



# Introduction to lattices

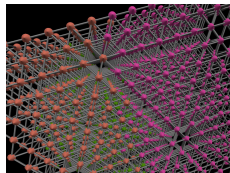
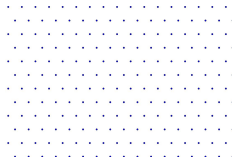
Barthel Jim

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Luxembourg National  
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# What is a lattice?



[http://www.gracebyte.com/lattice/  
images/ss\\_2.jpg](http://www.gracebyte.com/lattice/images/ss_2.jpg)



[https://docplayer.nl/19815420-Hoofdstuk-  
4-atoombouw-en-periodiek-systeem.html](https://docplayer.nl/19815420-Hoofdstuk-4-atoombouw-en-periodiek-systeem.html)



[http://www.aamt.edu.au/digital-  
resources/R10266/index.html](http://www.aamt.edu.au/digital-resources/R10266/index.html)

# Table of contents

- 1 Part I: Definitions
- 2 Part II: Comparing lattices
- 3 Part III: Gram-Schmidt Orthogonalization
- 4 Part IV: Determinant
- 5 Part V: Successive minima
- 6 Part VI: Minkowski's theorems
- 7 Part VII: Computational problems
- 8 References:

Part I:  
Definitions

Part II:  
Comparing  
lattices

Part III:  
Gram-Schmidt  
Orthogonal-  
ization

Part IV:  
Determinant

Part V:  
Successive  
minima

Part VI:  
Minkowski's  
theorems

Part VII:  
Computational  
problems

References:

# Part I: Definitions

# The formal definition of a lattice

## Definition (lattice)

A *lattice* is a discrete additive subgroup of  $\mathbb{R}^n$ .

In other words, a *lattice* is a subset  $\Lambda \subseteq \mathbb{R}^n$  satisfying the following properties:

- 1 (Subgroup property)  $\Lambda$  is closed under addition and subtraction.
- 2 (Discreteness) There is an  $\epsilon > 0$  such that any two distinct lattice points  $x \neq y \in \Lambda$  are at distance at least  $\|x - y\| \geq \epsilon$ .

# Constructing lattices

## Definition 1 (lattice generated by linearly independent vectors)

Let  $b_1, \dots, b_n \in \mathbb{R}^m$  be linearly independent vectors.

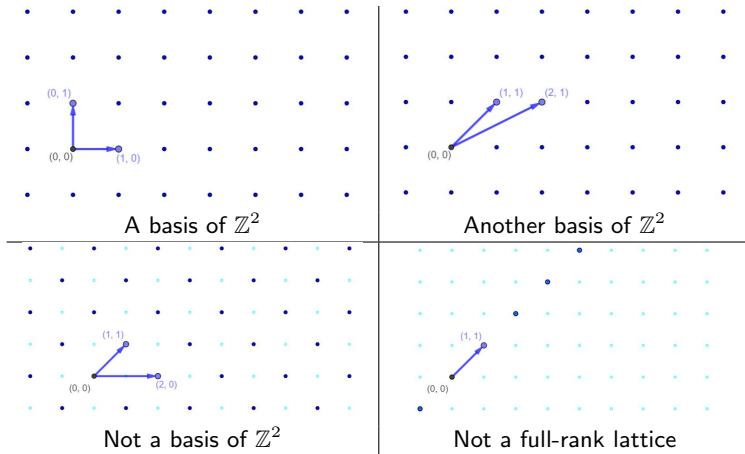
Let  $B = [b_1, \dots, b_n]$ .

- ① The lattice *generated by*  $B$  is the set

$$\mathcal{L}(B) = \{Bx \in \mathbb{Z}^m : x \in \mathbb{Z}^n\} = \left\{ \sum_{i=1}^n x_i b_i : x_i \in \mathbb{Z} \right\}.$$

- ② The matrix  $B$  is called the *basis* of the lattice  $\mathcal{L}(B)$ .
- ③ We call  $n$  the *rank* of  $\mathcal{L}(B)$  and  $m$  the *dimension* of  $\mathcal{L}(B)$ .
- ④ If  $n = m$ , then  $\mathcal{L}(B)$  is called a *full rank* lattice.

# Examples of lattices



## Definitions emerging from lattices

Part I:  
DefinitionsPart II:  
Comparing  
latticesPart III:  
Gram-Schmidt  
Orthogonal-  
izationPart IV:  
DeterminantPart V:  
Successive  
minimaPart VI:  
Minkowski's  
theoremsPart VII:  
Computational  
problems

## References:

## Definition 2

Let  $B$  be any lattice basis and let  $\mathcal{L}(B)$  be the corresponding lattice.

- ① The *span* of  $\mathcal{L}(B)$  is the vector space generated by  $B$ :

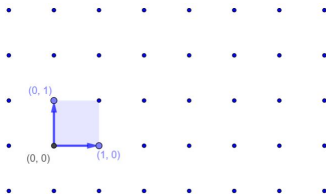
$$\text{span}(\mathcal{L}(B)) = \text{span}(B) = \langle B \rangle = \{Bx \in \mathbb{R}^m : x \in \mathbb{R}^n\}$$

- ② The fundamental parallelepiped of the lattice basis  $B$  is given by

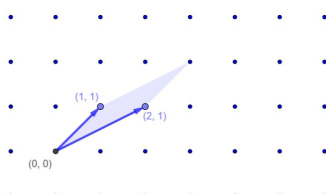
$$\begin{aligned} P(B) &= \{Bx \in \mathbb{R}^m : x \in \mathbb{R}^n, 0 \leq x_i < 1 \quad \forall 0 \leq i \leq n\} \\ &= \left\{ \sum_{i=1}^n x_i b_i : 0 \leq x_i < 1 \right\}. \end{aligned}$$



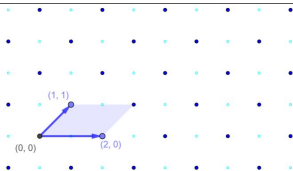
# More examples



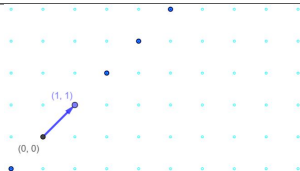
A basis of  $\mathbb{Z}^2$



Another basis of  $\mathbb{Z}^2$



Not a basis of  $\mathbb{Z}^2$



Not a full-rank lattice

# Part II: Comparing lattices

# A lattice and its possible bases (1)

## Lemma 3

Let  $\Lambda$  be a lattice of rank  $n$ , and let  $b_1, \dots, b_n \in \Lambda$  be linearly independent lattice vectors.

