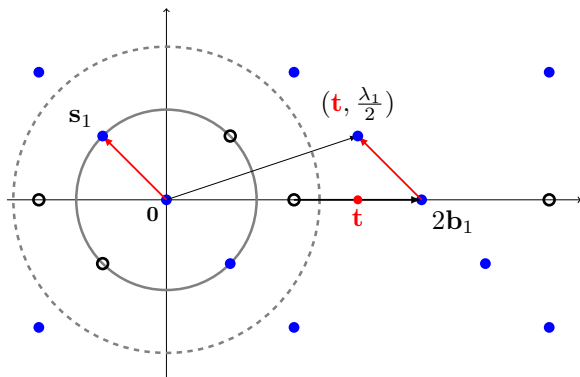
An attempt to circumvent the limitation

- ▶ Recall the limitation: $\lambda'_2 = \lambda'_1$



An attempt to circumvent the limitation

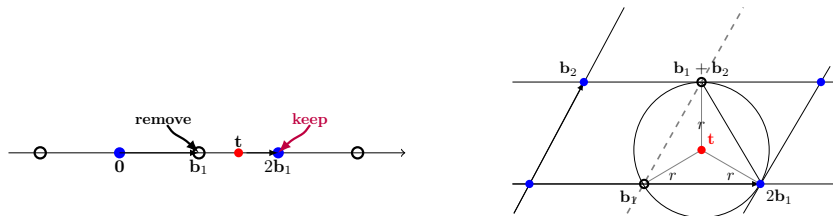
- ▶ Limitation is circumvented (for this example): $\lambda'_2 > \lambda'_1$ now!



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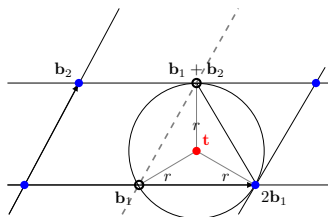
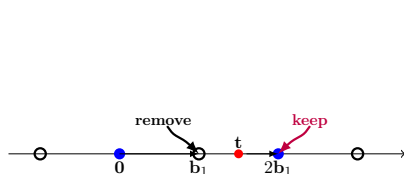
Deterministic sparsification is not enough

- ▶ Deterministic sparsification leads to a combinatorial explosion.



Deterministic sparsification is not enough

- ▶ Deterministic sparsification leads to a combinatorial explosion.



- ▶ But we want more...

- keep only closest vector to target \mathbf{t} .
- remove all other somewhat close vectors to \mathbf{t} .

Lattice Sparsification

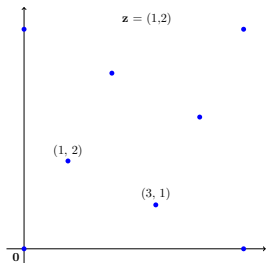
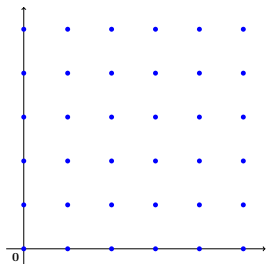
Khot's Lattice Sparsification [K03]

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

$$\mathcal{L}_{p,\mathbf{z}} = \{\mathbf{x} \in \mathcal{L}(\mathbf{B}) \mid \langle \mathbf{z}, \mathbf{B}^{-1}\mathbf{x} \rangle = 0 \pmod{p}\},$$

where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



Sparsification on lattice

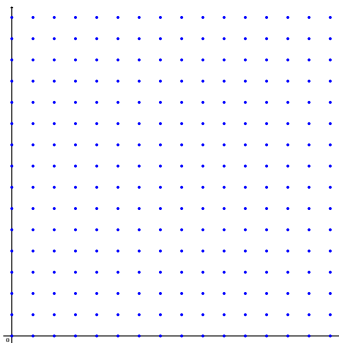
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- ▶ 1st sparsification:
 $p = 5, \mathbf{z} = (0, 0)$.

Sparsification on lattice

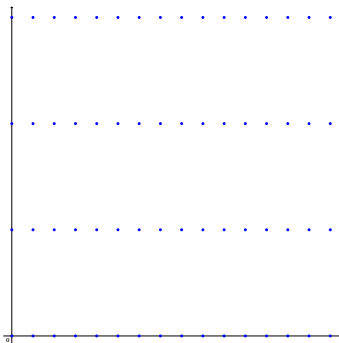
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- ▶ **2nd sparsification:**
 $p = 5, \mathbf{z} = (0, 1)$.

Sparsification on lattice

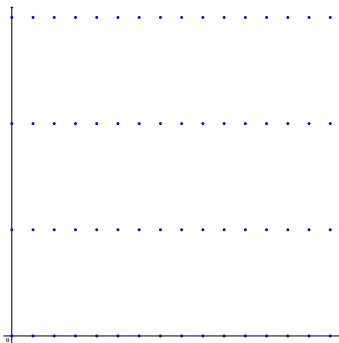
Khot's Lattice Sparsification

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where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **3rd sparsification:**
 $p = 5, \mathbf{z} = (0, 2)$.

Sparsification on lattice

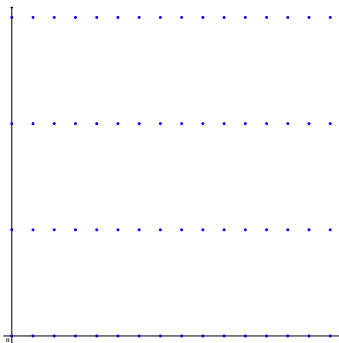
Khot's Lattice Sparsification

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- ▶ 4th sparsification:
 $p = 5, \mathbf{z} = (0, 3)$.

