

Dimension Preserving Reductions Between SVP and CVP in Different p -Norms



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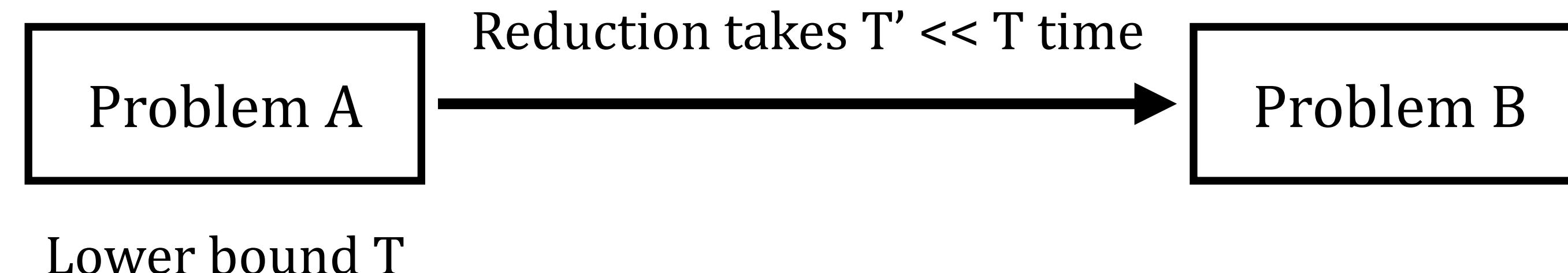
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- Many powerful crypto primitives are only known via lattices (such as **Fully homomorphic encryption**).
- Most popular candidate for post-quantum cryptography (classical crypto system secure even after quantum computer)

Exponential Time Reductions

- Polynomial Time Reduction are used to prove the NP-Hardness of the problem.
- Exponential Time Reductions:
 - If we believe problem is exponentially hard.
 - Implies the fine-grained hardness.