Then,  $b_1, \dots, b_n$  form a basis of  $\Lambda \Leftrightarrow P(b_1, \dots, b_n) \cap \Lambda = \{0\}$ .

## A lattice and its possible bases (2)

Part I:  
Definitions

Part II:  
Comparing  
lattices

Part III:  
Gram-Schmidt  
Orthogonal-  
ization

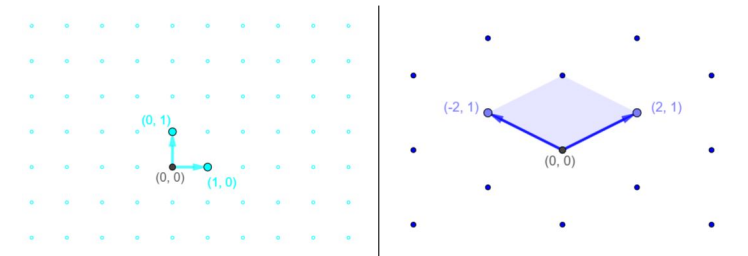
Part IV:  
Determinant

Part V:  
Successive  
minima

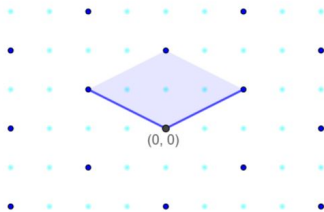
Part VI:  
Minkowski's  
theorems

Part VII:  
Computational  
problems

References:



We compare both lattices by superposing them:



# A lattice and its possible bases (3)

## **Proof of lemma 3:**

1)  $b_1, \dots, b_n$  form a basis of  $\Lambda \Rightarrow P(b_1, \dots, b_n) \cap \Lambda = \{0\}$ :

- By definition,

$$\Lambda = \left\{ \sum x_i b_i : x_i \in \mathbb{Z} \right\}.$$

- Furthermore,

$$P(b_1, \dots, b_n) = \left\{ \sum x_i b_i : 0 \leq x_i < 1 \right\}.$$

- Hence,

$$P(b_1, \dots, b_n) \cap \Lambda = \{0\}.$$

## A lattice and its possible bases (4)

2)  $P(b_1, \dots, b_n) \cap \Lambda = \{0\} \Rightarrow b_1, \dots, b_n$  form a basis of  $\Lambda$ :

① Since  $b_1, \dots, b_n \in \Lambda$ ,  $\mathcal{L}(b_1, \dots, b_n) \subseteq \Lambda$ .

② • Since  $\Lambda$  is a lattice of rank  $n$  and  $b_1, \dots, b_n$  are  $n$  linearly independent lattice vectors of  $\Lambda$ ,

$$\forall x \in \Lambda : x = \sum x_i b_i \quad (x_i \in \mathbb{R}).$$

• Let

$$x' = \sum \lfloor x_i \rfloor b_i \in \Lambda.$$

• Let

$$x'' = x - x' = \sum (x_i - \lfloor x_i \rfloor) b_i.$$

• Since  $\Lambda$  is closed under addition and subtraction

$$x'' \in \Lambda.$$

• Since  $0 \leq x_i - \lfloor x_i \rfloor < 1$  for all  $1 \leq i \leq n$ ,

$$x'' \in P(b_1, \dots, b_n).$$

## A lattice and its possible bases (4)

2)  $P(b_1, \dots, b_n) \cap \Lambda = \{0\} \Rightarrow b_1, \dots, b_n$  form a basis of  $\Lambda$ :

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$$\forall x \in \Lambda : x = \sum x_i b_i \ (x_i \in \mathbb{R}).$$

• Let

$$x' = \sum \lfloor x_i \rfloor b_i \in \Lambda.$$

• Let

$$x'' = x - x' = \sum (x_i - \lfloor x_i \rfloor) b_i \in P(b_1, \dots, b_n) \cap \Lambda.$$

• Since  $P(b_1, \dots, b_n) \cap \Lambda = \{0\}$ ,  $x'' = 0$ .

• Since  $b_1, \dots, b_n$  are linearly independent,

$$x_i = \lfloor x_i \rfloor \text{ for all } 1 \leq i \leq n.$$

In particular  $x_i$  is an integer for all  $1 \leq i \leq n$ .

• Hence,  $x \in \mathcal{L}(b_1, \dots, b_n)$  and so

$$\Lambda \subseteq \mathcal{L}(b_1, \dots, b_n).$$

# Equivalence of bases (1)

## Definition 4 (equivalence of lattices)

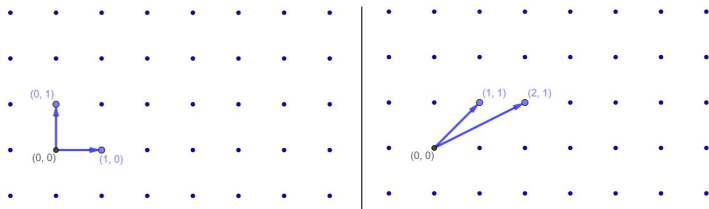
Let  $B_1, B_2$  be lattice bases. We say that  $B_1$  is *equivalent* to  $B_2$  if and only if  $\mathcal{L}(B_1) = \mathcal{L}(B_2)$ .

## Lemma 5

Two bases  $B_1, B_2$  of rank  $n$  are equivalent if and only if there exists a unimodular matrix  $U$  (i.e.  $U$  is a square matrix with integer coefficients and  $\det(U) = \pm 1$ ) such that  $B_2 = B_1 U$ .



