

Improved Reduction from BDD to uSVP in Lattices

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Séminaires Cryptologie et Sécurité, Université de Caen, Oct. 5, 2016

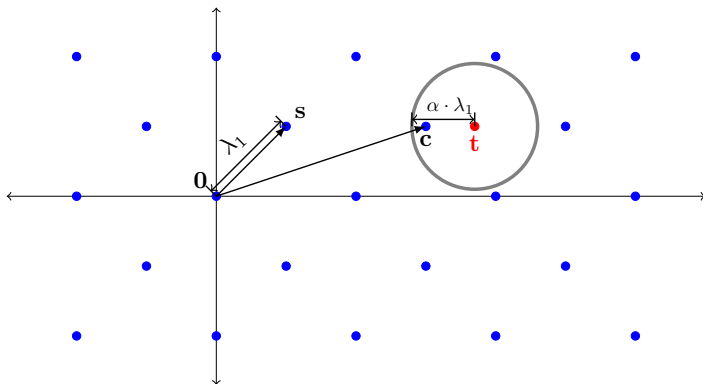


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Bounded Distance Decoding (BDD) and unique Shortest Vector Problem (uSVP)

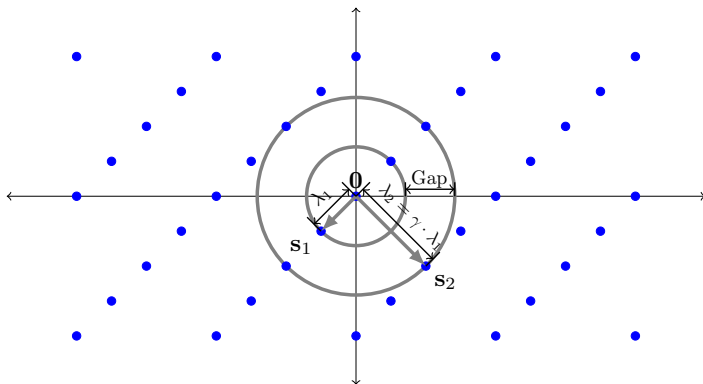


Bounded Distance Decoding for $\alpha \geq 0$ (BDD_α)

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} .

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



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

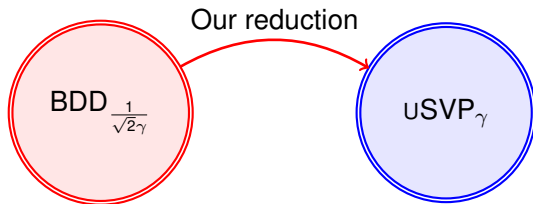
Input: $\mathbf{B} \in \mathbb{Q}^{n \times n}$ such that $\lambda_2(\mathcal{L}(\mathbf{B})) \geq \gamma \cdot \lambda_1(\mathcal{L}(\mathbf{B}))$.

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

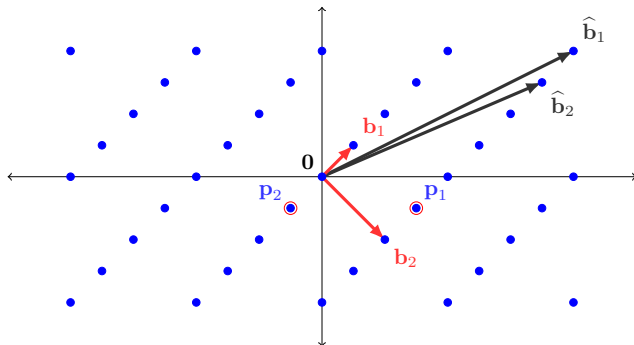
Improved reduction from BDD to uSVP (ICALP 2016)

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

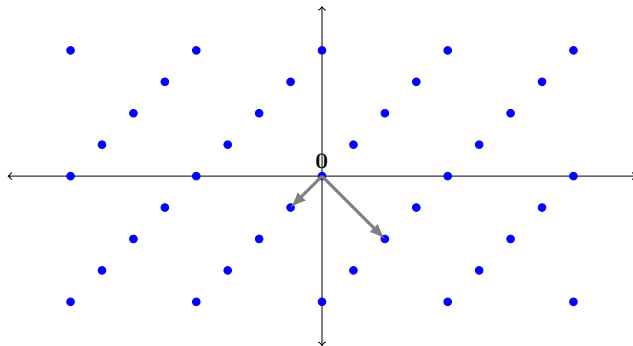


A definition of lattice

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

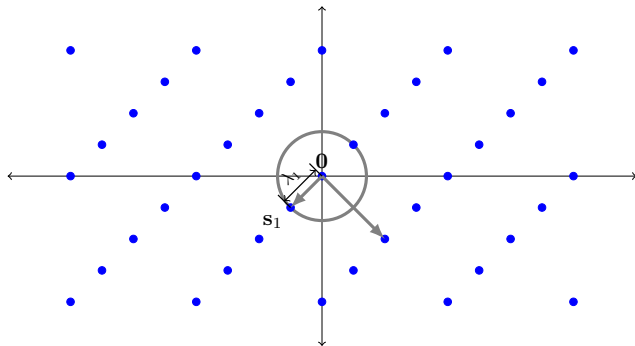


Lattice minimum

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

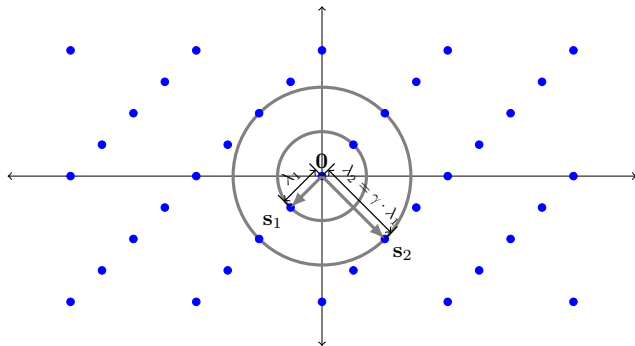


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Lattice Minima – second minimum

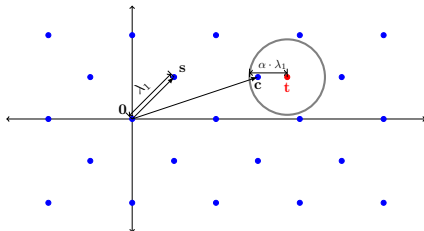


Lattice minimum

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Why is BDD interesting?