Lattices

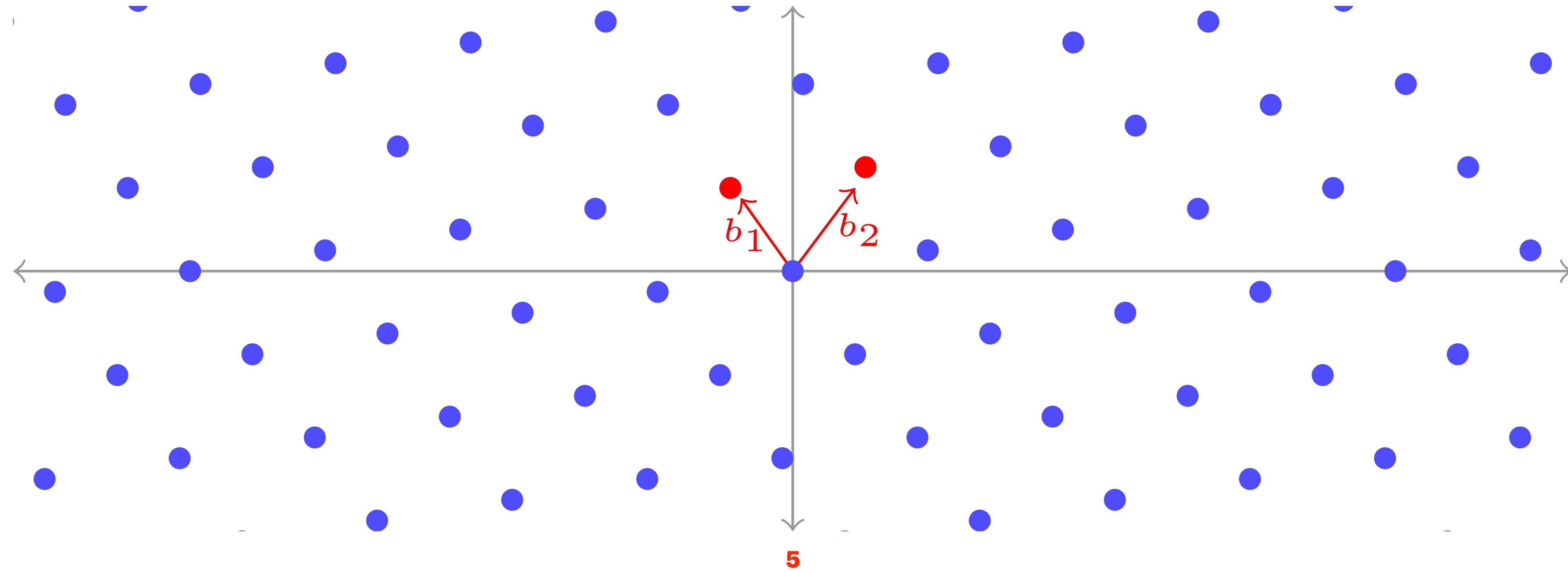
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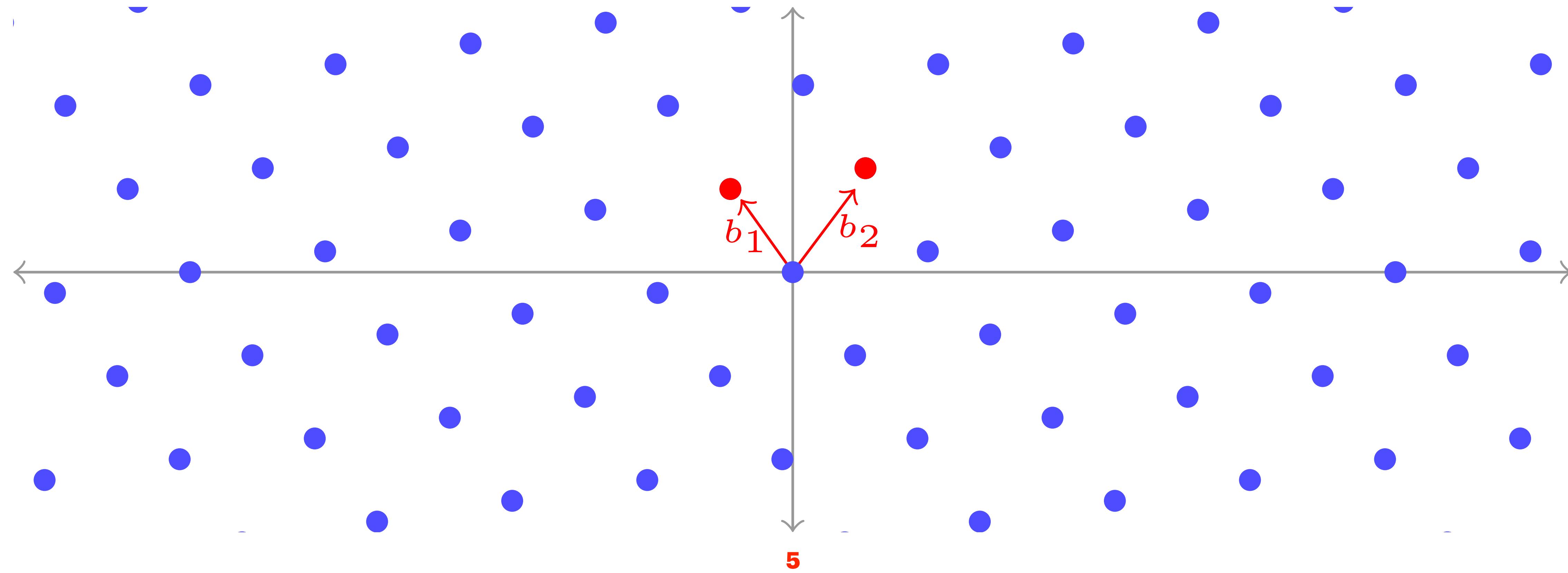
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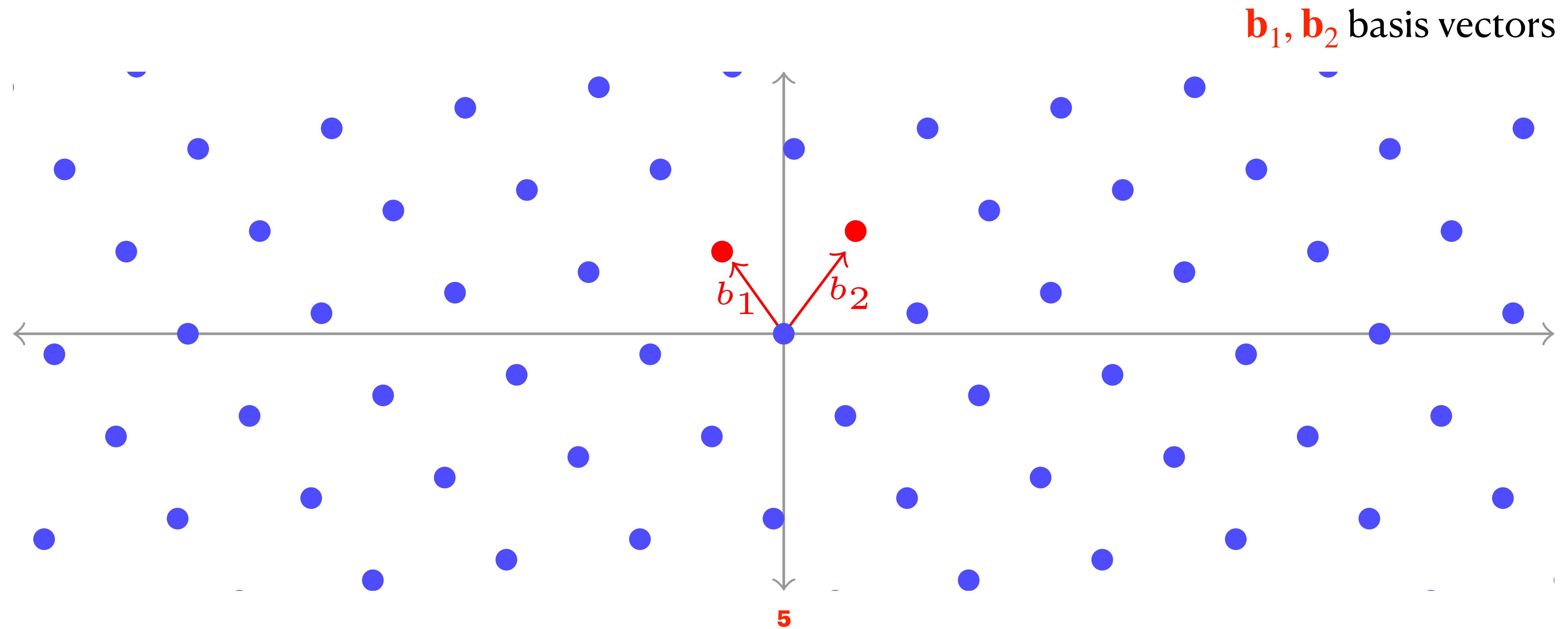
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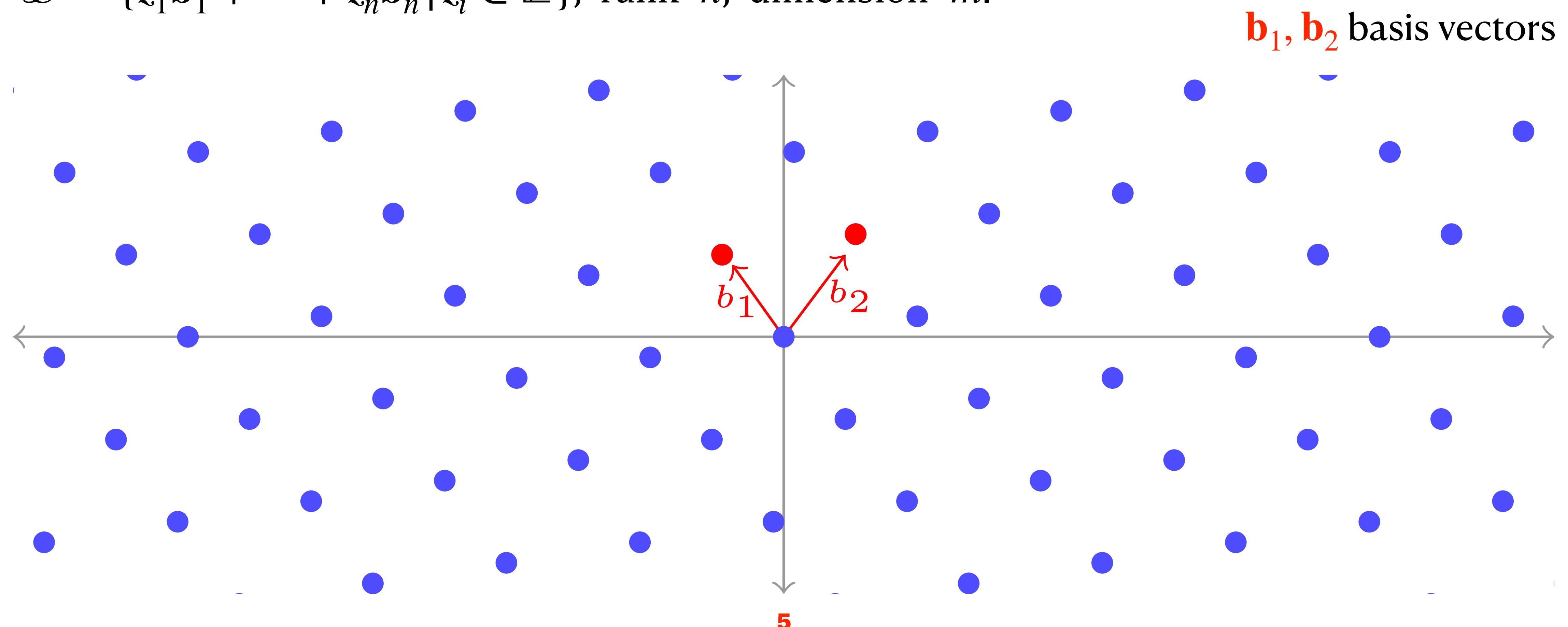
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- Specified by a basis $\mathbf{b}_1, \dots, \mathbf{b}_n$, linearly independent vectors.
- $\mathcal{L} = \{z_1\mathbf{b}_1 + \dots + z_n\mathbf{b}_n \mid z_i \in \mathbb{Z}\}$, rank= n , dimension= m .



