

# The Subspace Flatness Conjecture and Faster Integer Programming

**Thomas Rothvoss**

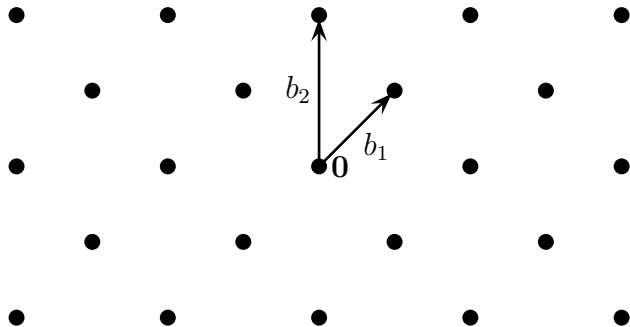
Joint work with Victor Reis



UNIVERSITY *of*  
WASHINGTON

# Covering radius

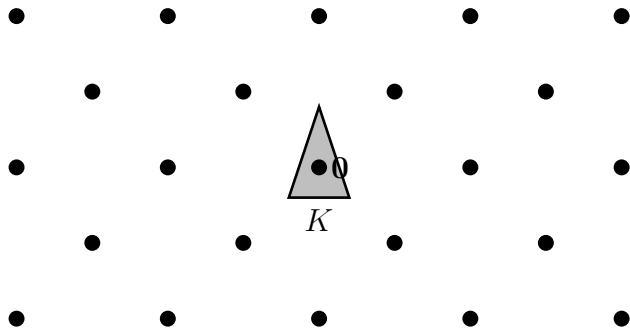
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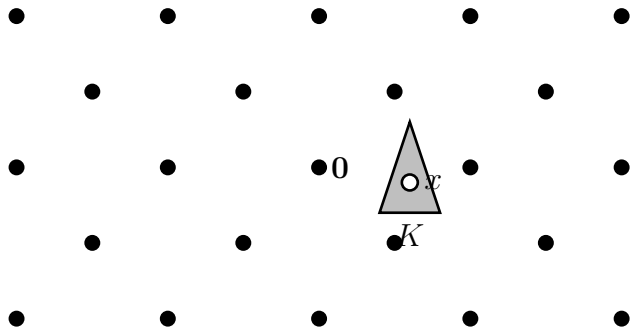
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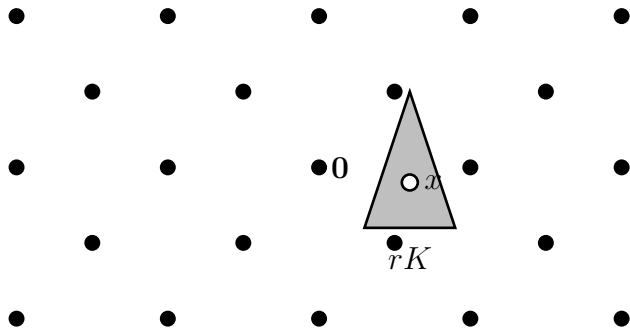
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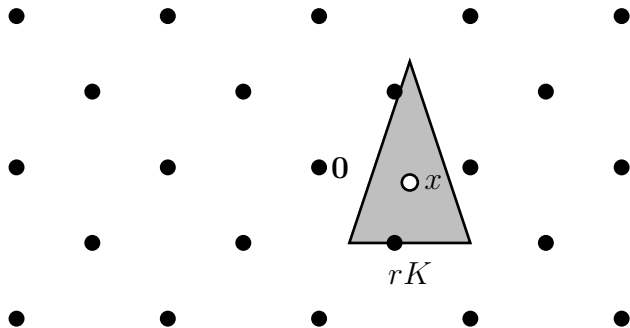
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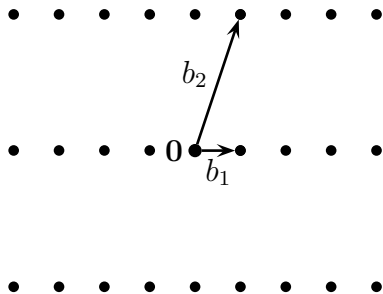
# Lower bounds on the covering radius