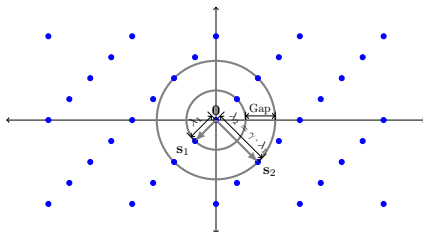
- ▶ In **cryptography**:

- ▶ 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:

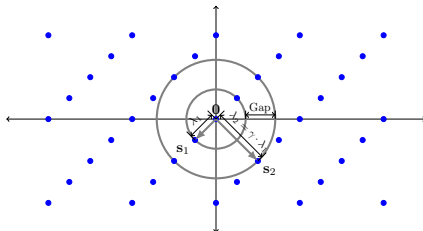
- ▶ Wifi, mobile phone *etc*;
- ▶ View message as a lattice point, Gaussian noise is added in channel transmission, decoding is solving BDD.

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.

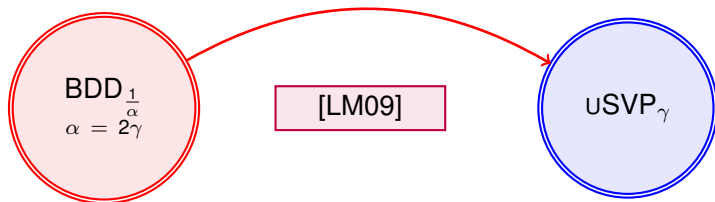
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 - ▶ First, reduce BDD to uSVP.
 - ▶ Second, solve **uSVP** by lattice reduction, e.g., LLL and BKZ.

$\text{BDD}_{\frac{1}{\text{poly}(n)}}$ and $\text{uSVP}_{\text{poly}(n)}$ 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.

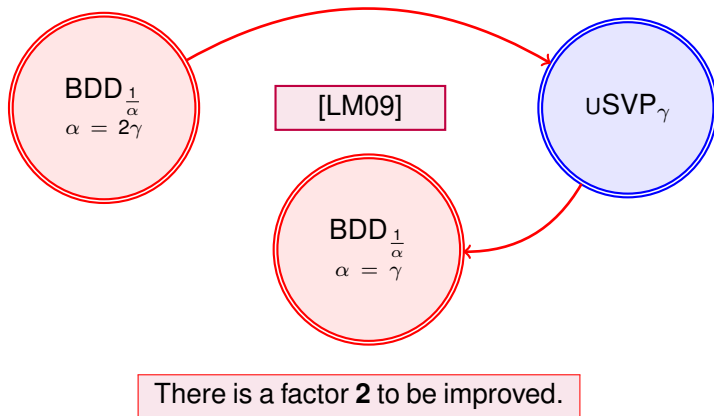
[LM09]: V. Lyubashevsky and D. Micciancio. On bounded distance decoding, unique shortest vectors, and the minimum distance problem, CRYPTO, 2009.

[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.

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Prior works on BDD to uSVP



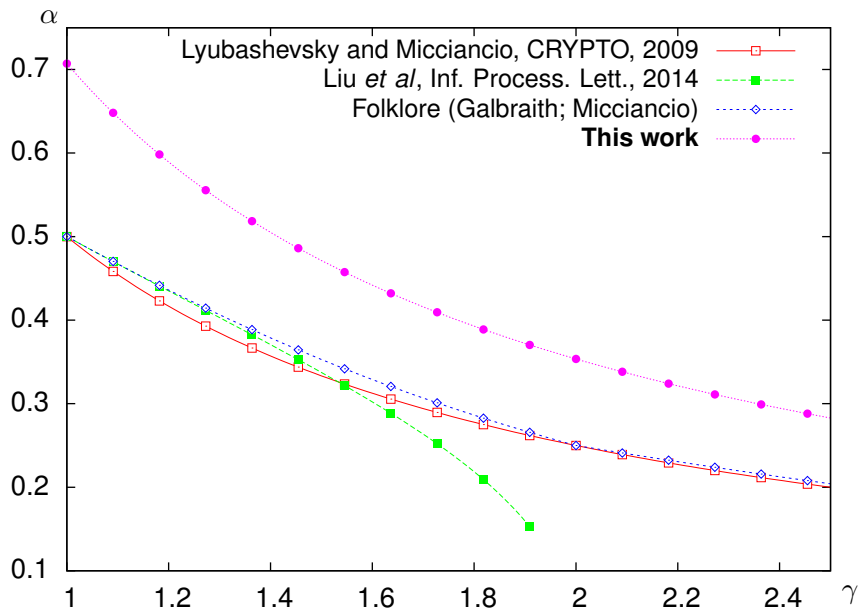
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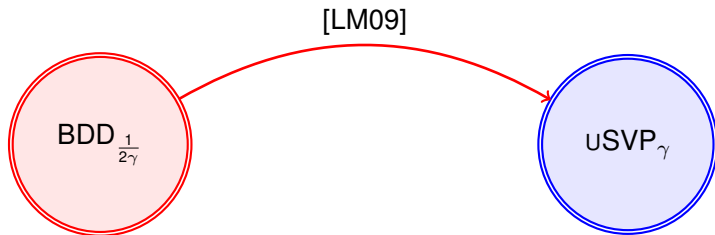
Comparison with prior works



The Lyubashevsky and Micciancio reduction

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

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



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

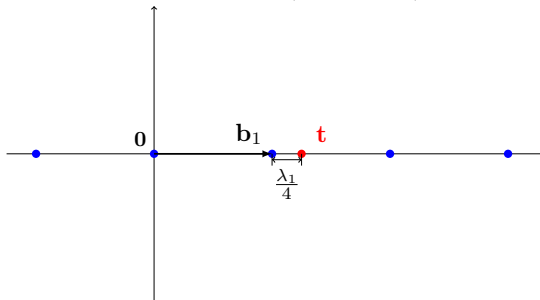
A simple case

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

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

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

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



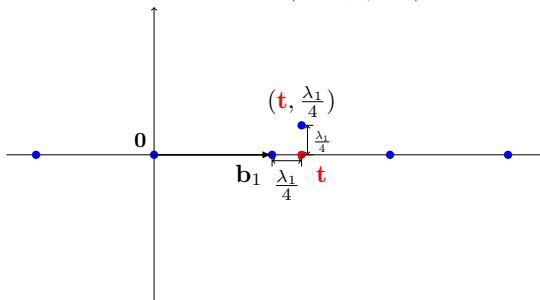
A simple case

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

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

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

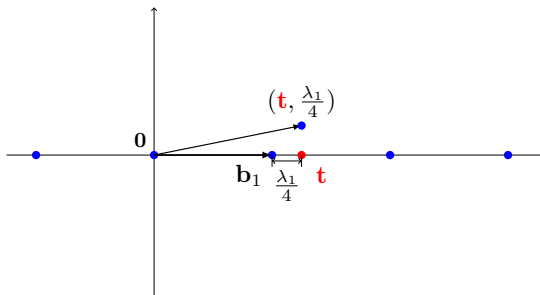
$$(n = 1, \gamma = 2)$$



A simple case

- Kannan embedding.