Notations

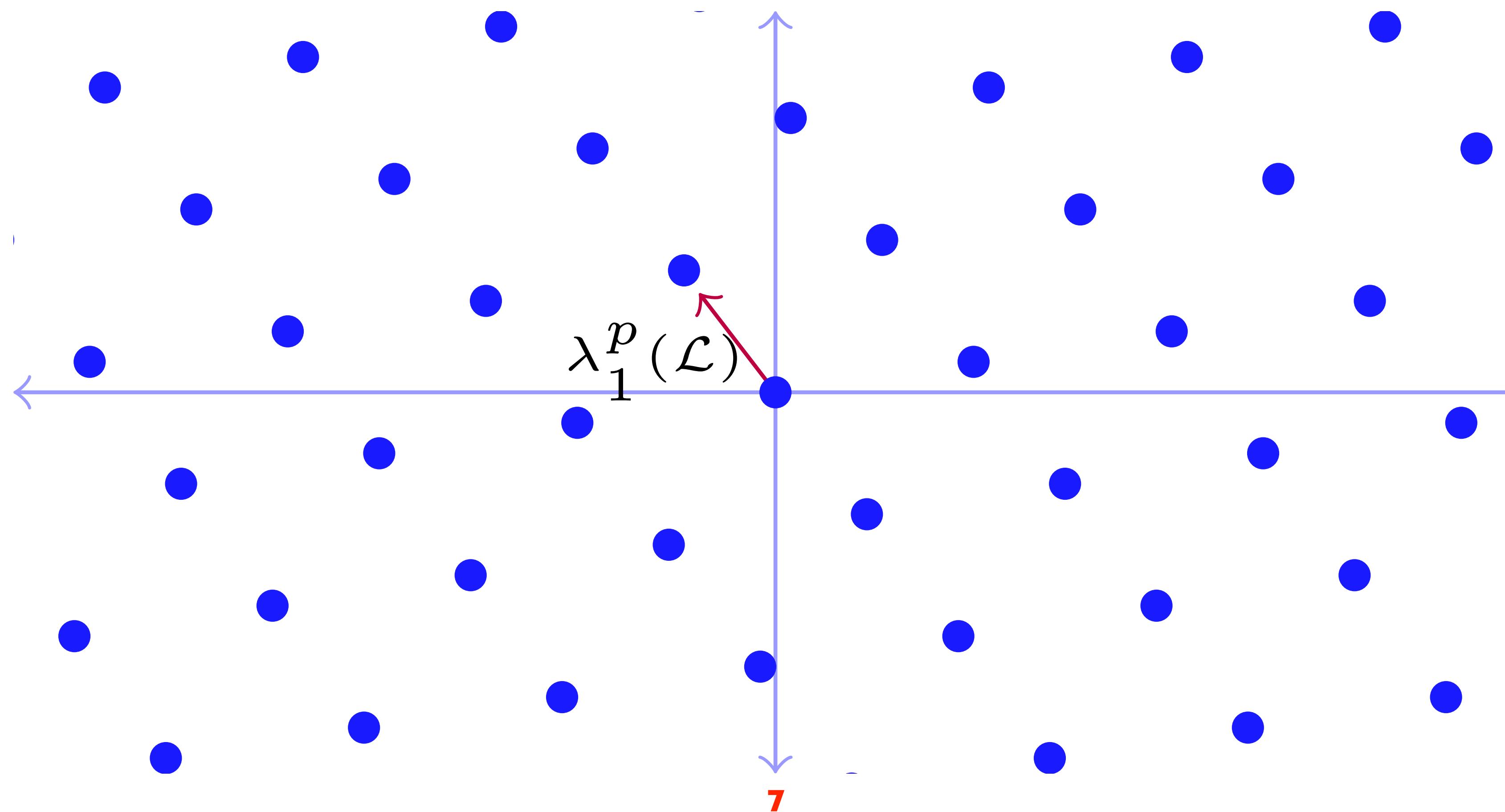
- ℓ_p norm:

$$p \in [1, \infty), \|\mathbf{v}\|_p := \left(\sum_{i=1}^m |\nu_i|^p \right)^{1/p}$$