Sparsification on lattice

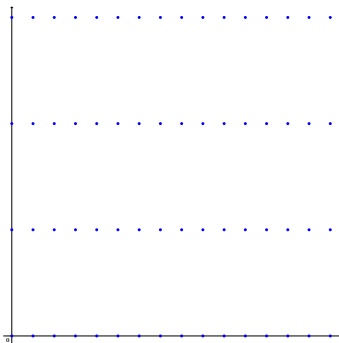
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Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

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where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **5th sparsification:**
 $p = 5, \mathbf{z} = (0, 4)$.

Sparsification on lattice

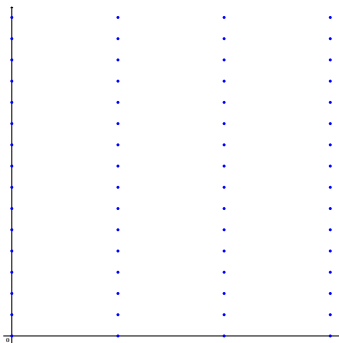
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Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

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where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **6th sparsification:**
 $p = 5, \mathbf{z} = (1, 0)$.

Sparsification on lattice

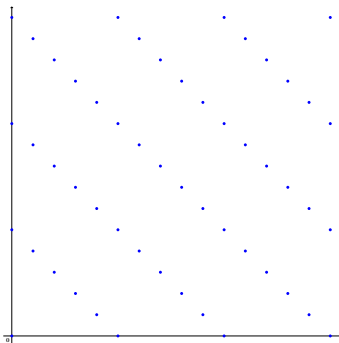
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

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- ▶ 7th sparsification:
 $p = 5, \mathbf{z} = (1, 1)$.

Sparsification on lattice

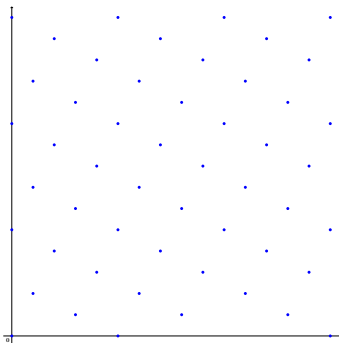
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

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where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **8th sparsification:**
 $p = 5, \mathbf{z} = (1, 2)$.

Sparsification on lattice

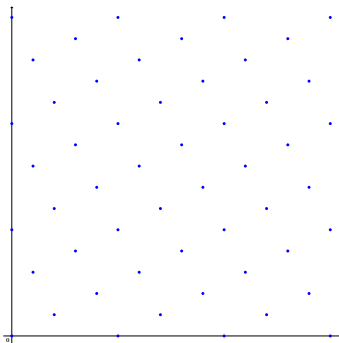
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

$$\mathcal{L}_{p,\mathbf{z}} = \{\mathbf{x} \in \mathcal{L}(\mathbf{B}) \mid \langle \mathbf{z}, \mathbf{B}^{-1}\mathbf{x} \rangle = 0 \pmod{p}\},$$

where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **9th sparsification:**
 $p = 5, \mathbf{z} = (1, 3)$.

Sparsification on lattice

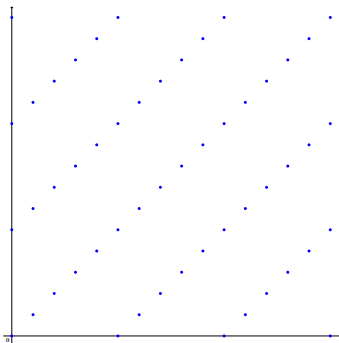
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

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where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **10th sparsification:**
 $p = 5, \mathbf{z} = (1, 4)$.

Sparsification on lattice

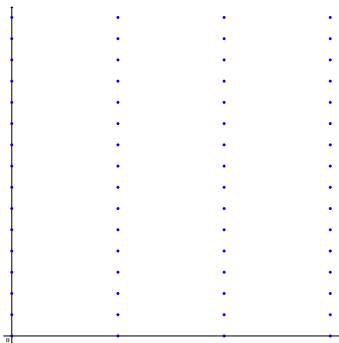
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

$$\mathcal{L}_{p,\mathbf{z}} = \{\mathbf{x} \in \mathcal{L}(\mathbf{B}) \mid \langle \mathbf{z}, \mathbf{B}^{-1}\mathbf{x} \rangle = 0 \pmod{p}\},$$

where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **11th sparsification:**
 $p = 5, \mathbf{z} = (2, 0)$.

Sparsification on lattice

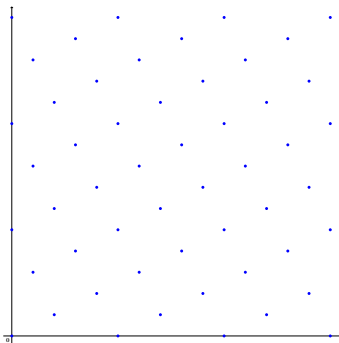
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

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where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **12th sparsification:**
 $p = 5, \mathbf{z} = (2, 1)$.

Sparsification on lattice

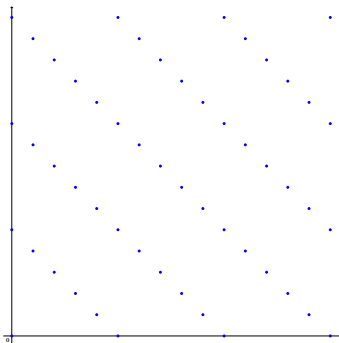
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

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where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **13th** sparsification:
 $p = 5, \mathbf{z} = (2, 2)$.