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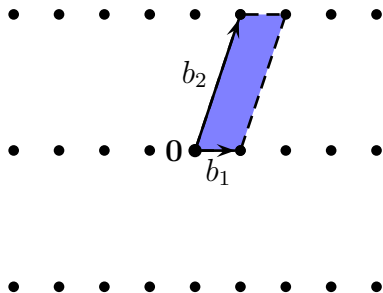
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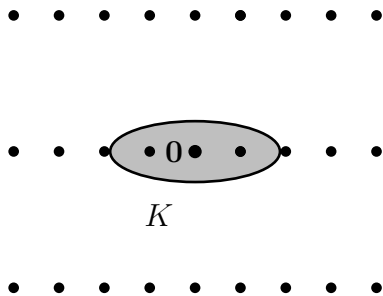


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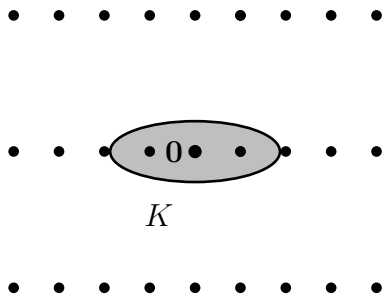
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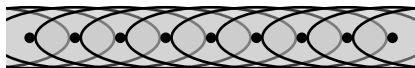
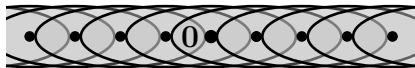
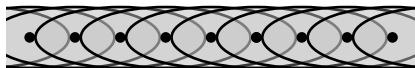
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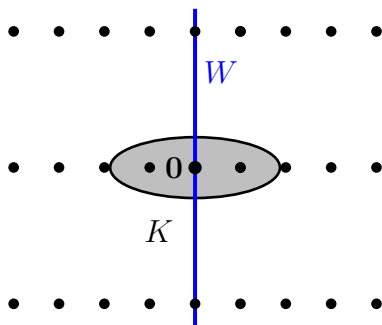
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- ▶ Simple lower bound:  $\mu(\Lambda, K) \geq \left(\frac{\det(\Lambda)}{\text{Vol}_n(K)}\right)^{1/n}$
- ▶ For any subspace  $\mu(\Lambda, K) \geq \mu(\Pi_W(\Lambda), \Pi_W(K))$

# Kannan, Lovász (1988)

- ▶ Consider the best volume-based lower bound

$$\mu_{KL}(\Lambda, K) = \max_{\substack{W \subseteq \text{span}(\Lambda) \text{ subspace} \\ d := \dim(W)}} \left( \frac{\det(\Pi_W(\Lambda))}{\text{Vol}_d(\Pi_W(K))} \right)^{1/d}$$

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## Theorem (Kannan, Lovász (1988))

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## Subspace Flatness Conjecture (Dadush 2012)

For full rank lattice  $\Lambda \subseteq \mathbb{R}^n$  and convex body  $K \subseteq \mathbb{R}^n$  one has

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- ▶ Dadush shows consequences for solving IPs.



# Main results

## Theorem (Reis, R.'23)

For full rank lattice  $\Lambda \subseteq \mathbb{R}^n$  and convex body  $K \subseteq \mathbb{R}^n$  one has

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## Theorem (Reis, R.'23)

For convex body  $K \subseteq \mathbb{R}^n$  one can find a point in  $K \cap \mathbb{Z}^n$  in time  $(\log n)^{O(n)}$ .

## Previously best known:

- ▶  $2^{O(n^2)}$  [Lenstra '83]
- ▶  $n^{O(n)}$  [Kannan '87]
- ▶  $2^{O(n)}n^n$  [Dadush '12], [Dadush, Eisenbrand, R. '22]

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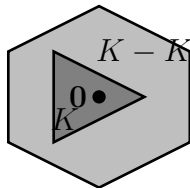


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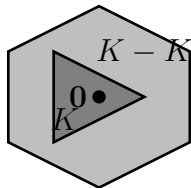


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Flatness constant in dimension  $n$  is at most  $O(n \log^8(n))$ .

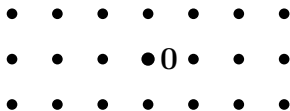
- ▶ Previously best known:  $O(n^{4/3} \log^{O(1)} n)$   
[Rudelson '98+Banaszczyk, Litvak, Pajor, Szarek '99]

# The Reverse Minkowski Theorem

## Reverse Minkowski Theorem (Regev, Stephens-Da.)

Let  $\Lambda \subseteq \mathbb{R}^n$  be a lattice that satisfies  $\det(\Lambda') \geq 1$  for all sublattices  $\Lambda' \subseteq \Lambda$ . Then for  $s = \Theta(\log n)$ ,

$$\rho_{1/s}(\Lambda) = \sum_{x \in \Lambda} \exp(-\pi s^2 \|x\|_2^2) \leq \frac{3}{2}$$

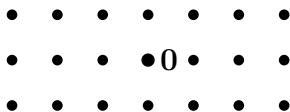


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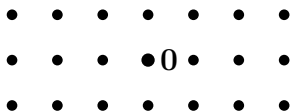
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## Definition

A lattice  $\Lambda$  is called **stable** if  $\det(\Lambda) = 1$  and  $\det(\Lambda') \geq 1$  for all sublattices  $\Lambda' \subseteq \Lambda$ .