$$p = \infty, \|\mathbf{v}\|_\infty := \max_i |\nu_i|.$$

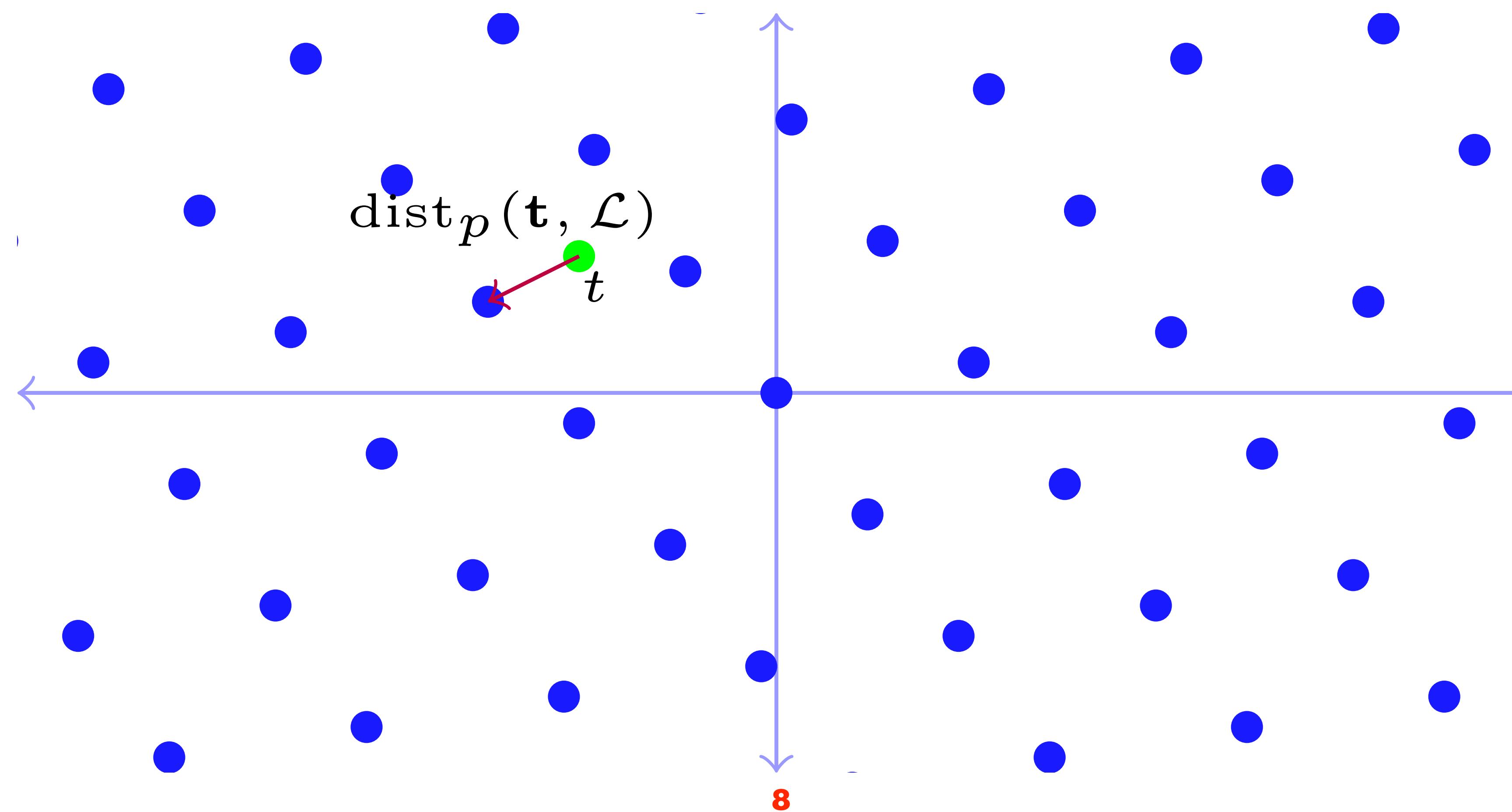
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- For any lattice \mathcal{L} , $\lambda_1^p(\mathcal{L}) = \min_{\mathbf{u} \neq \mathbf{0}, \mathbf{u} \in \mathcal{L}} \|\mathbf{u}\|_p$.



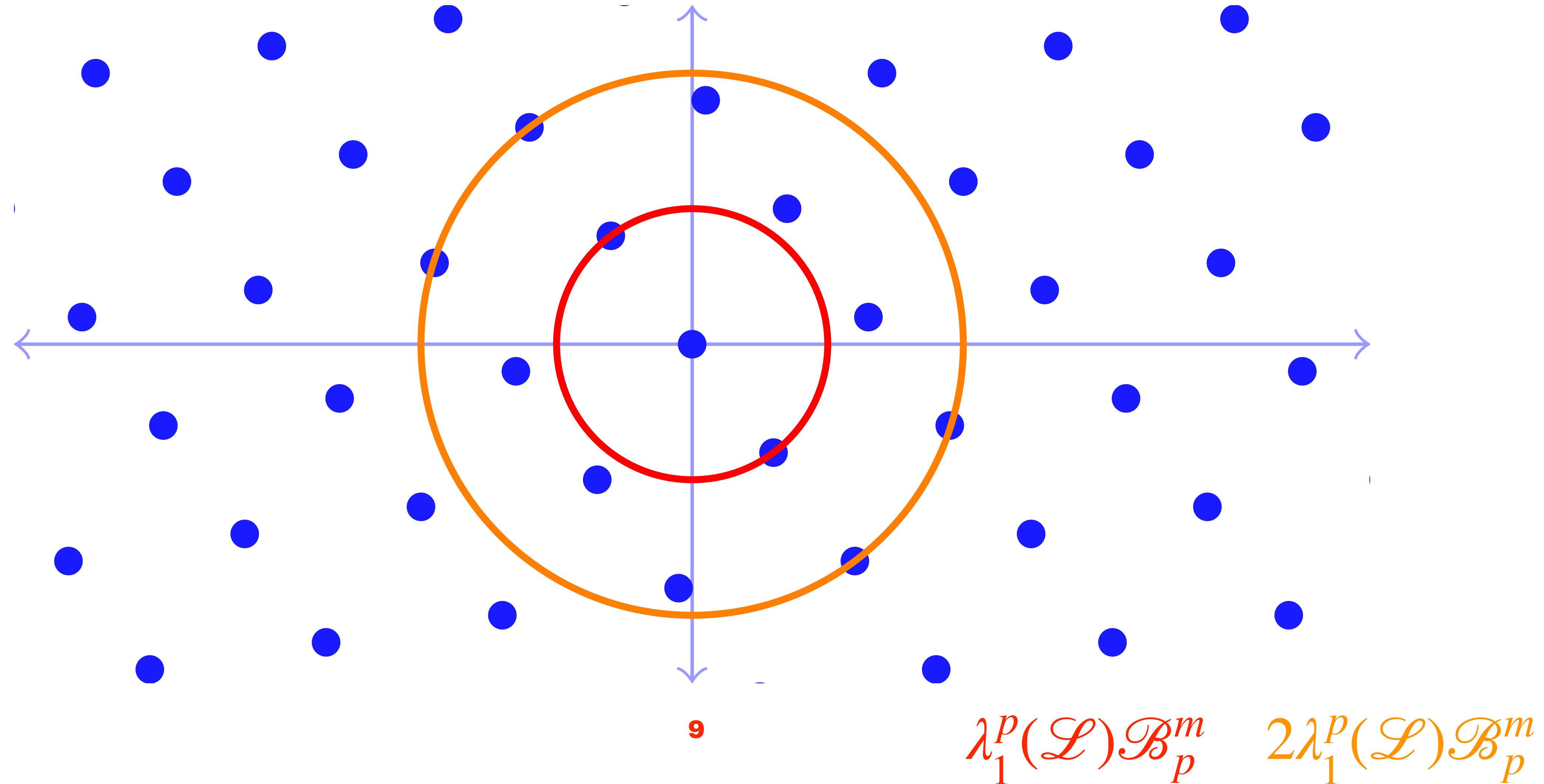
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- For any lattice \mathcal{L} and target vector t , $\text{dist}_p(t, \mathcal{L}) = \min_{u \in \mathcal{L}} \|u - t\|_p$.