Sparsification on lattice

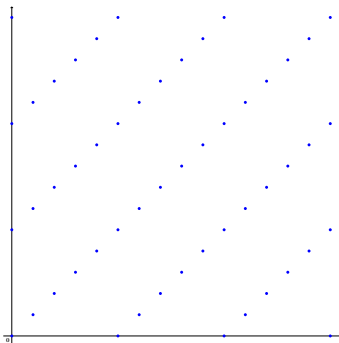
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

$$\mathcal{L}_{p,\mathbf{z}} = \{\mathbf{x} \in \mathcal{L}(\mathbf{B}) \mid \langle \mathbf{z}, \mathbf{B}^{-1}\mathbf{x} \rangle = 0 \pmod{p}\},$$

where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **14th sparsification:**
 $p = 5, \mathbf{z} = (2, 3)$.

Sparsification on lattice

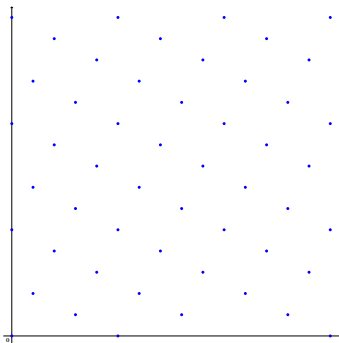
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

$$\mathcal{L}_{p,\mathbf{z}} = \{\mathbf{x} \in \mathcal{L}(\mathbf{B}) \mid \langle \mathbf{z}, \mathbf{B}^{-1}\mathbf{x} \rangle = 0 \pmod{p}\},$$

where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **15th sparsification:**
 $p = 5, \mathbf{z} = (2, 4)$.

Sparsification on lattice

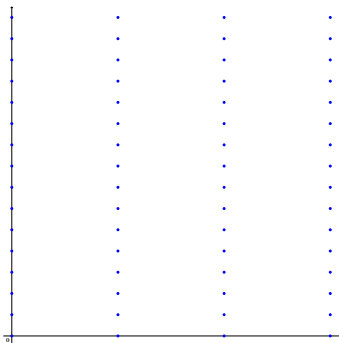
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

$$\mathcal{L}_{p,\mathbf{z}} = \{\mathbf{x} \in \mathcal{L}(\mathbf{B}) \mid \langle \mathbf{z}, \mathbf{B}^{-1}\mathbf{x} \rangle = 0 \pmod{p}\},$$

where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **16th sparsification:**
 $p = 5, \mathbf{z} = (3, 0)$.

Sparsification on lattice

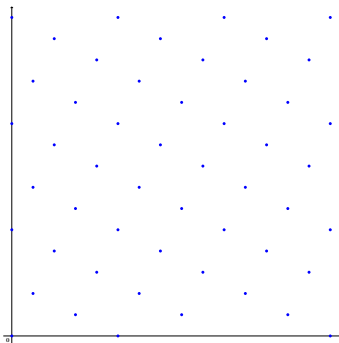
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

$$\mathcal{L}_{p,\mathbf{z}} = \{\mathbf{x} \in \mathcal{L}(\mathbf{B}) \mid \langle \mathbf{z}, \mathbf{B}^{-1}\mathbf{x} \rangle = 0 \pmod{p}\},$$

where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **17th sparsification:**
 $p = 5, \mathbf{z} = (3, 1)$.

Sparsification on lattice

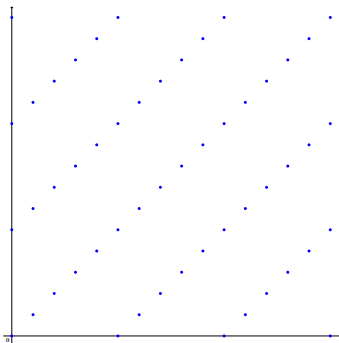
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

$$\mathcal{L}_{p,\mathbf{z}} = \{\mathbf{x} \in \mathcal{L}(\mathbf{B}) \mid \langle \mathbf{z}, \mathbf{B}^{-1}\mathbf{x} \rangle = 0 \pmod{p}\},$$

where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



► **18th** sparsification:
 $p = 5$, $\mathbf{z} = (3, 2)$.

Sparsification on lattice

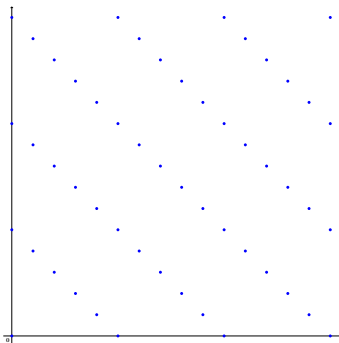
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

$$\mathcal{L}_{p,\mathbf{z}} = \{\mathbf{x} \in \mathcal{L}(\mathbf{B}) \mid \langle \mathbf{z}, \mathbf{B}^{-1}\mathbf{x} \rangle = 0 \pmod{p}\},$$

where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **19th** sparsification:
 $p = 5, \mathbf{z} = (3, 3)$.

Sparsification on lattice

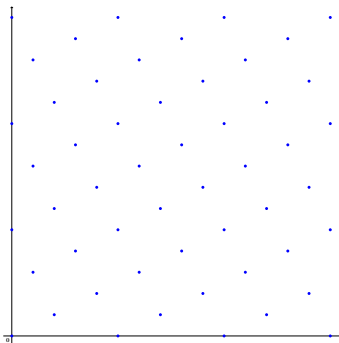
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

$$\mathcal{L}_{p,\mathbf{z}} = \{\mathbf{x} \in \mathcal{L}(\mathbf{B}) \mid \langle \mathbf{z}, \mathbf{B}^{-1}\mathbf{x} \rangle = 0 \pmod{p}\},$$

where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **20th sparsification:**
 $p = 5, \mathbf{z} = (3, 4)$.

Sparsification on lattice

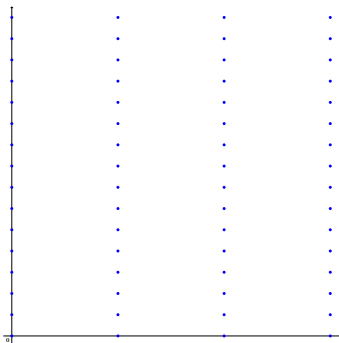
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

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- **21st** sparsification:
 $p = 5, \mathbf{z} = (4, 0)$.