## $\ell$ -position

- ▶ For a symmetric convex body  $K \subseteq \mathbb{R}^n$ ,

$$\ell_K = \mathbb{E}_{x \sim N(0, I_n)} [\|x\|_K^2]^{1/2}$$

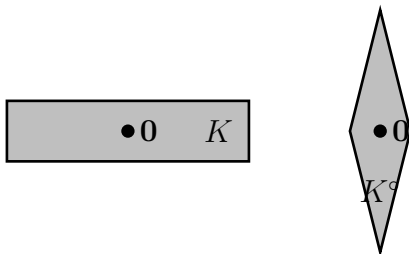
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- ▶ **Polar** is  $K^\circ = \{x \in \mathbb{R}^n \mid \langle x, y \rangle \leq 1 \ \forall y \in K\}$
- ▶ Possible that  $\ell_K$  and  $\ell_{K^\circ}$  arbitrarily large



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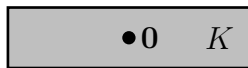
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## Theorem (Figiel, Tomczak-Jaegerman, Pisier)

For any symmetric convex body  $K \subseteq \mathbb{R}^n$ , there is an invertible linear map  $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$  so that  $\ell_{T(K)} \cdot \ell_{(T(K))^\circ} \leq O(n \log n)$ .



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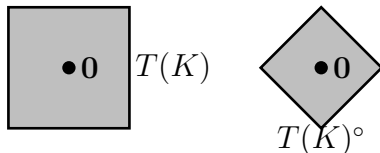
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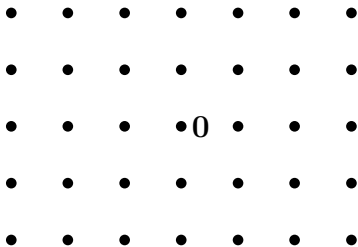
## Proof uses:

- ▶ Machinery of [Banaszczyk 96]
- ▶  $\rho_{1/s}(\Lambda^* \setminus \{\mathbf{0}\}) \leq \frac{1}{2}$  by **Reverse Minkowski Theorem** for  $s = \Theta(\log n)$ .

# Quotient lattices

## Definition

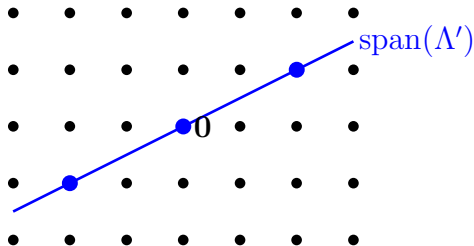
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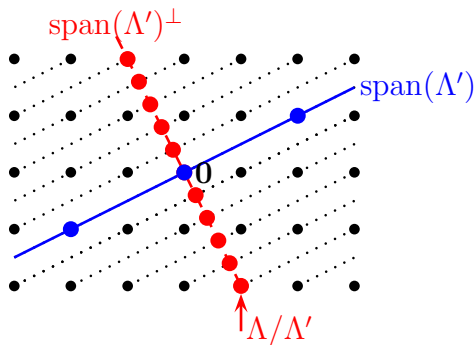




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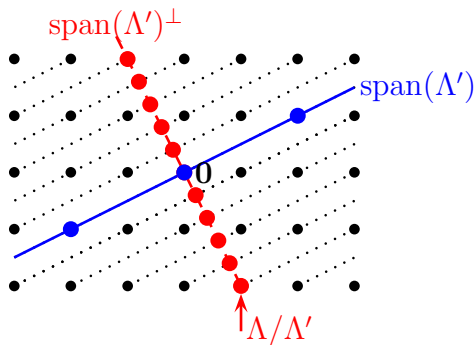
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- ▶ **Intuition:** We can factor  $\Lambda$  into  $\Lambda'$  and  $\Lambda/\Lambda'$
- ▶ For example  $\det(\Lambda) = \det(\Lambda') \cdot \det(\Lambda/\Lambda')$ .