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- ℓ_p ball, $r\mathcal{B}_p^m := \{\mathbf{x} \in \mathbb{R}^m : \|\mathbf{x}\|_p \leq r\}$.



Shortest Vector Problem

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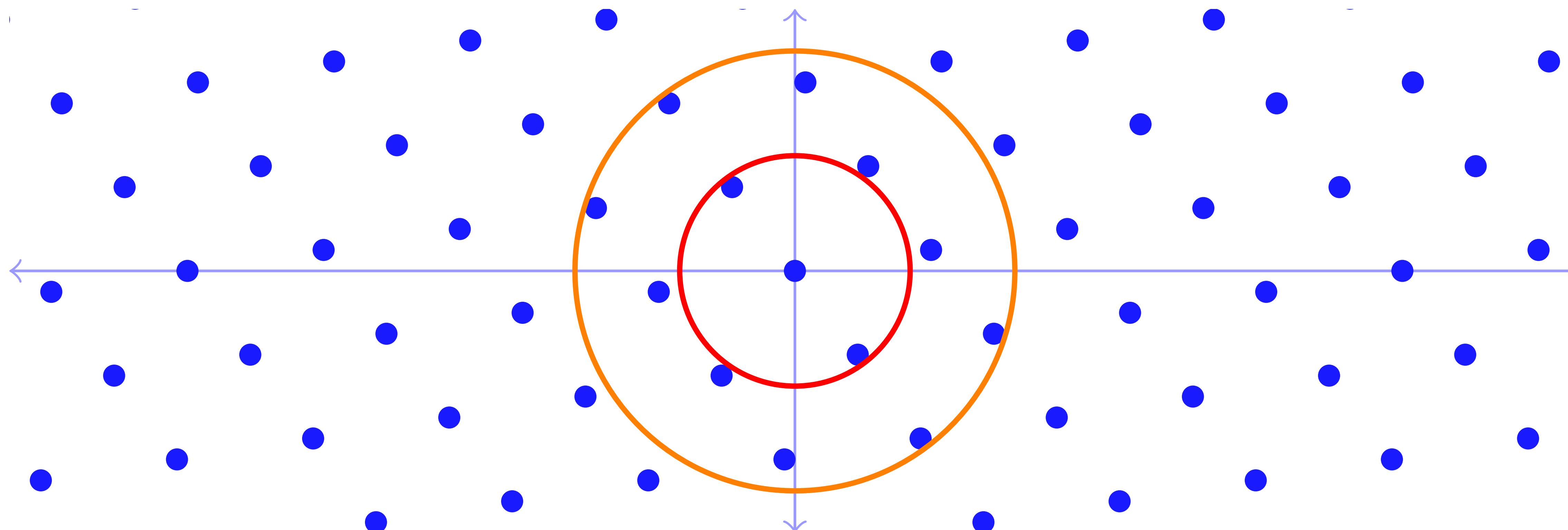
Shortest Vector Problem (γ -SVP _{p}) : Given a basis \mathbf{B} of lattice \mathcal{L} , find a lattice vector \mathbf{v} such that,