Sparsification on lattice

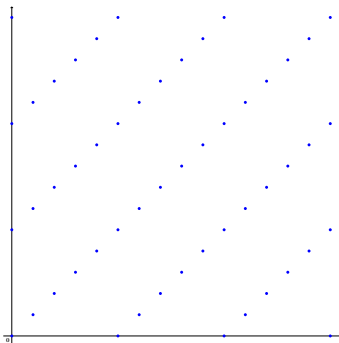
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

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where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **22nd** sparsification:
 $p = 5$, $\mathbf{z} = (4, 1)$.

Sparsification on lattice

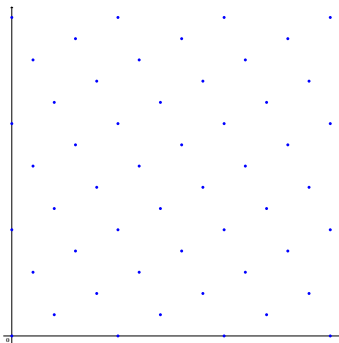
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

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where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **23rd** sparsification:
 $p = 5, \mathbf{z} = (4, 2)$.

Sparsification on lattice

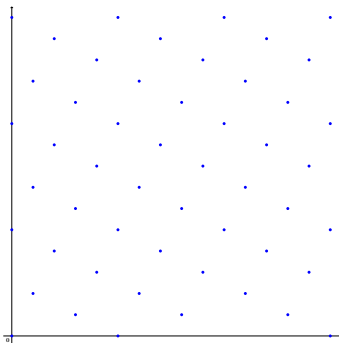
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Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

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where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **24th sparsification:**
 $p = 5, \mathbf{z} = (4, 3)$.

Sparsification on lattice

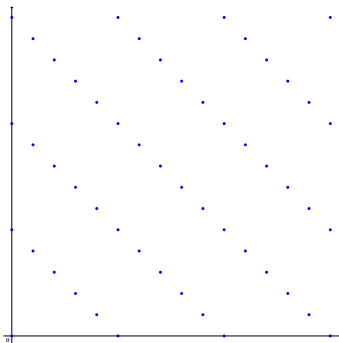
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

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where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



- ▶ **25th** sparsification:
 $p = 5, \mathbf{z} = (4, 4)$.

Sparsification on lattice

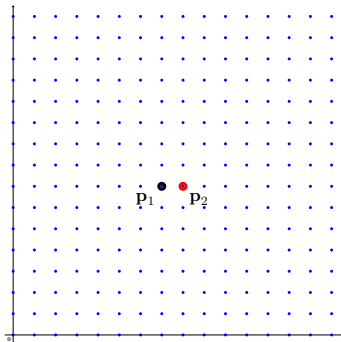
Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

$$\mathcal{L}_{p,\mathbf{z}} = \{\mathbf{x} \in \mathcal{L}(\mathbf{B}) \mid \langle \mathbf{z}, \mathbf{B}^{-1}\mathbf{x} \rangle = 0 \pmod{p}\},$$

where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



In the overall 25 sparsifications:

- ▶ \mathbf{p}_1 is in: 5 times; prob. $\frac{1}{5}$.
- ▶ \mathbf{p}_2 is out: 20 times; prob. $1 - \frac{1}{5}$.

Khot's Lattice Sparsification

Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$.

Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

$$\mathcal{L}_{p,\mathbf{z}} = \{\mathbf{x} \in \mathcal{L}(\mathbf{B}) \mid \langle \mathbf{z}, \mathbf{B}^{-1}\mathbf{x} \rangle = 0 \pmod{p}\},$$

where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.

- ▶ Each individual point is **kept** with probability $\frac{1}{p}$;
- ▶ and is **removed** with probability $1 - \frac{1}{p}$.
- ▶ Two issues:
 - ▶ The origin $\mathbf{0}$ is never removed.
 - ▶ There are dependencies among some points.

An argument of this probability result

A probabilistic argument on Khot's sparsification [S14]

Given a basis \mathbf{B} , vectors $\mathbf{v}_1, \dots, \mathbf{v}_N \in \mathcal{L}(\mathbf{B})$, and $\mathbf{B}^{-1}\mathbf{x} \notin \{\mathbf{B}^{-1}\mathbf{v}_i\}_{i \leq N}$, for any prime p , we have

$$\Pr_{\mathbf{z} \leftarrow U(\mathbb{Z}_q^n)} \left[\forall i, \begin{cases} \langle \mathbf{z}, \mathbf{B}^{-1}(\mathbf{x} + \mathbf{w}) \rangle = 0 \pmod{p} \\ \langle \mathbf{z}, \mathbf{B}^{-1}(\mathbf{v}_i + \mathbf{w}) \rangle \neq 0 \pmod{p} \end{cases} \right] \geq \frac{1}{p} - \frac{N}{p^2} - \frac{N}{p^{n-1}},$$

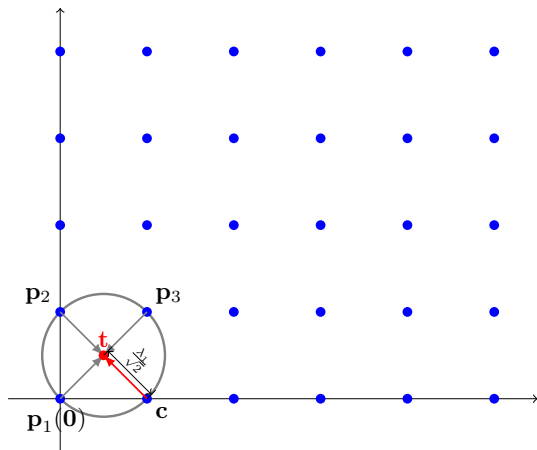
where $\mathbf{w} = \mathbf{B}\mathbf{u}$ for $\mathbf{u} \leftarrow U(\mathbb{Z}_q^n)$.

$$\frac{1}{p} - \frac{N}{p^2} \approx \frac{1}{p} \cdot \left(1 - \frac{1}{p}\right)^N.$$

- ▶ The latter formula is the (approximate) probability we get to
 - ▶ **keep** 1 point;
 - ▶ **remove** N points.