## Equivalence of bases (2)



Note that

$$\begin{pmatrix} 1 & 2 \\ 1 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 1 & 2 \end{pmatrix}$$

and that

$$\det \begin{pmatrix} 1 & 2 \\ 1 & 1 \end{pmatrix} = -1.$$

## Equivalence of bases (3)

**Proof of lemma 5:**

1)  $B_1, B_2$  are equivalent  $\Rightarrow \exists U$  unimodular such that  $B_2 = B_1 U$ :

- Since  $B_1$  and  $B_2$  are equivalent,

$$\mathcal{L}(B_1) = \mathcal{L}(B_2).$$

- Since  $\forall 1 \leq i \leq n : b_i \in B_2$ ,

$$b_i \in \mathcal{L}(B_2) = \mathcal{L}(B_1).$$

- By definition of the lattice  $\mathcal{L}(B_1)$ ,

$$\exists u_i \in \mathbb{Z}^n \text{ such that } b_i = B_1 u_i.$$

- Let  $U = (u_1, \dots, u_n)$ . Then clearly,

$$B_2 = B_1 U.$$

- Similarly, one can construct  $V \in \mathbb{Z}^{n \times n}$  such that

$$B_1 = B_2 V.$$

## Equivalence of bases (3)

**Proof of lemma 5:**

1)  $B_1, B_2$  are equivalent  $\Rightarrow \exists U$  unimodular such that  
 $B_2 = B_1 U$ :

- We deduce that  $B_2 = B_2 V U$  and so

$$B_2(Id - VU) = 0.$$

- Since the column vectors of  $B_2$  are linearly independent, its inverse exists and so

$$Id = VU.$$

- Since  $1 = \det(Id) = \det(V) \det(U)$  and  $U, V$  are integer matrices,

$$\det(U) = \pm 1.$$

## Equivalence of bases (4)

2)  $\exists U$  unimodular such that  $B_2 = B_1 U \Rightarrow B_1, B_2$  are equivalent:

- Since  $B_2 = B_1 U$  where  $B_2 = (b_1, \dots, b_n)$  and  $U = (u_1, \dots, u_n)$ ,

$$\forall 1 \leq i \leq n : b_i = B_1 u_i.$$

- Since  $U$  is unimodular,  $b_i \in \mathcal{L}(B_1)$  and hence

$$\mathcal{L}(B_2) \subseteq \mathcal{L}(B_1).$$

- Since any unimodular matrix has an inverse which is also unimodular, we first deduce that

$$B_1 = B_2 U^{-1}.$$

and then, the same argument as above yields

$$\mathcal{L}(B_1) \subseteq \mathcal{L}(B_2).$$

# Deducing one basis from another one

## Corollary 6

Two bases are equivalent if and only if one can be obtained from the other by the following operations on columns:

- 1  $b_i \leftarrow b_i + kb_j$  for some  $k \in \mathbb{Z}$  and  $i \neq j$ ,
- 2  $b_i \leftrightarrow b_j$ ,
- 3  $b_i \leftarrow -b_i$ .

# Part III: Gram-Schmidt Orthogonalization

# Gram-Schmidt orthogonalization

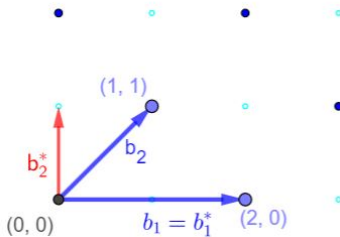
## Definition 7 (Gram-Schmidt orthogonalization)

Given any sequence of  $n$  linearly independent vectors  $b_1, \dots, b_n \in \mathbb{R}^m$ , we define their *Gram-Schmidt orthogonalization* as the sequence of vectors  $b_1^*, \dots, b_n^* \in \mathbb{R}^m$  defined recursively by

$$b_i^* = b_i - \sum_{j=1}^{i-1} \frac{\langle b_i, b_j^* \rangle}{\langle b_j^*, b_j^* \rangle} b_j^*.$$

In other words,  $b_i^*$  is the component of  $b_i$  orthogonal to  $b_1^*, \dots, b_{i-1}^*$ .

# An example of Gram-Schmidt



The vector  $b_2^*$  does not belong to the lattice.



# Properties of Gram-Schmidt orthogonalization

## Remark 8

Let  $b_1, \dots, b_n \in \mathbb{R}^m$  be  $n$  linearly independent vectors and let  $b_1^*, \dots, b_n^* \in \mathbb{R}^m$  be their Gram-Schmidt orthogonalization.

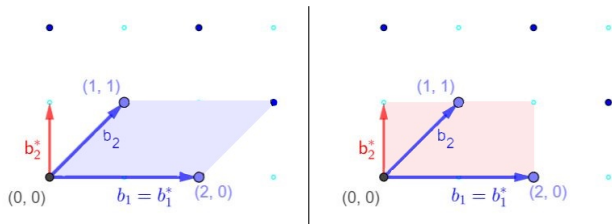
① (Orthogonality) For all  $i \neq j$  we have  $\langle b_i^*, b_j^* \rangle = 0$ .

② (Basis) For all  $1 \leq i \leq n$ ,  
 $\text{span}(b_1, \dots, b_i) = \text{span}(b_1^*, \dots, b_i^*)$ .

Note that in general  $\mathcal{L}(b_1, \dots, b_n) \neq \mathcal{L}(b_1^*, \dots, b_n^*)$  (most of the time  $b_i^* \notin \mathcal{L}(b_1, \dots, b_n)$ ) and that a lattice does not always admit an orthogonal basis!

③ (Order) The order of the Gram-Schmidt procedure matters.

# Volume of the fundamental parallelepiped



$$\text{vol}(P(b_1, b_2)) = \|b_1^*\| \|b_2^*\| = 2$$

## Remark 9

Let  $b_1, \dots, b_n \in \mathbb{R}^m$  be  $n$  linearly independent vectors and let  $b_1^*, \dots, b_n^* \in \mathbb{R}^m$  be their Gram-Schmidt orthogonalization.

Then:

$$\text{vol}(P(b_1, \dots, b_n)) = \prod_{i=1}^n \|b_i^*\|.$$

Part I:  
Definitions

Part II:  
Comparing  
lattices

Part III:  
Gram-Schmidt  
Orthogonal-  
ization

**Part IV:  
Determinant**

Part V:  
Successive  
minima

Part VI:  
Minkowski's  
theorems

Part VII:  
Computational  
problems

References:

# Part IV: Determinant

# Determinant

## Definition 10 (determinant of lattices)

Let  $\Lambda = \mathcal{L}(B)$  be a lattice of rank  $n$ . We define the *determinant* of  $\Lambda$  (denoted by  $\det(\Lambda)$ ) to be the  $n$ -dimensional volume of the fundamental parallelepiped  $P(B)$  associated to  $B$ . In symbols:

$$\det(\Lambda) = \text{vol}(P(B)) = \prod_{i=1}^n \|b_i^*\|.$$

# Properties of the determinant (1)

## Proposition 11

For any lattice basis  $B \in \mathbb{R}^{n \times m}$

- 1  $\det(\mathcal{L}(B)) = \sqrt{\det(B^T B)}$ ,
- 2 In particular if  $B \in \mathbb{R}^{n \times n}$  is a (non-singular) square matrix, then  $\det(\mathcal{L}(B)) = |\det(B)| = d$  and  $d\mathbb{Z}^n \subseteq \mathcal{L}(B)$ .
- 3 The determinant is independent of the basis.