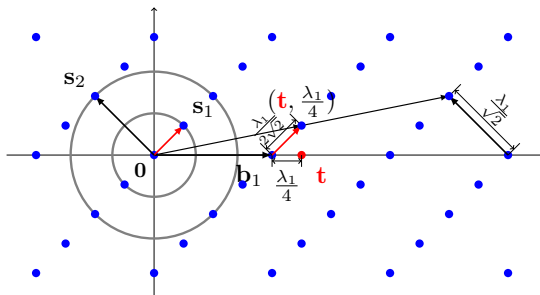
$$\begin{array}{ccc}
 \text{BDD}_{\frac{1}{2\gamma}} & & \text{uSVP}_{\gamma'} \\
 \left(\begin{array}{|c|} \hline \mathbf{B} \\ \hline \end{array}, \begin{array}{|c|} \hline \mathbf{t} \\ \hline \end{array} \right) & \xRightarrow[\text{embedding}]{\text{Kannan}} & \mathbf{B}' = \left[\begin{array}{|c|}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



A simple case

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

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



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

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

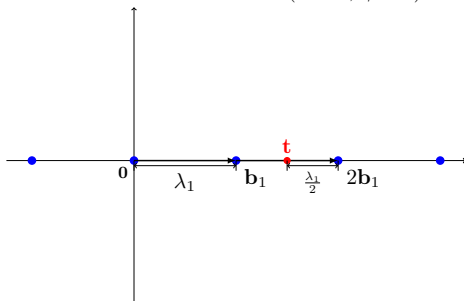
Limiting case of the reduction

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

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

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

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



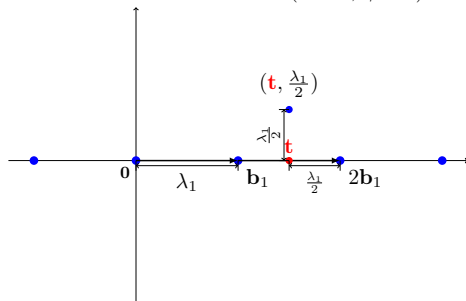
Limiting case of the reduction

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

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

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

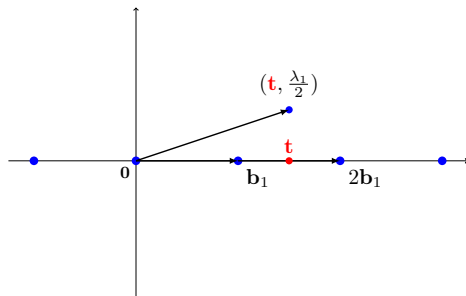
$$(n = 1, \gamma = 1)$$



Limiting case of the reduction

- Kannan embedding.

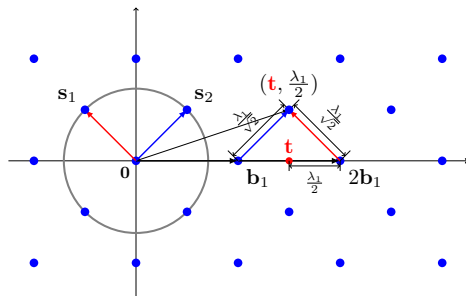
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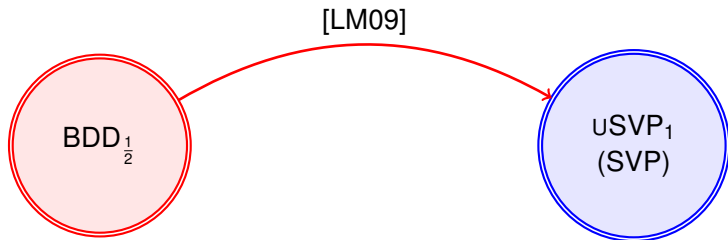
Limiting case of the reduction

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

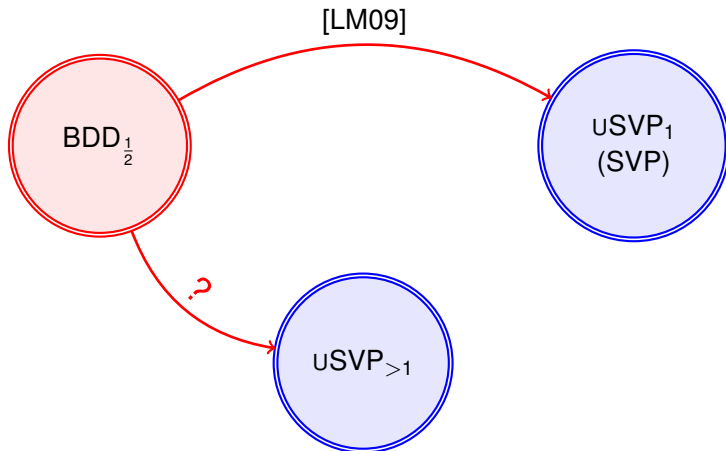
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 & & (n'=2, \gamma'=1)
 \end{array}$$



This is the best this reduction can achieve

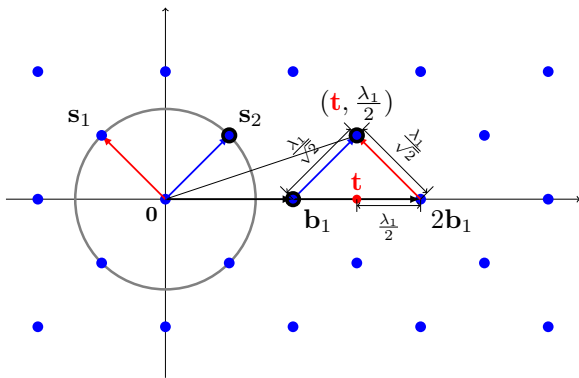


Can we improve it?