# The canonical filtration

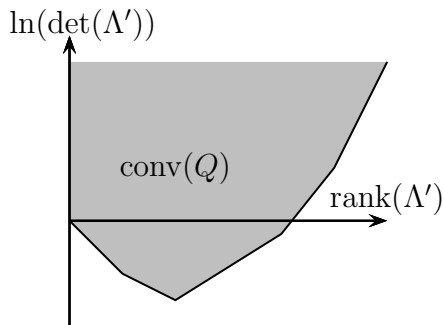
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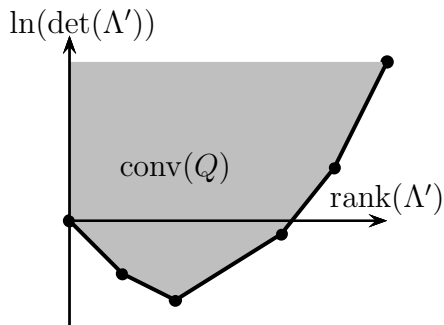


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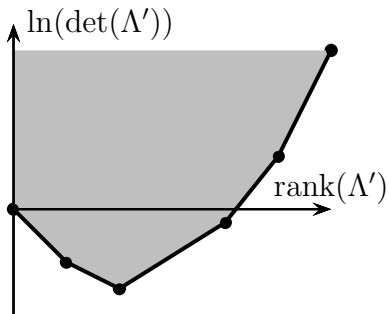
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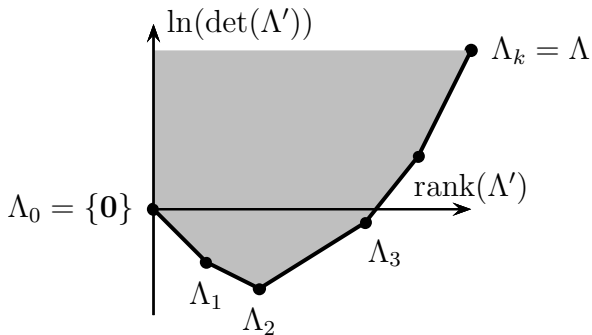
- ▶ Lower envelope of  $\text{conv}(Q)$  is called **canonical polygon**



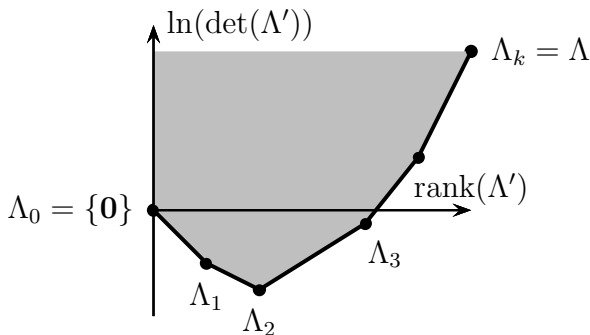
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### Theorem (Canonical filtration)

(a) *The vertices of the canonical plot form a chain*

$$\{0\} = \Lambda_0 \subset \Lambda_1 \subset \dots \subset \Lambda_k = \Lambda.$$

(b)  $r_i := \det(\Lambda_i/\Lambda_{i-1})^{1/\text{rank}(\Lambda_i/\Lambda_{i-1})}$  satisfy  $r_1 < \dots < r_k$

(c) *Each  $\frac{1}{r_i}(\Lambda_i/\Lambda_{i-1})$  is stable.*



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- ▶ Consider canonical filtration  $\{\mathbf{0}\} = \Lambda_0 \subset \dots \subset \Lambda_k = \Lambda$ .
- ▶ Define

$$d_i := \text{rank}(\Lambda_i/\Lambda_{i-1}) \quad \text{and} \quad r_i := \det(\Lambda_i/\Lambda_{i-1})^{1/d_i}$$

- ▶ **Goal:**
  - ▶ Bound  $\mu(\Lambda, K)$  in terms of  $r_i, d_i$ .
  - ▶ Bound  $r_i, d_i$  in terms of  $\mu_{KL}(\Lambda, K)$ .

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► Recall:  $\frac{1}{r_i}(\Lambda_i/\Lambda_{i-1})$  stable

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- ▶ Recall:  $\frac{1}{r_i}(\Lambda_i/\Lambda_{i-1})$  stable
- ▶ Use monotonicity of  $\ell$ -value:  $\ell_{K \cap W} \leq \ell_K$  for subspace  $W$   
(potentially huge loss!!)