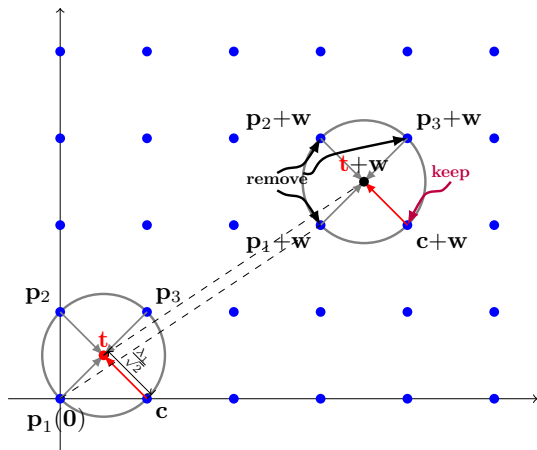
New reduction for $\gamma = 1$

- ▶ $\text{BDD}_{1/\sqrt{2}}$ instance: $(\mathcal{L}(\mathbf{B}), \mathbf{t})$.



New reduction for $\gamma = 1$

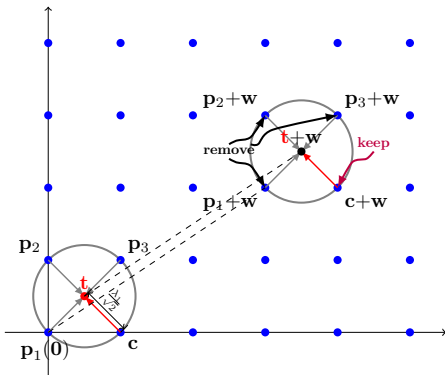
- ▶ Remove annoying points around the shifted target $\mathbf{t} + \mathbf{w}$.



New reduction for $\gamma = 1$

- ▶ Sparsify it!

$$\mathcal{L}_{p,z} = \{\mathbf{x} \in \mathcal{L}(\mathbf{B}) \mid \langle \mathbf{z}, \mathbf{B}^{-1} \mathbf{x} \rangle = 0 \pmod{p}\}$$



Choose p a prime, $\mathbf{z} \leftarrow \mathbb{Z}_p^n$; and hope the following conditions hold.

$$\begin{cases} \langle \mathbf{z}, \mathbf{B}^{-1}(\mathbf{c} + \mathbf{w}) \rangle = 0 \pmod{p} \\ \forall i, \langle \mathbf{z}, \mathbf{B}^{-1}(\mathbf{p}_i + \mathbf{w}) \rangle \neq 0 \pmod{p}. \end{cases}$$

Equivalently, we have

$$\begin{cases} \mathbf{c} + \mathbf{w} \in \mathcal{L}_{p,z} \\ \forall i, \mathbf{p}_i + \mathbf{w} \notin \mathcal{L}_{p,z}. \end{cases}$$

Algorithm for solving BDD

Version 2. The $\text{BDD}_{1/(\sqrt{2}\gamma)}$ to $\text{uSVP}_{\gamma'}$ reduction.

Input: a basis $\mathbf{B} = \{\mathbf{b}_i\}_{i \in [n]}$, and a target point \mathbf{t} .

Output: a lattice point \mathbf{c} such that $\|\mathbf{c} - \mathbf{t}\| = \text{dist}(\mathbf{t}, \mathcal{L})$.

0. Choose $p > N$ to be prime;

sample $\mathbf{z}, \mathbf{u} \leftarrow \mathbb{Z}_p^n$; compute $\mathbf{w} = \mathbf{B}\bar{\mathbf{u}} \in \mathcal{L}$.

Let $\mathbf{B}_{p,\mathbf{z}}$ denote the basis of $\mathcal{L}_{p,\mathbf{z}}$.

1. Define

$$\mathbf{B}' = \begin{pmatrix} \mathbf{B}_{p,\mathbf{z}} & \mathbf{t} + \mathbf{w} \\ \mathbf{0} & \frac{\lambda_1(\mathcal{L}(\mathbf{B}))}{2\gamma} \end{pmatrix}.$$

2. Run the $\text{uSVP}_{\gamma'}$ solver on input \mathbf{B}' . Let $\mathbf{s}' = \begin{pmatrix} \mathbf{s}'_1 \\ \mathbf{s}'_2 \end{pmatrix}$ be its output.

3. Output $\mathbf{t} - \mathbf{s}'_1$.

Algorithm for solving BDD

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Let $\mathbf{B}_{p,\mathbf{z}}$ denote the basis of $\mathcal{L}_{p,\mathbf{z}}$.

1. Define

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2. Run the $\text{uSVP}_{\gamma'}$ solver on input \mathbf{B}' . Let $\mathbf{s}' = \begin{pmatrix} \mathbf{s}'_1 \\ \mathbf{s}'_2 \end{pmatrix}$ be its output.

3. Output $\mathbf{t} - \mathbf{s}'_1$.

How sparse can the sublattice $\mathcal{L}_{p,\mathbf{z}}$ be?

How sparse?

Recall the probability to **keep** 1 point and **remove** N points:

$$\frac{1}{p} - \frac{N}{p^2}$$

How sparse?