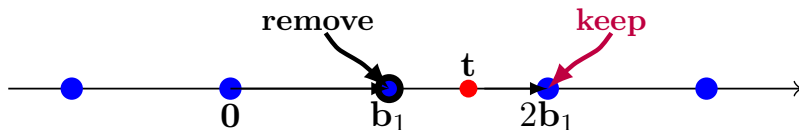
An attempt to circumvent the limitation

- Limitation in the Lyubushevsky and Micciancio reduction.



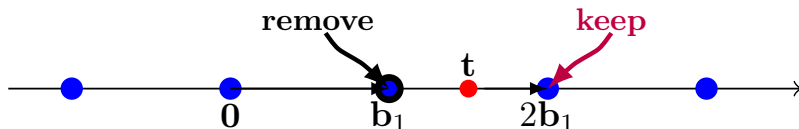
An attempt to circumvent the limitation

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

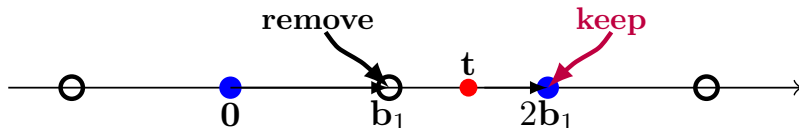


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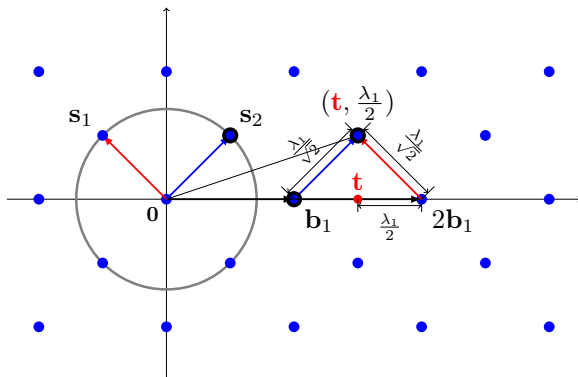


- ▶ Lattice $\mathcal{L}(\tilde{\mathbf{B}})$ with $\tilde{\mathbf{B}} = [2\mathbf{b}_1]$.



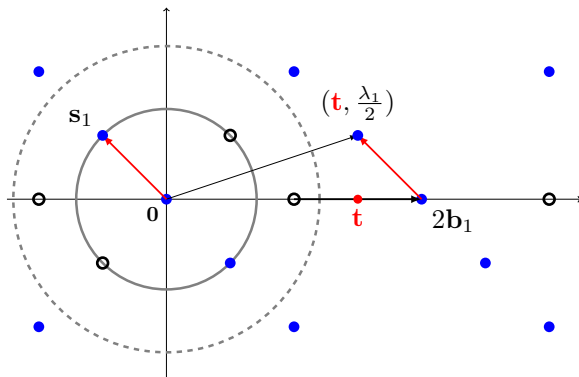
An attempt to circumvent the limitation

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



An attempt to circumvent the limitation

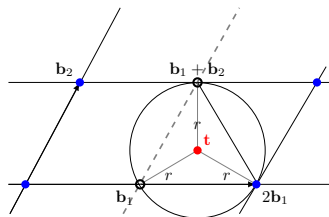
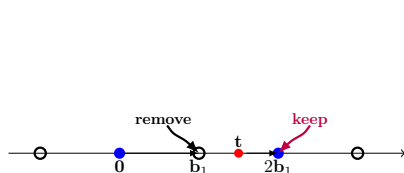
- Limitation is circumvented (for this example): $\lambda'_2 > \lambda'_1$ now!



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

Main idea

- Deterministic sparsification leads to a combinatorial explosion.



- But we want more...

- keep only 1 closest vector to target t .
- remove all other somewhat close N vectors to t .

Lattice Sparsification

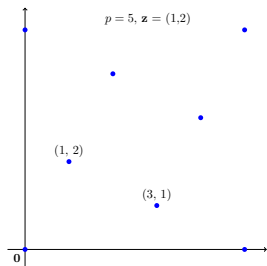
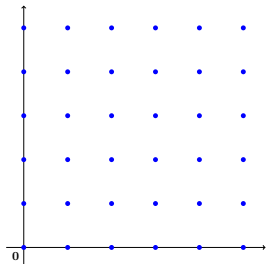
Khot's Lattice Sparsification [K03]

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

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

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

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



[K03]: S. Khot. Hardness of approximating the shortest vector problem in high L_p norms. FOCS'03, 2003.

Sparsification on lattice

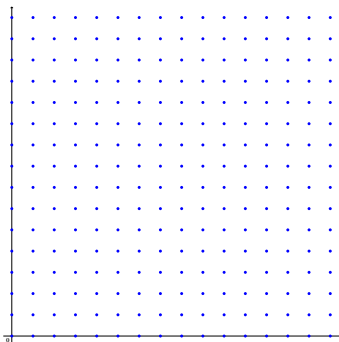
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- 1st sparsification:
 $p = 5, \mathbf{z} = (0, 0)$.

Sparsification on lattice

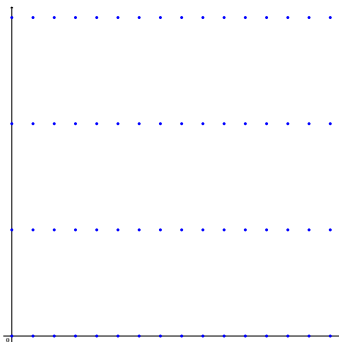
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► 2nd sparsification:
 $p = 5, \mathbf{z} = (0, 1)$.

Sparsification on lattice

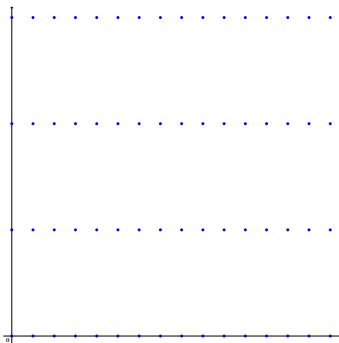
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where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



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

Sparsification on lattice

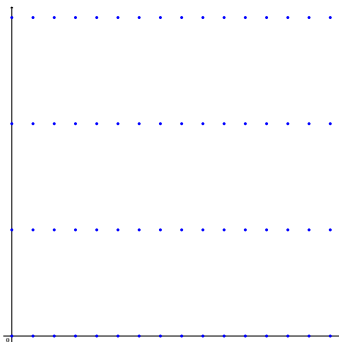
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► 4th sparsification:
 $p = 5, \mathbf{z} = (0, 3)$.