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- ▶ Recall:  $\frac{1}{r_i}(\Lambda_i/\Lambda_{i-1})$  stable
- ▶ Use monotonicity of  $\ell$ -value:  $\ell_{K \cap W} \leq \ell_K$  for subspace  $W$   
(potentially huge loss!!)
- ▶ Group indices of similar density together:  $r_i \leq \frac{1}{2}r_{i+2}$

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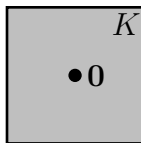
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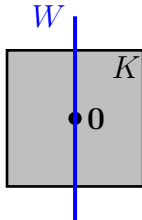
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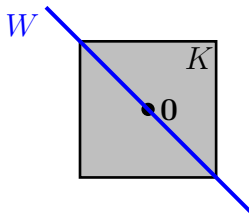
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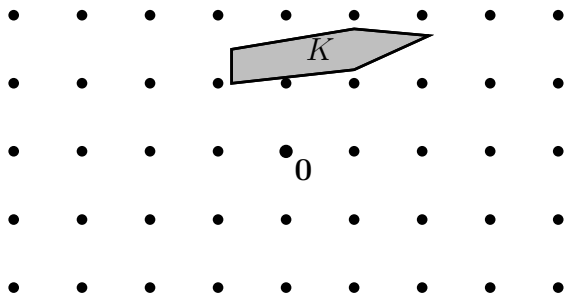
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# Dadush's algorithm

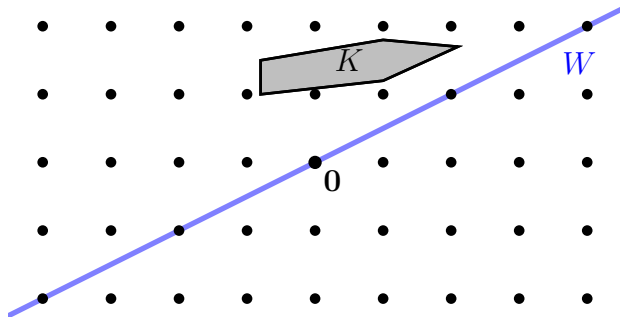


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**Output:** Point in  $K \cap \Lambda$

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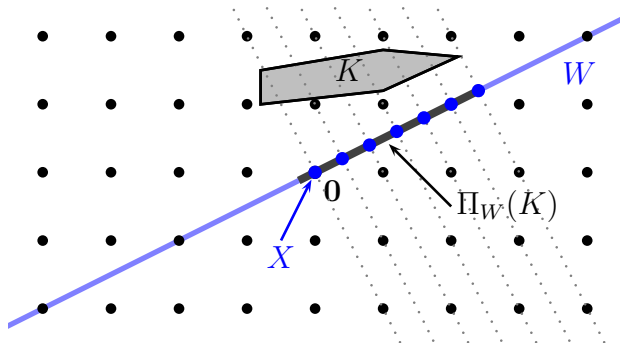
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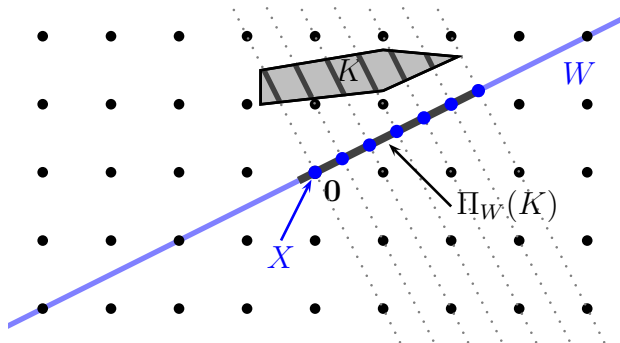


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For full rank lattice  $\Lambda \subseteq \mathbb{R}^n$  and convex body  $P \subseteq \mathbb{R}^n$  one has

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Thanks for your attention!

# Generalization to non-symmetric $K$

- ▶ Translate  $K$  so that **Santaló point** is at  $\mathbf{0}$  (i.e. barycenter of  $K^\circ$  is at  $\mathbf{0}$ ).
- ▶ Run proof with  $K_{\text{sym}} := K \cap (-K)$  (inner symmetrizer)
- ▶ We need:

$$\text{Vol}_d(\Pi_W(K))^{1/d} \lesssim \left(\frac{n}{d}\right)^3 \cdot \text{Vol}_d(\Pi_W(K_{\text{sym}}))^{1/d}$$

More involved! This is a polar version of Rudelson's result on sections of the difference body:

$$\text{Vol}_d((K - K) \cap W)^{1/d} \lesssim \frac{n}{d} \cdot \max_{x \in \mathbb{R}^n} \left\{ \text{Vol}_d(K \cap (x + W))^{1/d} \right\}$$