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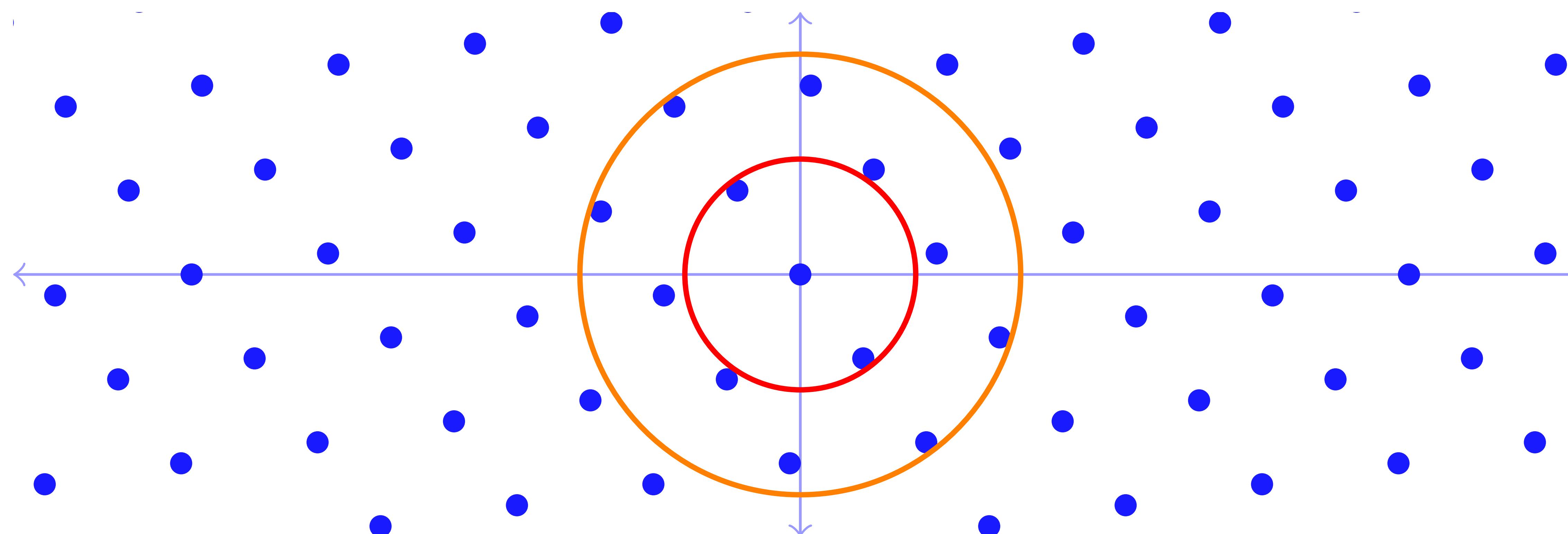
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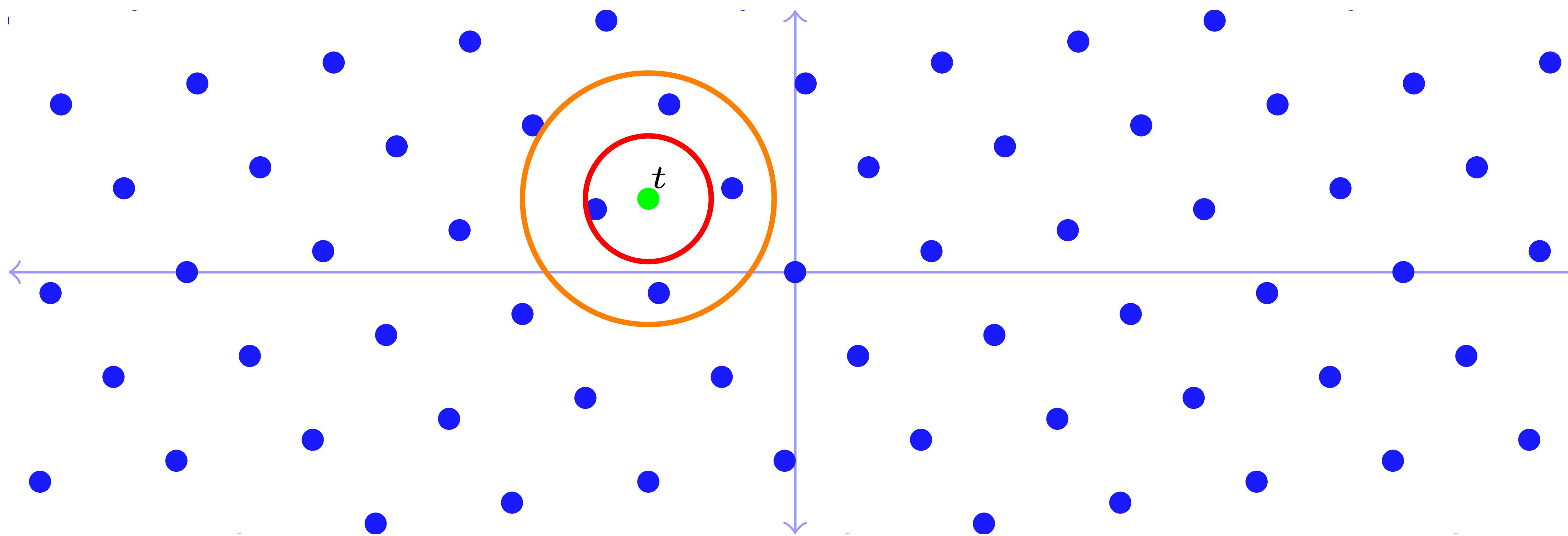
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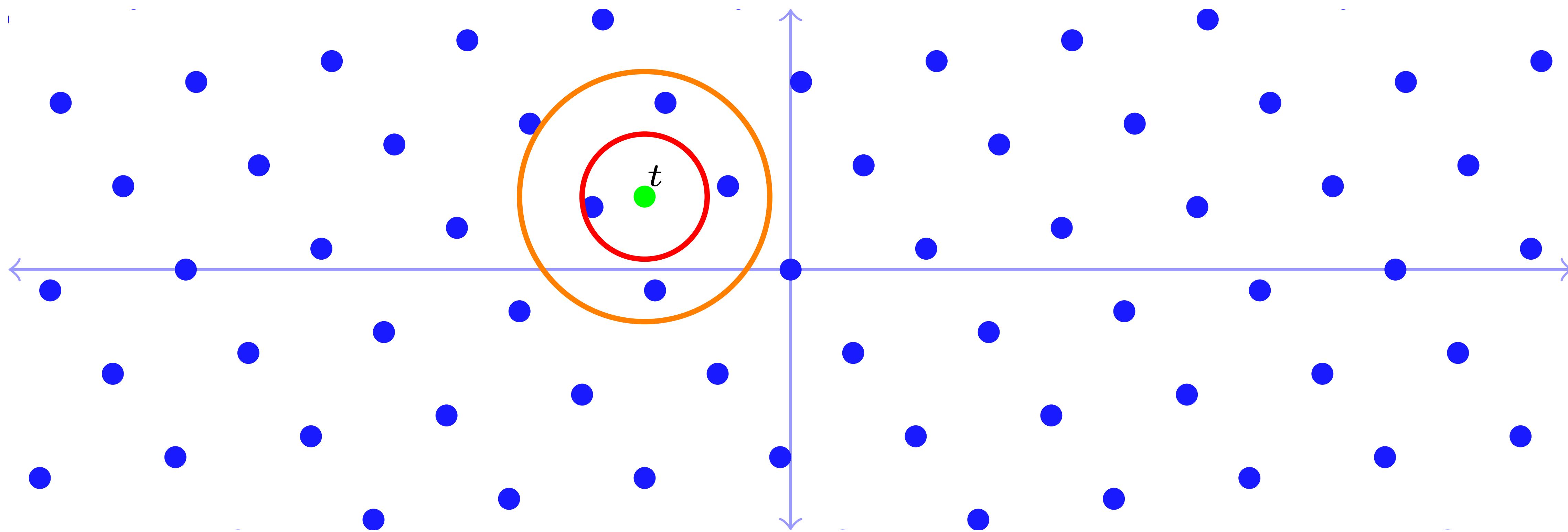
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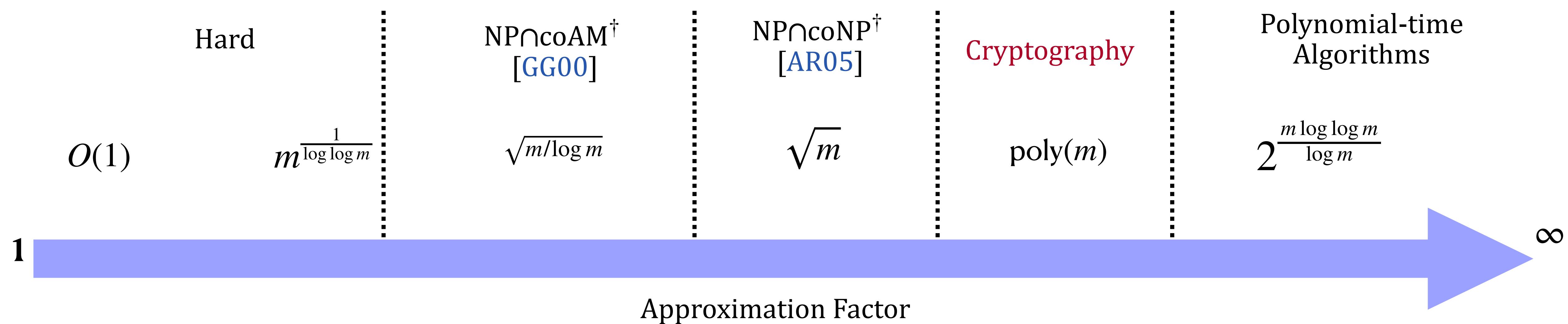
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$\text{dist}_p(\mathbf{t}, \mathcal{L}) \mathcal{B}_p^m$