Sparsification on lattice

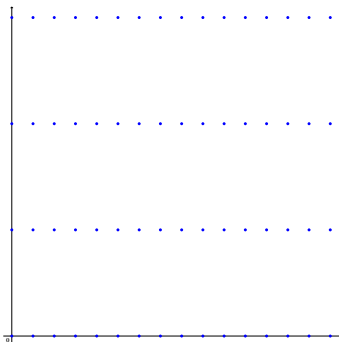
Khot's Lattice Sparsification

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



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

Sparsification on lattice

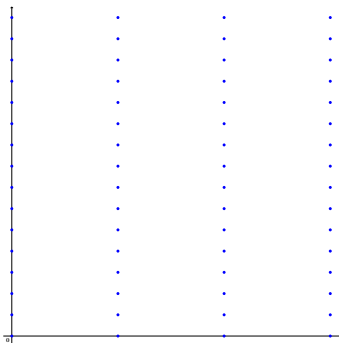
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Output: a sparsified sub-lattice of $\mathcal{L}(\mathbf{B})$:

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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

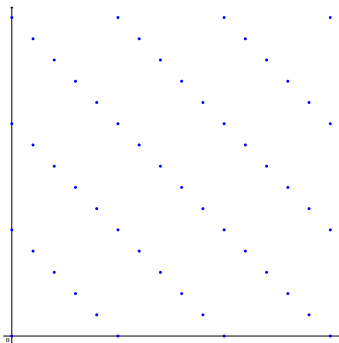
Khot's Lattice Sparsification

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

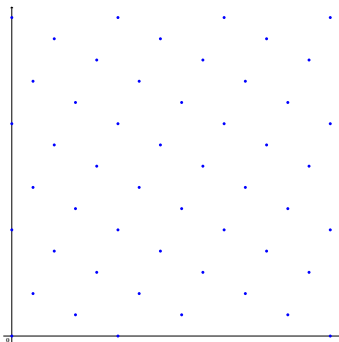
Khot's Lattice Sparsification

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

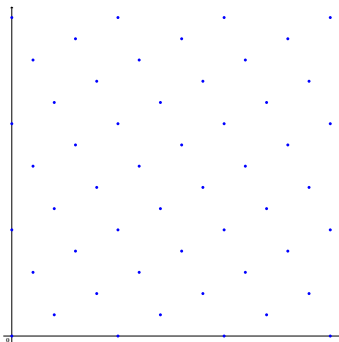
Khot's Lattice Sparsification

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

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

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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

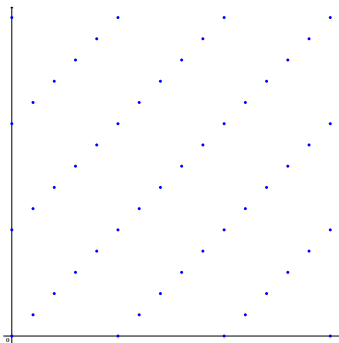
Khot's Lattice Sparsification

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

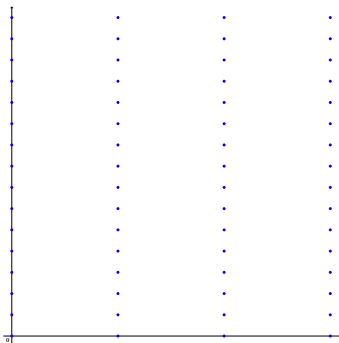
Khot's Lattice Sparsification

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

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

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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

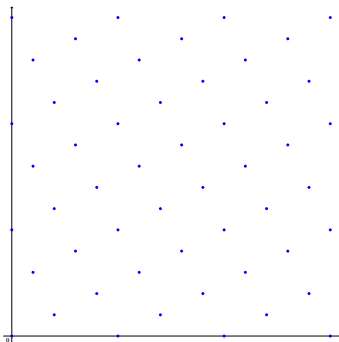
Khot's Lattice Sparsification

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

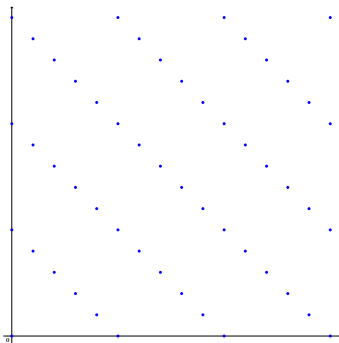
Khot's Lattice Sparsification

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

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

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

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

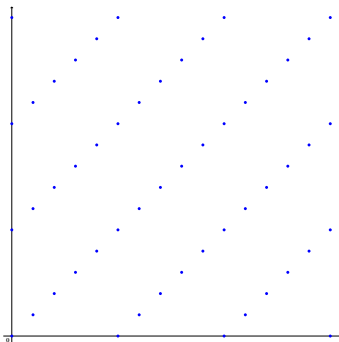
Khot's Lattice Sparsification

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

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

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

Sparsification on lattice

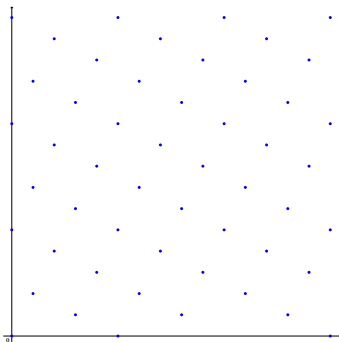
Khot's Lattice Sparsification

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

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

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

Sparsification on lattice

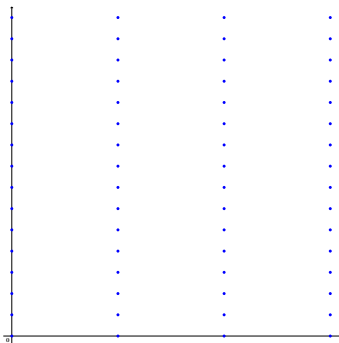
Khot's Lattice Sparsification

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

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

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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