# Properties of the determinant (2)

## Proof of proposition 11:

1)  $\det(\mathcal{L}(B)) = \sqrt{\det(B^T B)}$  :

- By the Gram-Schmidt orthogonalization procedure, we know that

$$B = B^* M$$

where  $M$  is an upper triangular matrix with 1's on the diagonal and

$$\frac{\langle b_i, b_j^* \rangle}{\langle b_j^*, b_j^* \rangle} \|b_j^*\| \text{ for all } j < i.$$

- Hence,

$$\sqrt{\det(B^T B)} = \sqrt{\det(M^T (B^*)^T B^* M)} = \sqrt{\det(M^T) \det((B^*)^T B^*) \det(M)}.$$

- Since  $M$  is upper triangular and has only 1's at its diagonal,

$$\det(M) = \det(M^T) = 1.$$

- Furthermore, by orthogonality of the columns of  $B^*$ ,

$$\det((B^*)^T B^*) = \prod_{i=1}^n (\|b_i^*\|)^2 = (\det(\mathcal{L}(B)))^2.$$

- Since  $\det(\mathcal{L}(B)) \geq 0$  by definition,

$$\sqrt{\det((B^*)^T B^*)} = \det(\mathcal{L}(B)).$$

## Properties of the determinant (3)

2) If  $B \in \mathbb{R}^{n \times n}$  is a (non-singular) square matrix, then  $\det(\mathcal{L}(B)) = |\det(B)| = d$  and  $d\mathbb{Z}^n \subseteq \mathcal{L}(B)$ .

- Since  $B$  is a square matrix,

$$\det(\mathcal{L}(B)) = \sqrt{\det(B^T B)} = \sqrt{(\det(B))^2} = |\det(B)|.$$

- Let  $v = dy \in d\mathbb{Z}^n$  where  $y \in \mathbb{Z}^n$ .
- Since  $B$  is non-singular, there is

$$x = B^{-1}dy \in \mathbb{R}^n.$$

- By Cramer's rule:

$$\begin{aligned} x_i &= \frac{\det((b_1, \dots, b_{i-1}, dy, b_{i+1}, \dots, b_n))}{\det(B)} \\ &= \pm \det((b_1, \dots, b_{i-1}, dy, b_{i+1}, \dots, b_n)) \in \mathbb{Z} \end{aligned}$$

- Thus,

$$x \in \mathbb{Z}^n.$$

- Hence,

$$v = Bx \in \mathcal{L}(B).$$

# Properties of the determinant (4)

3) The determinant is independent of the basis.

- Let  $B_1, B_2$  be equivalent bases. Then, there is a unimodular matrix  $U$  such that

$$B_2 = B_1 U.$$

- Thus,

$$\begin{aligned}\det(\mathcal{L}(B_2)) &= \sqrt{\det(B_2^T B_2)} \\ &= \sqrt{\det(U^T B_1^T B_1 U)} \\ &= \sqrt{(\det(U))^2 \det(B_1^T B_1)} \\ &= \sqrt{\det(B_1^T B_1)} \\ &= \det(\mathcal{L}(B_1))\end{aligned}$$



## Remarks about the determinant

## Remark 12

For any lattice basis  $B \in \mathbb{R}^{n \times m}$

① (Hadamard inequality)

$$\det(\mathcal{L}(B)) = \prod_{i=1}^n \|b_i^*\| \leq \prod_{i=1}^n \|b_i\|$$

(since  $\|b_i^*\| \leq \|b_i\|$ ).

- ② Geometrically, the determinant represents the inverse of the density of lattice points in space (e.g., the number of lattice points in a large and sufficiently regular region of space  $A$  should be approximately equal to the volume of  $A$  divided by the determinant.)

Small determinant = Dense lattice

# An example of a determinant

Part I:  
Definitions

Part II:  
Comparing  
lattices

Part III:  
Gram-Schmidt  
Orthogonal-  
ization

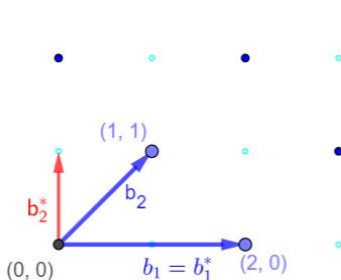
Part IV:  
Determinant

Part V:  
Successive  
minima

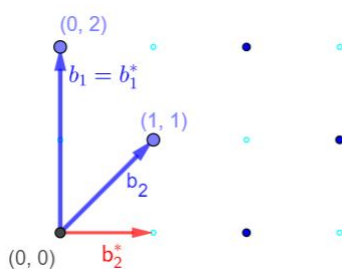
Part VI:  
Minkowski's  
theorems

Part VII:  
Computational  
problems

References:



By previous computations:  
 $\det(\mathcal{L}(b_1, b_2)) = \|b_1^*\| \|b_2^*\| = 2$   
 Hadamar's inequality is satisfied:  
 $\det(\mathcal{L}(b_1, b_2)) \leq 2\sqrt{2} = \|b_1\| \|b_2\|$



$$\begin{aligned} \det(\mathcal{L}(\begin{pmatrix} 0 & 1 \\ 2 & 1 \end{pmatrix})) &= \sqrt{\det((\begin{pmatrix} 0 & 2 \\ 1 & 1 \end{pmatrix})(\begin{pmatrix} 0 & 1 \\ 2 & 1 \end{pmatrix}))} \\ &= \sqrt{\det((\begin{pmatrix} 4 & 2 \\ 2 & 2 \end{pmatrix}))} = 2 \end{aligned}$$

Note

$$\det(\mathcal{L}(\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix})) = 1$$

Hence,  $\mathcal{L}(\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix})$  (i.e. dark and light blue points) is denser than  $\mathcal{L}(\begin{pmatrix} 0 & 1 \\ 2 & 1 \end{pmatrix})$ .