$2\text{dist}_p(\mathbf{t}, \mathcal{L}) \mathcal{B}_p^m$

Approximation of SVP and CVP



[†] Only for ℓ_2 norm. Extend to ℓ_p norm by [RR06]

Reduction between Lattice Problems

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- When $p > 2$, there is super constant increment in dimension.

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- $2^{(1-\epsilon)n}$ lower bound for small constant approximation of \mathbf{CVP}_p and \mathbf{SVP}_∞ assuming (gap) Strong Exponential time Hypothesis [[BGS17; AS18; Aggarwal, Bennett, Golovnev, Stephens-Davidowitz '20](#)].

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- Uses some specific properties of the $O(1)$ -**SVP**₂ algorithm.

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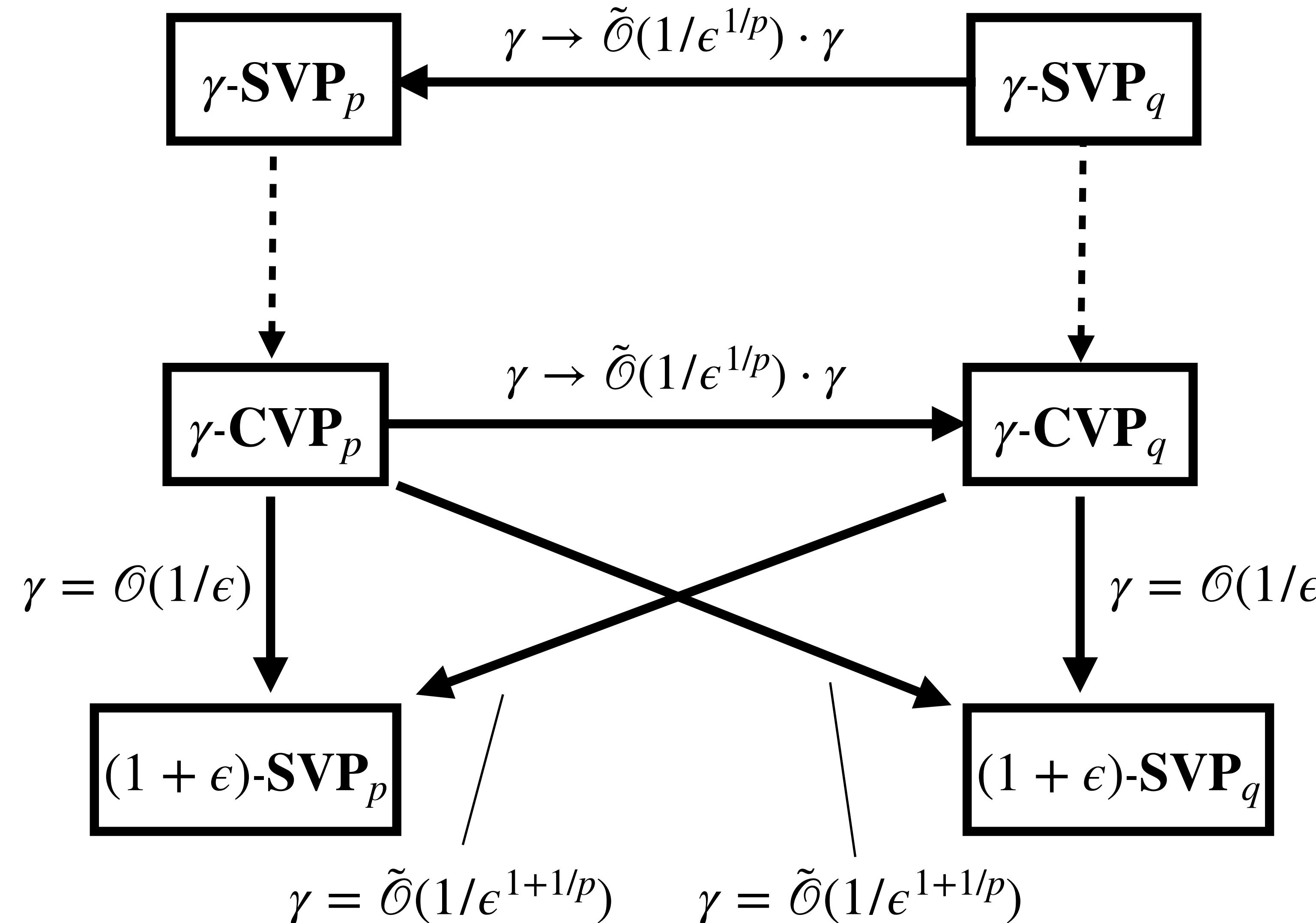
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Either preserves the rank
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Our Result: Exponential time reductions