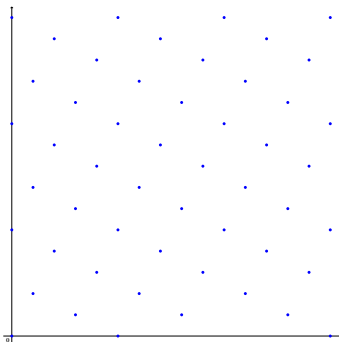
Khot's Lattice Sparsification

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

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

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

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

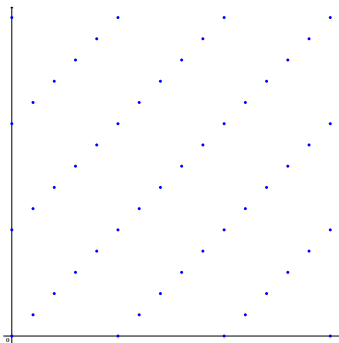
Khot's Lattice Sparsification

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

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

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

Sparsification on lattice

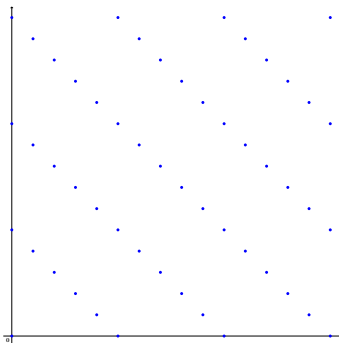
Khot's Lattice Sparsification

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

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

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

Sparsification on lattice

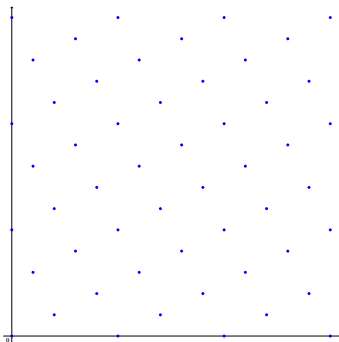
Khot's Lattice Sparsification

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

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

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

Sparsification on lattice

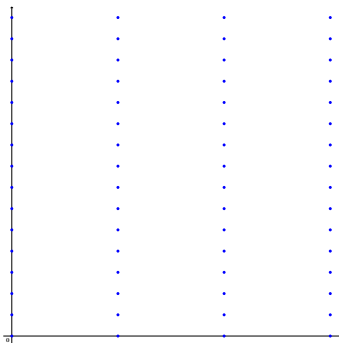
Khot's Lattice Sparsification

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

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

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



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

Sparsification on lattice

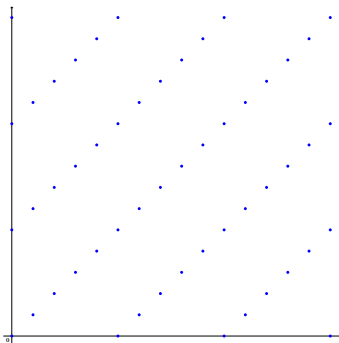
Khot's Lattice Sparsification

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

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

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

Sparsification on lattice

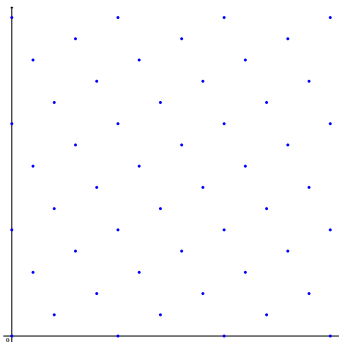
Khot's Lattice Sparsification

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

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

Sparsification on lattice

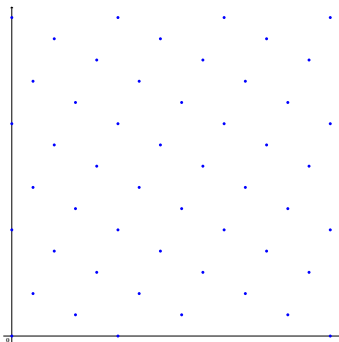
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where p is a prime integer and $\mathbf{z} \in \mathbb{Z}_p^n$.



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

Sparsification on lattice

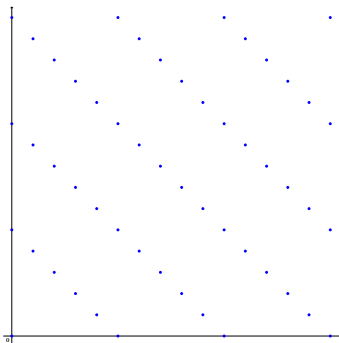
Khot's Lattice Sparsification

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

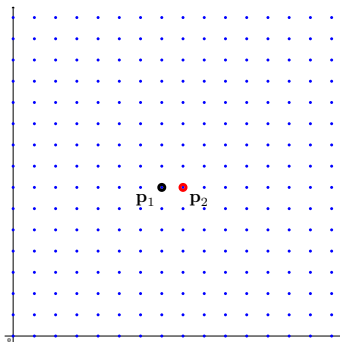
Khot's Lattice Sparsification

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

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

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

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

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

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

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

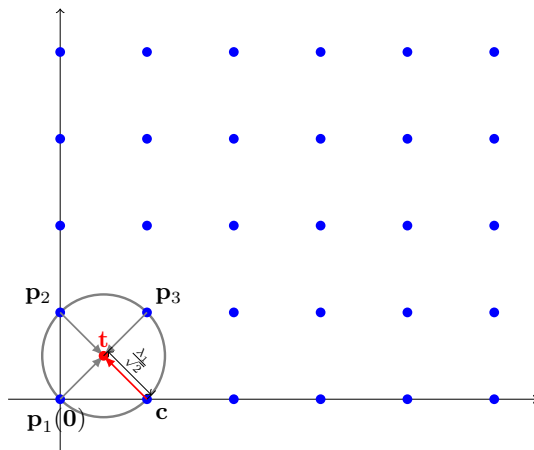
$$\frac{1}{p} - \frac{N}{p^2} \approx \frac{1}{p} \cdot \left(1 - \frac{1}{p}\right)^N.$$

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

[S14]: N. Stephens-Davidowitz. Discrete Gaussian sampling reduces to CVP and SVP. In *Proc. of SODA*, 2016.