Part I:  
Definitions

Part II:  
Comparing  
lattices

Part III:  
Gram-Schmidt  
Orthogonal-  
ization

Part IV:  
Determinant

Part V:  
Successive  
minima

Part VI:  
Minkowski's  
theorems

Part VII:  
Computational  
problems

References:

# Part V: Successive minima

## Successive minima (1)

## Definition 13a (minimum distance)

Let  $\Lambda = \mathcal{L}(B)$  be a lattice of rank  $n$ .

The *minimum distance*  $\lambda_1$  of  $\Lambda$  is the smallest distance between any two lattice points:

$$\lambda_1(\Lambda) = \inf\{\|x - y\| : x, y \in \Lambda, x \neq y\}.$$

Equivalently, the *minimum distance* can be defined as the shortest non-zero vector of  $\Lambda$ :

$$\lambda_1(\Lambda) = \inf\{\|v\| : v \in \Lambda \setminus \{0\}\}.$$

Equivalently, the *minimum distance* is the smallest  $r > 0$  such that  $\Lambda$  contains at least one vector of length bounded by  $r$ ,

$$\lambda_1(\Lambda) = \inf\{r \in \mathbb{R}_{>0} : \dim(\text{span}(\Lambda \cap B(0, r))) \geq 1\}$$

where  $B(0, r) = \{x \in \mathbb{R}^m : \|x\| \leq r\}$  is the closed ball of radius  $r$  around 0.

## Successive minima (2)

## Definition 13b (successive minima)

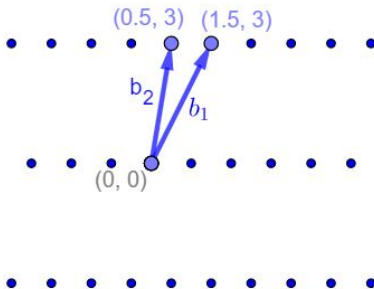
Let  $\Lambda = \mathcal{L}(B)$  be a lattice of rank  $n$ .

For  $i \in \{1, \dots, n\}$ , we define the  $i^{th}$  *successive minimum* as the smallest  $r > 0$  such that  $\Lambda$  contains at least  $i$  linearly independent vectors of length bounded by  $r$ ,

$$\lambda_i(\Lambda) = \inf\{r \in \mathbb{R}_{>0} : \dim(\text{span}(\Lambda \cap B(0, r))) \geq i\}$$

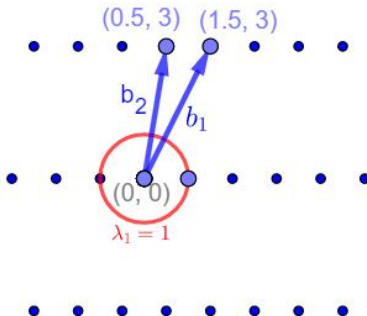
where  $B(0, r) = \{x \in \mathbb{R}^m : \|x\| \leq r\}$  is the closed ball of radius  $r$  around 0.

# An example of successive minima



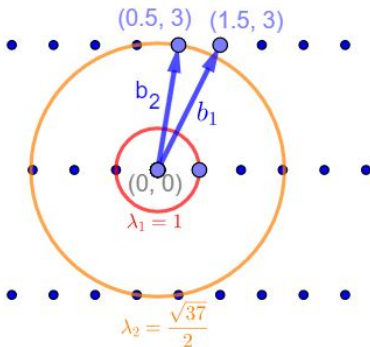
Consider  $\mathcal{L} \begin{pmatrix} 1/2 & 3/2 \\ 3 & 3 \end{pmatrix}$ .

# An example of successive minima



Start to grow a circle at the origin until you meet a point to find  $\lambda_1$ .

## An example of successive minima



Keep growing the circle until you meet a second point that lies not on the line given by the minimal vector to find  $\lambda_2$ .



## Rough lower bound (1)

### Theorem 14

Let  $B$  be a rank  $n$  lattice basis and let  $B^*$  be its Gram-Schmidt orthogonalization. Then:

$$\lambda_1(\Lambda) \geq \min_{i=1,\dots,n} \|b_i^*\| > 0.$$

Thus, for any two non-equal lattice points  $x, y \in \Lambda$

$$\|x - y\| \geq \min_{i=1,\dots,n} \|b_i^*\| > 0$$

## Rough lower bound (2)

Proof of theorem 14:

- Let,

$$Bx \in \mathcal{L}(B) \setminus \{0\}$$

be a generic lattice vector where  $x \in \mathbb{Z}^n \setminus \{0\}$ .

- Let

$$k = \max\{k \in \{1, \dots, n\} : x_k \neq 0\}.$$

- Then, by orthogonality

$$|\langle Bx, b_k^* \rangle| = \left| \sum_{i \leq k} \langle b_i x_i, b_k^* \rangle \right| = |x_k \langle b_k, b_k^* \rangle| = |x_k| \|b_k^*\|^2.$$

- By Cauchy-Schwartz,

$$|\langle Bx, b_k^* \rangle| \leq \|Bx\| \|b_k^*\|.$$

- Since  $|x_k| \geq 1$  and  $\|b_k^*\| \neq 0$ ,

$$\|b_k^*\| \leq \|Bx\|.$$

# The successive minima are achieved (1)

## Theorem 15

The successive minima of a lattice are achieved.

In other words, for every  $1 \leq i \leq n$ , there exists a vector  $v_i \in \Lambda$  with  $\|v_i\| = \lambda_i(\Lambda)$ .

# The successive minima are achieved (2)

## Proof of theorem 14:

- Let,

$$S = B(0, 2\lambda_1(\Lambda)) = \{x \in R^m : \|x\| < 2\lambda_1(\Lambda)\}.$$

- By definition of the minimal distance, there is at least one lattice point  $x \in S$ .