$$p \leq q$$



- Dimension and rank increase by at most one.
- Dotted lines represent polynomial time reduction by Goldreich, Micciancio, Safra, Seifert '99.
- Solid lines represent $2^{\epsilon m}$ time reductions .

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- Improves on Kannan's celebrated polynomial time reduction from \sqrt{m} - \mathbf{CVP}_2 to \mathbf{SVP}_2 .

Reduction from CVP_q to SVP_p

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- Eisenbrand and Venzin observed that it suffices to simultaneously find a lattice vector close to $k\mathbf{t}$ and a lattice vector close to $(k - 1)\mathbf{t}$.

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- Otherwise, repeat this.

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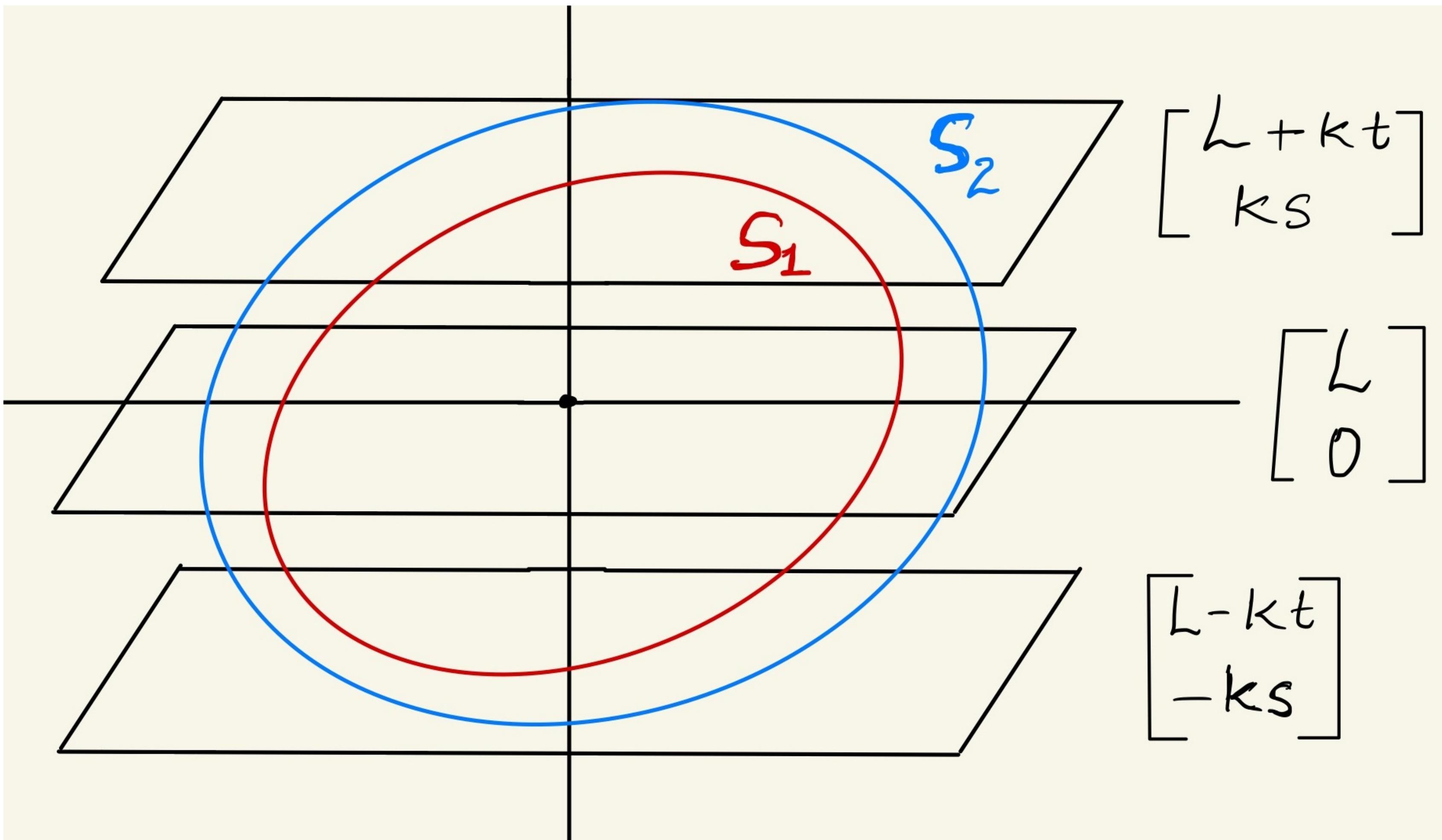
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- We use generalisation of this to any ℓ_p norm, to sample a uniformly random lattice vector.

Proof sketch

- $\exists r > 0, r' = r - m^{1/p-1/q} \text{dist}_q(\mathbf{t}, \mathcal{L}) - s, S_1 = r' \mathcal{B}_p^m, S_2 = r \mathcal{B}_p^m$ such that

$$\frac{|S_2|}{|S_1|} \leq 2^{\epsilon m}.$$
- $S \subseteq S_1 \cap (\mathcal{L} + k'\mathbf{t}, k's)$
- $T = S + (-\mathbf{v} + \mathbf{t}, s) \in S_2$
- $2^{2\epsilon m} |S| = 2^{2\epsilon m} |T| \geq |S_2|$



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Questions?