Recall the probability to **keep** 1 point and **remove** N points:

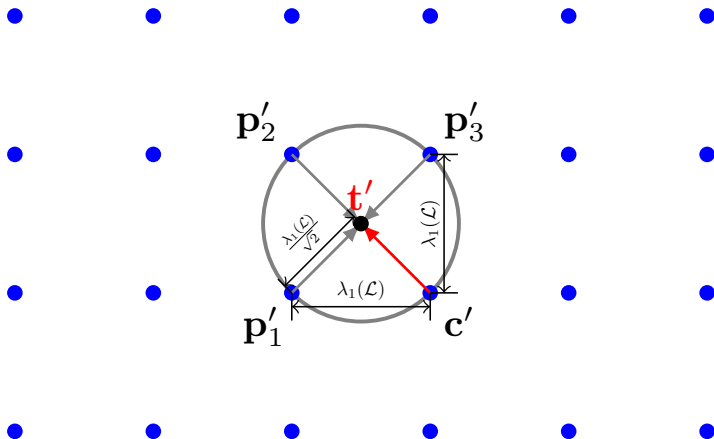
$$\frac{1}{p} - \frac{N}{p^2}$$

- ▶ We want it to be at least $\frac{1}{\text{poly}(n)}$;
- ▶ thus, $p \geq N$ and both should be $\leq \text{poly}(n)$.

We can sparsify the lattice by removing polynomially many points.

How many points around the target within $\lambda_1(\mathcal{L})/\sqrt{2}$

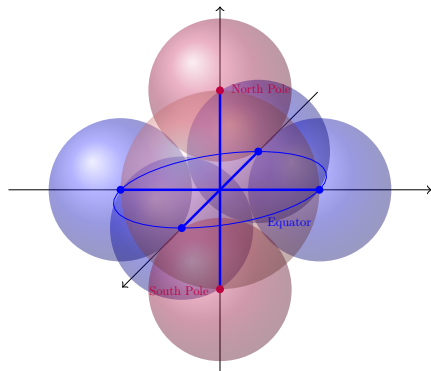
- ▶ How many points are there:
 - ▶ within $\lambda_1(\mathcal{L})/\sqrt{2}$ distance to target \mathbf{t}' ;
 - ▶ thus waiting for removal.



Sparsification works well within $\lambda_1/2$

Within $\lambda_1/\sqrt{2}$, adapted from [MG02, Th. 5.2]

For any n -dimensional lattice \mathcal{L} and any vector $\mathbf{t} \in \text{Span}(\mathcal{L})$, we have $\#\mathcal{L} \cap \mathcal{B}(\mathbf{t}, \lambda_1(\mathcal{L})/\sqrt{2}) \leq 2n$.



Extremely dense lattice, adapted from [MG02, Lem. 4.1]

For any $\alpha > 1/\sqrt{2}$, there exists $\epsilon > 0$ such that for any sufficiently large n we can find an n -dimensional lattice \mathcal{L} and a vector $\mathbf{t} \in \text{Span}(\mathcal{L})$, such that $\#\mathcal{L} \cap \mathcal{B}(\mathbf{t}, \alpha \cdot \lambda_1(\mathcal{L})) \geq 2^{n^\epsilon}$.

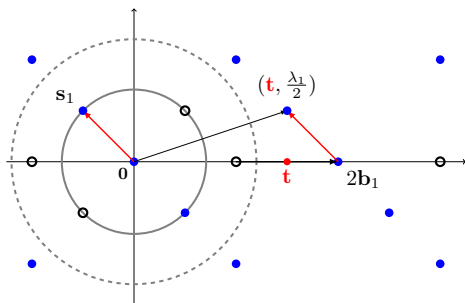
- ▶ This extremely dense lattice is the **Schnorr-Adleman** prime number lattice [MG02].
- ▶ Thus, worst-case list decoding radius is $\lambda_1/\sqrt{2}$.

Optimize the embedding

- ▶ Recall the last embedded lattice we got.

$$\text{BDD}_{\frac{1}{2\gamma}} \left(\begin{array}{c} \boxed{\mathbf{B}} \\ \boxed{\mathbf{t}} \end{array} \right) \xrightarrow[\text{embedding}]{\text{Kannan}} \text{USVP}_{\sqrt{2}\gamma'} \left[\begin{array}{c} \boxed{\tilde{\mathbf{B}}} \\ \boxed{\mathbf{t}} \\ \boxed{\mathbf{0}} \\ \boxed{\frac{\lambda_1}{2\gamma'}} \end{array} \right]$$

$(n = 1, \gamma = 1)$ $(n' = 2, \gamma' > 1)$

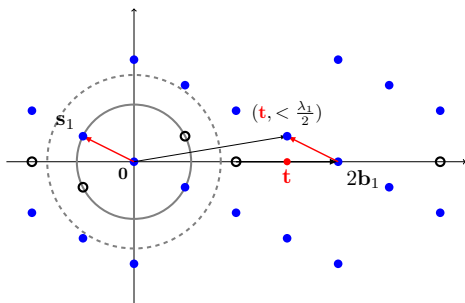


Optimize the embedding

- ▶ Decrease embedding height – focus on the original lattice.

$$\text{BDD}_{\frac{1}{2\gamma}} \left(\begin{array}{c} \boxed{\mathbf{B}} \\ (n, \gamma = 1) \end{array}, \begin{array}{c} \boxed{t} \end{array} \right) \xrightarrow[\text{embedding}]{\text{Kannan}} \text{USVP}_{\sqrt{2}\gamma'} \mathbf{B}' = \begin{array}{c} \boxed{\tilde{\mathbf{B}}} \\ \boxed{0} \end{array} \begin{array}{c} \boxed{t} \\ \boxed{\frac{\lambda_1}{2\gamma n}} \end{array}$$