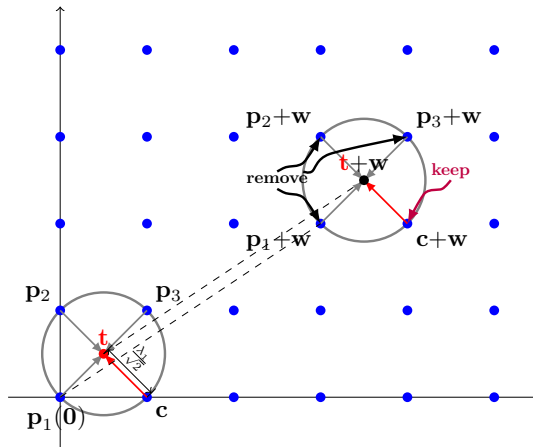
New reduction for $\gamma = 1$

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



New reduction for $\gamma = 1$

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



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 $\mathbf{w} = \mathbf{B}\mathbf{u} \in \mathcal{L}$.
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 $\mathbf{s}' = \begin{pmatrix} \mathbf{s}'_1 \\ s'_2 \end{pmatrix}$ be its output.
3. Output $\mathbf{t} - \mathbf{s}'_1$.

Algorithm for solving BDD

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

How sparse?

Recall the probability to **keep** $\boxed{1}$ point and **remove** \boxed{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 $\leq \text{poly}(n)$.

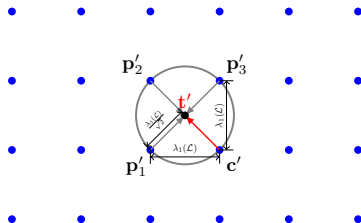
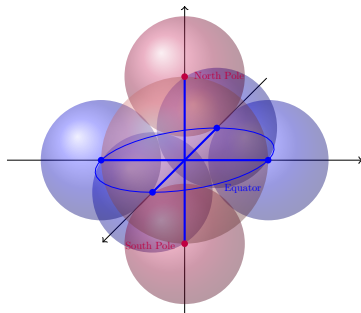
We can sparsify the lattice by removing polynomially many points.

So what is the worst list decoding radius?

Approaching the limit

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 $\#\mathcal{L} \cap \mathcal{B}(\mathbf{t}, \lambda_1(\mathcal{L})/\sqrt{2}) \leq 2n$.

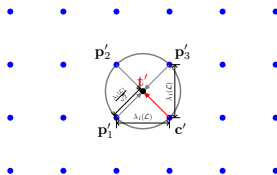
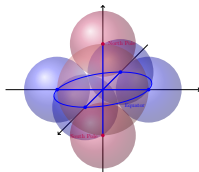


[MG02]: D. Micciancio and S. Goldwasser. Complexity of lattice problem: A cryptography perspective. Kluwer, 2009.

Approaching the limit

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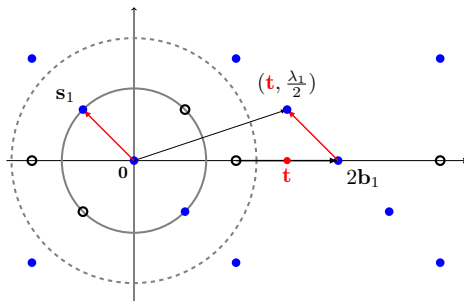
Extremely dense lattice, adapted from [MG02, Lem. 4.1]

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 $\#\mathcal{L} \cap \mathcal{B}(\mathbf{t}, \alpha \cdot \lambda_1(\mathcal{L})) \geq 2^{n^\epsilon}$.

Optimize the embedding

- Recall the last embedded lattice we got.

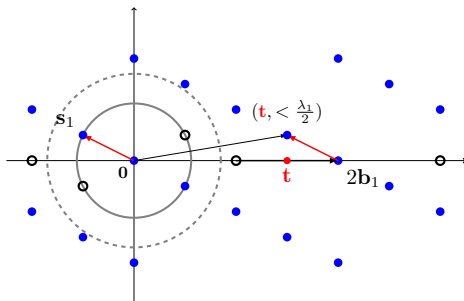
$$\begin{array}{ccc}
 \text{BDD}_{\frac{1}{2\gamma}} & & \text{USVP}_{\gamma'} \\
 \left(\begin{array}{c} \boxed{\mathbf{B}} \\ (n=1, \gamma=1) \end{array}, \begin{array}{c} \boxed{\mathbf{t}} \end{array} \right) & \xrightarrow[\text{embedding}]{\text{Kannan}} & \mathbf{B}' = \begin{bmatrix} \boxed{\tilde{\mathbf{B}}}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



Optimize the embedding

- Decrease embedding height

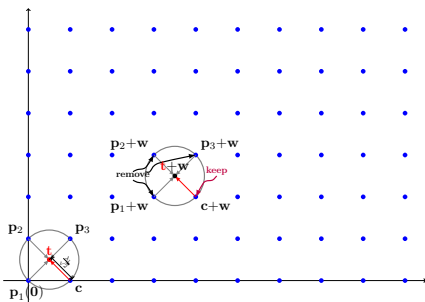
$$\begin{array}{ccc}
 \text{BDD}_{\frac{1}{2\gamma}} & & \text{USVP}_{\gamma'} \\
 \left(\begin{array}{c} \boxed{\mathbf{B}} \\ (n, \gamma = 1) \end{array}, \begin{array}{c} \boxed{\mathbf{t}} \end{array} \right) & \xrightarrow[\text{embedding}]{\text{Kannan}} & \mathbf{B}' = \begin{bmatrix} \boxed{\tilde{\mathbf{B}}}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



Optimize the embedding

- Decrease embedding height \rightarrow focus on the original lattice.

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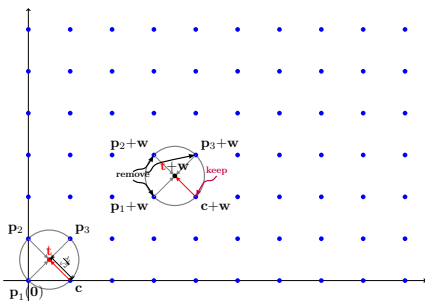


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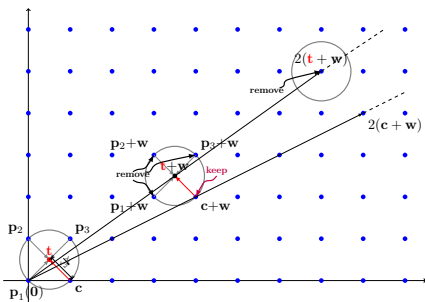
$\lambda'_2 > \lambda_1 / \sqrt{2}$



Optimize the embedding

- ▶ Accumulate embedding height – sparsify sufficiently many balls.