- Thus,

$$\lambda_1(\Lambda) = \inf\{\|x\| : x \in \Lambda \cap S \setminus \{0\}\}.$$

- Consider a small sphere of radius  $\frac{1}{2}\lambda_1(\Lambda)$  around each lattice point:

$$B\left(x, \frac{1}{2}\lambda_1(\Lambda)\right) \text{ for all } x \in \Lambda.$$

- Since the minimal distance between lattice points is  $\lambda_1(\Lambda)$ ,

$$B\left(x, \frac{1}{2}\lambda_1(\Lambda)\right) \cap B\left(y, \frac{1}{2}\lambda_1(\Lambda)\right) = \emptyset \text{ for all } x \neq y \in \Lambda.$$

- For all  $x \in S \cap \Lambda$ ,

$$B\left(x, \frac{1}{2}\lambda_1(\Lambda)\right) \subseteq B(0, 3\lambda_1(\Lambda)) = S'.$$

# The successive minima are achieved (3)

- Notice that:

$$\text{vol} \left( B \left( x, \frac{1}{2} \lambda_1(\Lambda) \right) \right) = C_n \left( \frac{1}{2} \lambda_1(\Lambda) \right)^n \text{ and}$$

$$\text{vol}(0, 3\lambda_1(\Lambda)) = C_n(3\lambda_1(\Lambda))^n$$

- Hence, there are at most  $6^n$  lattice points in  $S$ . So,

$$\lambda_1(\Lambda) = \inf \{ \|x\| : x \in \Lambda \cap S \setminus \{0\} \} = \min \{ \|x\| : x \in \Lambda \cap S \setminus \{0\} \}.$$

- By a similar argument, one proves the theorem for the other successive minima.

Part I:  
Definitions

Part II:  
Comparing  
lattices

Part III:  
Gram-Schmidt  
Orthogonal-  
ization

Part IV:  
Determinant

Part V:  
Successive  
minima

Part VI:  
Minkowski's  
theorems

Part VII:  
Computational  
problems

References:

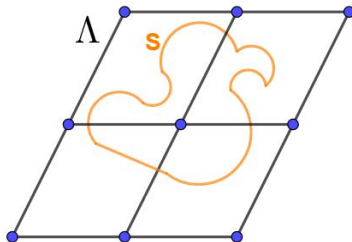
# Part VI: Minkowski's theorems

# Blichfeld's theorem (1)

## Theorem 16 (Blichfeld)

Let  $\Lambda = \mathcal{L}(B) \subseteq \mathbb{R}^n$  be a full-rank lattice and let  $S \subseteq \mathbb{R}^n$  be a subset with  $\text{vol}(S) > \det(\Lambda)$ . Then, there exist two nonequal points  $z_1, z_2 \in S$  such that  $z_1 - z_2 \in \Lambda$ .

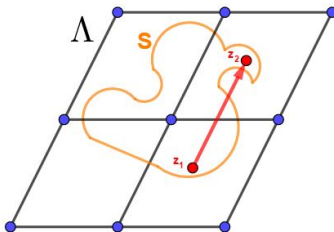
## Blichfeld example



Consider the lattice  $\Lambda$  and  $S \subseteq \mathbb{R}^n$  with  $\text{vol}(S) > \det(\Lambda)$ .

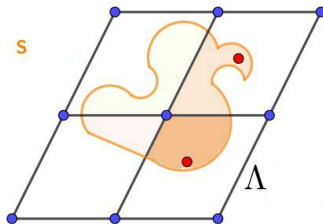


## Blichfeld example



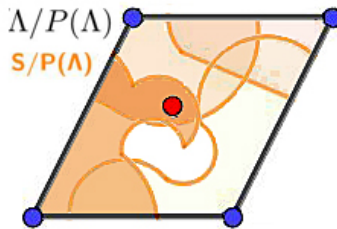
Then, we want to find  $z_1, z_2 \in S$  such that  $z_1 - z_2 \in \Lambda$ .

## Blichfeld example



To do so, consider  $\mathbb{R}^2$  partitioned by the lattice.

## Blichfeld example



Reduce all of  $\mathbb{R}^2$  to the fundamental parallelepiped and look for intersections, this gives us the two points we are looking for.

## Blichfeld's theorem (2)

**Proof of theorem 16:**

- As  $x$  ranges over all of  $\Lambda$ , we can partition  $\mathbb{R}^n$  by considering the sets

$$x + P(B) = \{x + y : y \in P(B)\}.$$

- For any  $x \in \Lambda$ , define

$$S_x = S \cap (x + P(B)).$$

- Since  $x + P(B)$  partitions  $\mathbb{R}^n$ , it does so with  $S$ . Hence,

$$S_x \cap S_y = \emptyset \ (\forall x \neq y) \text{ and } S = \cup_{x \in \Lambda} S_x.$$

- This implies that

$$\text{vol}(S) = \sum_{x \in \Lambda} \text{vol}(S_x).$$

- Translate the pieces  $S_x$  into the fundamental parallelepiped by defining

$$\hat{S}_x = S_x - x.$$

Clearly  $\hat{S}_x \subseteq P(B)$  and  $\text{vol}(S_x)$ .

## Blichfeld's theorem (3)

- Furthermore,

$$\sum_{x \in \Lambda} \text{vol}(\hat{S}_x) = \sum_{x \in \Lambda} \text{vol}(S_x) = \text{vol}(S) > \text{vol}(P(B)).$$

Thus, there must exist  $x \neq y \in \Lambda$  such that

$$\hat{S}_x \cap \hat{S}_y \neq \emptyset.$$

- Let  $z \in \hat{S}_x \cap \hat{S}_y$ . Then,

$$z + x \in S_x \subseteq S \text{ and } z + y \in S_y \subseteq S.$$

- Since  $x, y \in \Lambda$ ,

$$(z + x) - (z + y) = x - y \in \Lambda.$$

# Minkowski's convex body theorem

(1)

## Definition 17

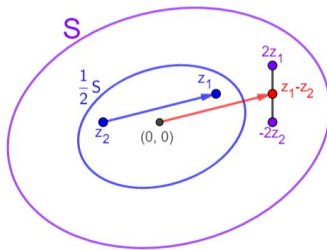
A subset  $S \subseteq \mathbb{R}^n$  is called:

- ① *centrally-symmetric* if for any  $x \in S$  we also have  $-x \in S$ ,
- ② *convex* if for any  $x, y \in S$  we also have  $\mu x + (1 - \mu)y \in S$  for all  $\mu \in [0, 1]$ .

## Theorem 18 (Minkowski - convex body)

Let  $\Lambda$  be a full-rank lattice of rank  $n$ . Then, any centrally-symmetric convex set  $S$  with  $\text{vol}(S) > 2^n \det(\Lambda)$  contains a non-zero lattice point.

# Minkowski convex body example



Since  $\text{vol}(S) > 4 \det(\Lambda)$ :

$$\text{vol}\left(\frac{1}{2}S\right) = \frac{1}{4}\text{vol}(S) > \det(\Lambda).$$

Minkowski's theorem implies the existence of a non-zero lattice point  $z_1 - z_2$  which happens to be also in  $S$ .