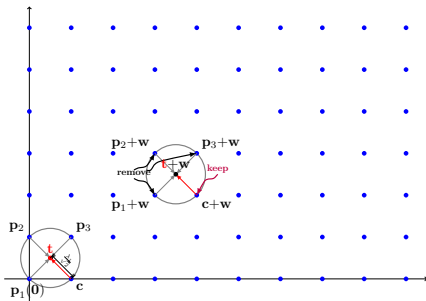
$(n' = n + 1, \gamma' > 1)$



Optimize the embedding

- ▶ Decrease embedding height – focus on the original lattice.

$$\begin{array}{ccc}
 \text{BDD}_{\frac{1}{2\gamma}} & & \text{USVP}_{\sqrt{2}\gamma'} \\
 \left(\begin{array}{c} \boxed{\mathbf{B}} \\ \boxed{\mathbf{t}} \end{array} \right) & \xrightarrow[\text{embedding}]{\text{Kannan}} & \mathbf{B}' = \begin{bmatrix} \boxed{\tilde{\mathbf{B}}} \\ \boxed{\mathbf{0}} \end{bmatrix} \begin{bmatrix} \boxed{\mathbf{t}'} \\ \boxed{\frac{\lambda_1}{2\gamma n}} \end{bmatrix} \\
 (n, \gamma = \frac{1}{\sqrt{2}}) & & (n' = n + 1, \gamma' > \frac{1}{\sqrt{2}})
 \end{array}$$

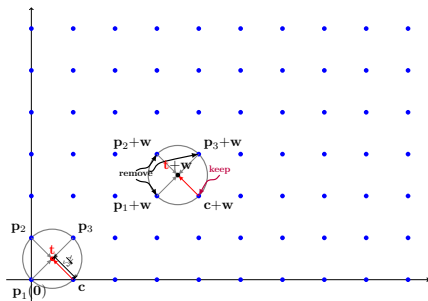


Optimize the embedding

- ▶ Decrease embedding height – focus on the original lattice.

$$\text{BDD } \frac{1}{\sqrt{2}\gamma} \left(\begin{array}{|c|} \hline \mathbf{B} \\ \hline \end{array} , \begin{array}{|c|} \hline \mathbf{t} \\ \hline \end{array} \right) \xrightarrow[\text{embedding}]{\text{Kannan}} \mathbf{B}' = \begin{array}{|c|} \hline \begin{array}{|c|} \hline \tilde{\mathbf{B}} \\ \hline \end{array} \\ \hline \mathbf{0} \end{array} \begin{array}{|c|} \hline \mathbf{t}' \\ \hline \frac{\lambda_1}{\sqrt{2-\gamma}} \end{array} \right)$$

$(n, \gamma = 1)$ $(n' = n + 1, \gamma' > 1)$



Optimize the embedding

- ▶ Decrease embedding height – focus on the original lattice.

$$\text{BDD } \frac{1}{\sqrt{2}\gamma}$$

$$\left(\begin{array}{|c|} \hline \mathbf{B} \\ \hline \end{array}, \begin{array}{|c|} \hline \mathbf{t} \\ \hline \end{array} \right)$$

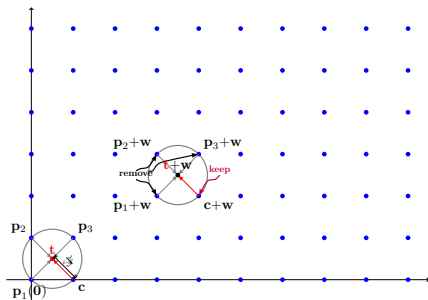
$$(n, \gamma = 1)$$

Kannan
embedding

$$\text{USVP } \gamma'$$

$$\mathbf{B}' = \begin{array}{|c|c|} \hline \begin{array}{|c|} \hline \tilde{\mathbf{B}} \\ \hline \end{array} & \begin{array}{|c|} \hline \mathbf{t}' \\ \hline \end{array} \\ \hline \begin{array}{|c|} \hline \mathbf{0} \\ \hline \end{array} & \begin{array}{|c|} \hline \frac{\lambda_1}{\sqrt{2-\gamma}} \\ \hline \end{array} \\ \hline \end{array}$$

$(n' = n + 1, \gamma' > 1)$



Optimize the embedding

- Accumulate embedding height – sparsify sufficiently many balls.

$$\text{BDD } \frac{1}{\sqrt{2}\gamma}$$

$$\left(\begin{array}{c} \boxed{\mathbf{B}} \\ \boxed{\mathbf{t}} \end{array} \right)$$

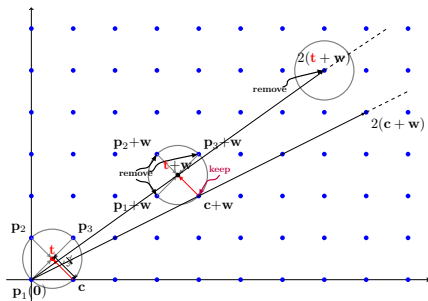
$$(n, \gamma = 1)$$

Kannan
embedding

$$\text{USVP } \gamma'$$

$$\mathbf{B}' = \begin{array}{c} \boxed{\tilde{\mathbf{B}}} \\ \boxed{0} \end{array} \begin{array}{c} \boxed{\mathbf{t}'} \\ \boxed{\frac{\lambda_1}{\sqrt{2-\gamma'}}} \end{array}$$

$$(n' = n + 1, \gamma' > 1)$$



Optimize the embedding

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$$\text{BDD } \frac{1}{\sqrt{2}\gamma}$$

$$\left(\begin{array}{c} \boxed{\mathbf{B}} \\ \boxed{\mathbf{t}} \end{array} \right)$$