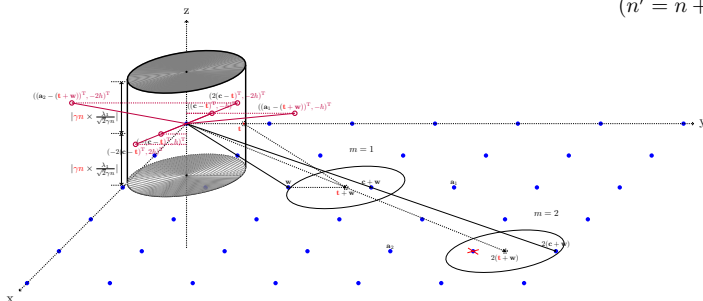
$$\text{BDD} \xrightarrow{\frac{1}{\sqrt{2}\gamma}} \text{USVP}_{\gamma'} \quad \begin{pmatrix} \boxed{\mathbf{B}} & \boxed{\mathbf{t}} \end{pmatrix} \xrightarrow[\text{embedding}]{\text{Kannan}} \mathbf{B}' = \begin{bmatrix} \boxed{\tilde{\mathbf{B}}} & \boxed{\mathbf{t}'} \\ \boxed{0} & \boxed{\frac{\lambda_1}{\sqrt{2}\gamma n}} \end{bmatrix}$$



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 (n, \gamma = 1) & & \begin{array}{l} \lambda'_2 > \lambda_1/\sqrt{2} \quad ([1, \gamma n]) \\ \lambda'_2 > \lambda_1/\sqrt{2} \quad (> \gamma n) \\ (n' = n + 1, \gamma' > 1) \end{array}
 \end{array}$$



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

1. Define

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

2. Run the uSVP_γ solver on input \mathbf{B}' . Let $\mathbf{s}' = \begin{pmatrix} \mathbf{s}'_1 \\ s'_2 \end{pmatrix}$ be its output.
3. Output $\mathbf{t} - \mathbf{s}'_1$.

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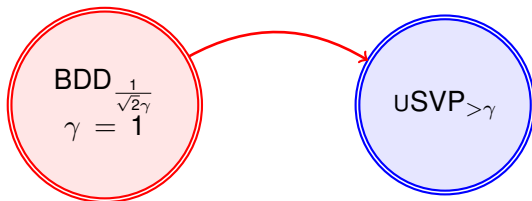
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Reduction is efficient for $\gamma \leq \text{poly}(n)$.

Our result

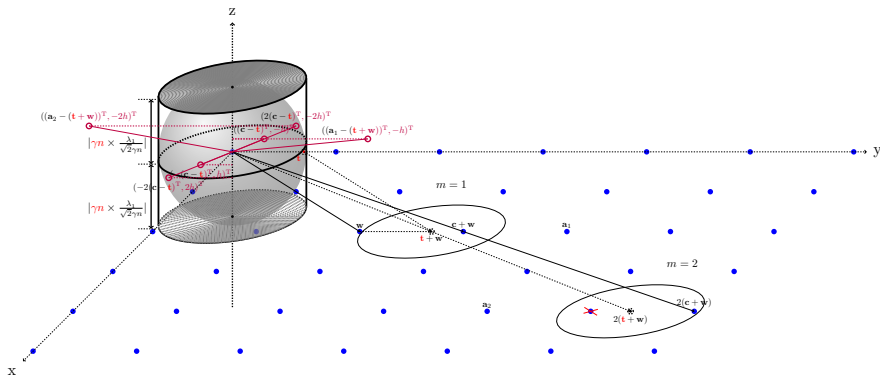


The reduction algorithm also works for any γ with $\gamma \leq \text{poly}(n)$.

This algorithm actually works for any $\gamma \geq 1$,
thanks Stephens-Davidowitz for the observation.

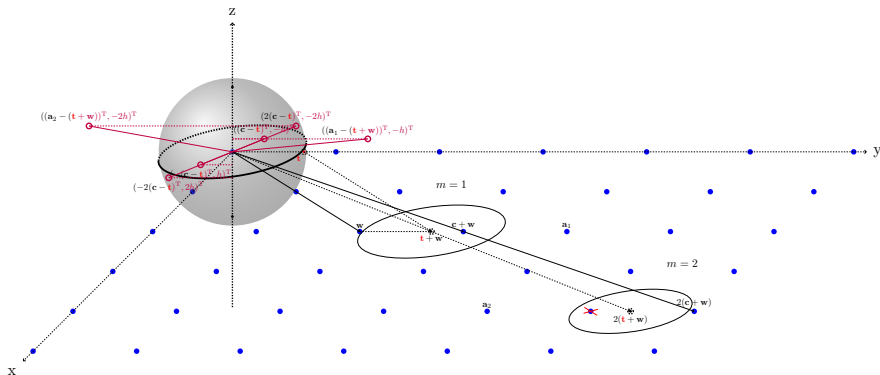
Intuition of extension for any $\gamma > 1$

- A tighter view on the reduction.



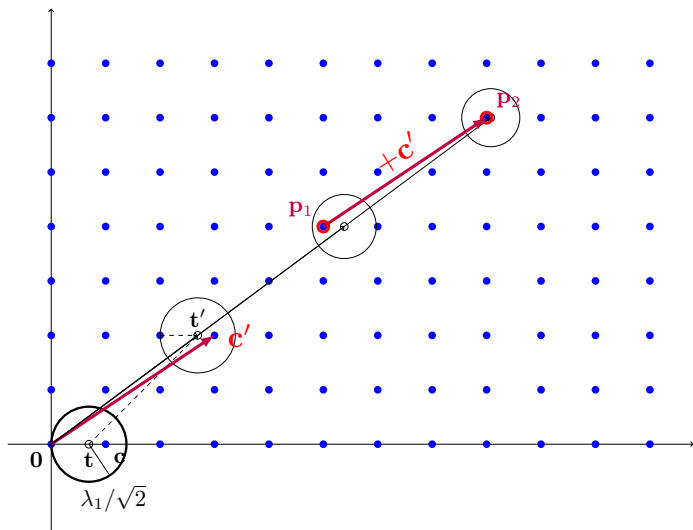
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Intuition of extension for any $\gamma > 1$

- Intuition of the improvement.



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

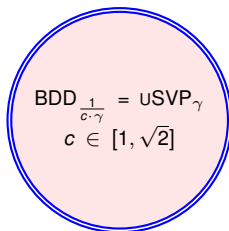
- ▶ The sparsification in our reduction heavily relies on randomness.

- Problem 1. Remove the sparsification randomness.

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

- Problem 2. Is sparsification just an artifact?

- 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 $c = \sqrt{2}$?