# Minkowski's convex body theorem

## (2)

Part I:  
DefinitionsPart II:  
Comparing  
latticesPart III:  
Gram-Schmidt  
Orthogonal-  
izationPart IV:  
DeterminantPart V:  
Successive  
minimaPart VI:  
Minkowski's  
theoremsPart VII:  
Computational  
problems

References:

### Proof of theorem 18:

- Define

$$\hat{S} = \frac{1}{2}S = \{x \in \mathbb{R}^n : 2x \in S\}.$$

- Clearly

$$\text{vol}(\hat{S}) = 2^{-n} \text{vol}(S) > \det(\Lambda).$$

- Blichfeld's theorem implies that

$$\exists z_1 \neq z_2 \in \hat{S} \text{ such that } 0 \neq z_1 - z_2 \in \Lambda.$$

- By definition,

$$2z_1, 2z_2 \in S.$$

- Since  $S$  is centrally-symmetric,

$$-2z_2 \in S.$$

- Since  $S$  is convex,

$$\frac{2z_1 - 2z_2}{2} = z_1 - z_2 \in S.$$



# Minkowski's first theorem (1)

## Theorem 19 (Minkowski - 1)

For any full-rank lattice  $\Lambda$  of rank  $n$ ,

$$\lambda_1(\Lambda) \leq \sqrt{n}(\det(\Lambda))^{1/n}.$$

$\sqrt{n}(\det(\Lambda))^{1/n}$  is called the *Minkowski bound*.

# Minkowski's first theorem (2)

## Proof of theorem 19:

- Consider the open sphere  $B(0, \lambda_1(\Lambda))$  centered at 0 of radius  $\lambda_1(\Lambda)$ .
- By definition,  $B(0, \lambda_1(\Lambda))$  is centrally-symmetric and convex but contains no non-zero lattice points.

- Minkowski's convex body theorem implies that

$$\text{vol}(B(0, \lambda_1(\Lambda))) \leq 2^n \det(\Lambda).$$

- Note that  $B(0, \lambda_1(\Lambda))$  contains a cube of side length  $\frac{2\lambda_1(\Lambda)}{\sqrt{n}}$  and so

$$\left( \frac{2\lambda_1(\Lambda)}{\sqrt{n}} \right)^n \leq \text{vol}(B(0, \lambda_1(\Lambda))).$$

- Thus,

$$\lambda_1(\Lambda) \leq \sqrt{n}(\det(\Lambda))^{1/n}.$$

# Minkowski's first theorem (3)

## Remarks 19

- ① Using the fact that  $\text{vol}(B(0, \lambda_1(\Lambda))) = \frac{\pi^{n/2}}{\Gamma(\frac{n}{2}+1)} (\lambda_1(\Lambda))^n$ , one can obtain an upper bound for  $\lambda_1(\Lambda)$  that is a lot tighter than  $\sqrt{n}(\det(\Lambda))^{1/n}$ .
- ②  $\lambda_1(\Lambda)$  can be very small compared to the Minkowski bound. Indeed, consider in dimension 2 the lattice given by  $(1, 0)^T$  and  $(0, N)^T$  where  $N \in \mathbb{N} \setminus \{0\}$ . Then, the Minkovsky bound is  $\sqrt{2}\sqrt{N}$  but  $\lambda_1(\Lambda) = 1$ .
- ③  $\lambda_1(\Lambda)$  can be very close to the Minkowski bound. Indeed, one can show, that in any dimension, there exists a lattice with shortest vector at least  $c\sqrt{n}(\det(\Lambda))^{1/n}$  for some constant  $c$ .
- ④ It has been shown that  $O(\sqrt{(n)}) \det(\Lambda)^{1/n}$  is the best upper bound one can possibly prove.
- ⑤ The term  $(\det(\Lambda))^{1/n}$  makes sure that the expressions scale properly. Indeed,  $\lambda_1(c\Lambda) = c\lambda_1(\Lambda)$  and  $(\det(c\Lambda))^{1/n} = c(\det(\Lambda))^{1/n}$ .

# Minkowski's second theorem (1)

## Theorem 20 (Minkowski - 2)

For any full-rank lattice  $\Lambda$  of rank  $n$ ,

$$\left( \prod_{i=1}^n \lambda_i(\Lambda) \right)^{1/n} \leq \sqrt{n} (\det(\Lambda))^{1/n}.$$

## Minkowski's second theorem (2)

Proof of theorem 20:

- Let  $x_1, \dots, x_n \in \Lambda$  be linearly independent vectors achieving the successive minima (i.e.  $\|x_i\| = \lambda_i(\Lambda)$ ).
- Let  $x_1^*, \dots, x_n^*$  be their Gram-Schmidt orthogonalization.
- Consider the open ellipsoid  $T$  with axes  $x_1^*, \dots, x_n^*$  and lengths  $\lambda_1(\Lambda), \dots, \lambda_n(\Lambda)$ ,

$$T = \left\{ y \in \mathbb{R}^n : \sum_{i=1}^n \left( \frac{\langle y, x_i^* \rangle}{\|x_i^*\| \lambda_i(\Lambda)} \right)^2 < 1 \right\}.$$

- Let  $y \in \Lambda$  and let

$$k = \max\{k \in \{1, \dots, n\} : \|y\| \geq \lambda_k(\Lambda)\}.$$

- Then,

$$y \in \text{span}(x_1^*, \dots, x_k^*) = \text{span}(x_1, \dots, x_k),$$

else  $x_1, \dots, x_k, y$  would be  $k+1$  linearly independent vectors of length less than  $\lambda_{k+1}(\Lambda)$ .