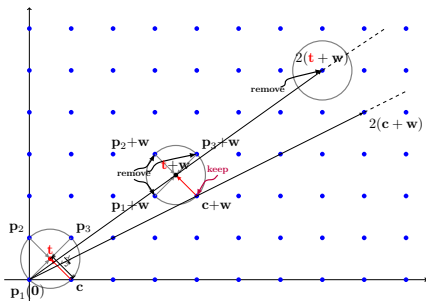
$$(n, \gamma = 1)$$

Kannan
embedding

$$\text{USVP } \gamma'$$

$$\mathbf{B}' = \begin{array}{|c|c|} \hline \boxed{\tilde{\mathbf{B}}} & \boxed{\mathbf{t}'} \\ \hline \boxed{0} & \boxed{\frac{\lambda_1}{\sqrt{2}\gamma n}} \\ \hline \end{array}$$

$\lambda'_2 > \lambda_1/\sqrt{2}$ ($[1, \gamma n]$)
 $\lambda_2 \geq \lambda_1/\sqrt{2}$ ($\geq \gamma n$)
 $(n' = n + 1, \gamma' \geq 1)$



Optimize the embedding

- Accumulate embedding height – sparsify sufficiently many balls.

$$\text{BDD } \frac{1}{\sqrt{2}\gamma}$$

$$\left(\begin{array}{c} \boxed{\mathbf{B}} \\ \boxed{\mathbf{t}} \end{array} \right)$$

$$(n, \gamma = 1)$$

Kannan
embedding

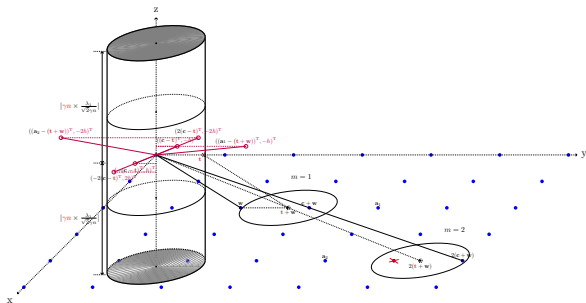
$$\text{USVP } \gamma'$$

$$\mathbf{B}' = \begin{array}{c} \boxed{\tilde{\mathbf{B}}} \quad \boxed{\mathbf{t}'} \\ \boxed{\mathbf{0}} \quad \boxed{\frac{\lambda_1}{\sqrt{2}\gamma n}} \end{array}$$

$$\lambda_2' \geq \lambda_1/\sqrt{2} \quad ([1, \gamma n])$$

$$\lambda_2 \geq \lambda_1/\sqrt{2} \quad (\geq \gamma n)$$

$$(n' = n + 1, \gamma' \geq 1)$$



Algorithm for solving BDD

Version 3. The $\text{BDD}_{1/(\sqrt{2}\gamma)}$ to USVP_γ reduction.

Input: a basis $\mathbf{B} = \{\mathbf{b}_i\}_{i \in [n]}$, and a target point \mathbf{t} .

Output: a lattice point \mathbf{c} such that $\|\mathbf{c} - \mathbf{t}\| = \text{dist}(\mathbf{t}, \mathcal{L})$.

0. Choose $p \geq \gamma n \cdot 2n$ to be prime;
sample $\mathbf{z}, \mathbf{u} \leftarrow \mathbb{Z}_p^n$; compute $\mathbf{w} = \mathbf{B}\bar{\mathbf{u}} \in \mathcal{L}$.
Let $\mathbf{B}_{\rho, \mathbf{z}}$ denote the basis of $\mathcal{L}_{\rho, \mathbf{z}}$.

1. Define

$$\mathbf{B}' = \begin{pmatrix} \mathbf{B}_{\rho, \mathbf{z}} & \mathbf{t} + \mathbf{w} \\ \mathbf{0} & \frac{\lambda_1(\mathcal{L}(\mathbf{B}))}{\sqrt{2}\gamma n} \end{pmatrix}.$$

2. Run the USVP_γ solver on input \mathbf{B}' . Let $\mathbf{s}' = \begin{pmatrix} \mathbf{s}'_1 \\ \mathbf{s}'_2 \end{pmatrix}$ be its output.

3. Output $\mathbf{t} - \mathbf{s}'_1$.

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3. Output $\mathbf{t} - \mathbf{s}'_1$.

Reduction is efficient for $\gamma \leq \text{poly}(n)$.

Main idea of the new reduction

- Target: **extend** the gap of the first two minima.
 - ▶ **Increase** the second minimum.
 - ▶ Khot's sparsification on lattice.
 - ▶ Sphere packing within $\lambda_1(\mathcal{L})/\sqrt{2}$ -radius ball gives the limit.
 - ▶ **Decrease** the first minimum.
 - ▶ Decrease the embedding height.

- Background
- The Lyubashevsky and Micciancio reduction and its limitation
- New reduction:
 - lattice sparsification.
 - example for $\gamma = 1$.
 - sphere packing.
- Open problems

- ▶ The sparsification in our reduction heavily relies on randomness.
 - Problem 1. Remove the sparsification randomness.
- ▶ Our reduction can only reduce $\text{BDD}_{1/(\sqrt{2}\gamma)}$ to USVP_γ for polynomially large γ .
 - Problem 2. Handle exponentially large γ .

Open problems

- ▶ Conjecture: BDD and uSVP are computationally identical.

$$\text{BDD}_{\frac{1}{c \cdot \gamma}} = \text{uSVP}_{\gamma}$$
$$c \in [1, \sqrt{2}]$$

• Problem 3. What is the constant c ? Is $c = \sqrt{2}$?

- ▶ In practice, randomly chosen lattice is sparse enough such that there is almost no point within $\lambda_1/\sqrt{2}$ -radius.

• Problem 4. Is sparsification just an artifact?