## Minkowski's second theorem (3)

- Thus,

$$\begin{aligned} \sum_{i=1}^n \left( \frac{\langle y, x_i^* \rangle}{\|x_i^*\| \lambda_i(\Lambda)} \right)^2 &= \sum_{i=1}^k \left( \frac{\langle y, x_i^* \rangle}{\|x_i^*\| \lambda_i(\Lambda)} \right)^2 \\ &\geq \frac{1}{(\lambda_k(\Lambda))^2} \sum_{i=1}^k \left( \frac{\langle y, x_i^* \rangle}{\|x_i^*\|} \right)^2 \\ &= \frac{\|y\|^2}{(\lambda_k(\Lambda))^2} \\ &\geq 1. \end{aligned}$$

- Hence,

$$y \notin T.$$

# Minkowski's second theorem (4)

- By Minkowski's convex body theorem,

$$\text{vol}(T) \geq 2^n \det(\Lambda).$$

- On the other hand, by the volume formula for ellipsoids,

$$\text{vol}(T) = \left( \prod_{i=1}^n \lambda_i(\Lambda) \right) \text{vol}(B(0, 1)) \geq \left( \prod_{i=1}^n \lambda_i(\Lambda) \right) \left( \frac{2}{\sqrt{n}} \right)^n.$$

- Combining both bounds yields

$$\left( \prod_{i=1}^n \lambda_i(\Lambda) \right)^{1/n} \leq \sqrt{n} (\det(\Lambda))^{1/n}.$$

# Minkowski's second theorem (5)

## Remarks 21

- ① Using the fact that  $\text{vol}(B(0, 1)) = \frac{\pi^{n/2}}{\Gamma(\frac{n}{2}+1)}$ , one can obtain a better upper bound for the geometric mean  $\left(\prod_{i=1}^n \lambda_i(\Lambda)\right)^{1/n}$ .
- ② The two previous results can easily be converted for any other norm.
- ③ The two previous results can be adapted to lattices of general rank.



Part I:  
Definitions

Part II:  
Comparing  
lattices

Part III:  
Gram-Schmidt  
Orthogonal-  
ization

Part IV:  
Determinant

Part V:  
Successive  
minima

Part VI:  
Minkowski's  
theorems

Part VII:  
Computational  
problems

References:

# Part VII: Computational problems

# Shortest vector problem

## Search SVP

Given a lattice basis  $B \in \mathbb{Z}^{m \times n}$ ,  
find  $v \in \mathcal{L}(B)$  such that  $0 \neq \|v\| = \lambda_1(\mathcal{L}(B))$ .

## Optimization SVP

Given a lattice basis  $B \in \mathbb{Z}^{m \times n}$ ,  
find  $\lambda_1(\mathcal{L}(B))$ .

## Decisional SVP

Given a lattice basis  $B \in \mathbb{Z}^{m \times n}$  and a rational  $r \in \mathbb{Q}$ ,  
determine whether  $\lambda_1(\mathcal{L}(B)) \leq r$  or not.

Surprisingly:

Search SVP  $\Leftrightarrow$  Optimization SVP  $\Leftrightarrow$  Decisional SVP

# Approximate shortest vector problem

Let  $\gamma \geq 1$ .

## Search $\text{SVP}_\gamma$

Given a lattice basis  $B \in \mathbb{Z}^{m \times n}$ ,  
find  $v \in \mathcal{L}(B)$  such that  $0 \neq \|v\| \leq \gamma \lambda_1(\mathcal{L}(B))$ .

## Optimization $\text{SVP}_\gamma$

Given a lattice basis  $B \in \mathbb{Z}^{m \times n}$ ,  
find  $d$  such that  $d \leq \lambda_1(\mathcal{L}(B)) \leq \gamma d$ .

## Promise $\text{SVP}_\gamma$ or $\text{GapSVP}_\gamma$

Given a lattice basis  $B \in \mathbb{Z}^{m \times n}$  and a rational  $r \in \mathbb{Q}$ ,  
determine whether  $(B, r)$  belongs to the YES instance ( $= \lambda_1(\mathcal{L}(B)) \leq r$ )  
or to the NO instance ( $\lambda_1(\mathcal{L}(B)) > \gamma r$ ).

Surprisingly:

$$\text{Search } \text{SVP}_\gamma \Rightarrow \text{Optimization } \text{SVP}_\gamma \Leftrightarrow \text{Promise } \text{SVP}_\gamma$$

# Closest vector problem

Let  $\gamma \geq 1$ .

## Search $\text{CVP}_\gamma$

Given a lattice basis  $B \in \mathbb{Z}^{m \times n}$  and a vector  $t \in \mathbb{Z}^m$ ,  
find  $v \in \mathcal{L}(B)$  such that  $\|v - t\| \leq \gamma \text{dist}(t, \mathcal{L}(B))$ .

## Optimization $\text{CVP}_\gamma$

Given a lattice basis  $B \in \mathbb{Z}^{m \times n}$  and a vector  $t \in \mathbb{Z}^m$ ,  
find  $d$  such that  $d \leq \text{dist}(t, \mathcal{L}(B)) \leq \gamma d$ .

## Promise $\text{CVP}_\gamma$ or Gap $\text{CVP}_\gamma$

Given a lattice basis  $B \in \mathbb{Z}^{m \times n}$ , a rational  $r \in \mathbb{Q}$  and a vector  $t \in \mathbb{Z}^m$ ,  
determine whether  $(B, r, t)$  belongs to the YES instance  
( $= \text{dist}(t, \mathcal{L}(B)) \leq r$ ) or to the NO instance ( $= \text{dist}(t, \mathcal{L}(B)) > \gamma r$ ).

# Miscellaneous lattice problems

## SIVP

Given a lattice basis  $B \in \mathbb{Z}^{m \times n}$ ,  
find  $n$  linearly independent vectors  $v_1, \dots, v_n \in \mathcal{L}(B)$  such that  
 $0 \neq \|v_i\| \leq \gamma \lambda_i(\mathcal{L}(B))$ .

## Bounded distance decoding

Given a lattice basis  $B \in \mathbb{Z}^{m \times n}$  and a vector  $t \in \mathbb{Z}^m$  such that  
 $\text{dist}(t, \mathcal{L}(B)) < \frac{\lambda_1(\mathcal{L}(B))}{n}$  for a given  $n \in \mathbb{N}$ ,  
find  $v \in \mathcal{L}(B)$  such that  $\|v - t\| < \frac{\lambda_1(\text{dist}(t, \mathcal{L}(B)))}{n}$ .

## Covering radius problem

Given a lattice basis  $B \in \mathbb{Z}^{m \times n}$ ,  
find the largest distance from any vector to the lattice.

# Remarks about lattice problems

## Remarks 21

- 1 Many lattice problems are conjectured to be hard!
- 2 Finding the shortest vector is hard however, finding a short vector is manageable using different algorithms.
- 3 Many cryptographic schemes based on lattice problems seem to be secure and are even conjectured to be quantum secure.

# References

## References:

- Oded Regev's course notes "Lattices in Computer Science" (from 2004) from the Tel Aviv University that are accessible via the link: [https://cims.nyu.edu/~regev/teaching/lattices\\_fall\\_2009/](https://cims.nyu.edu/~regev/teaching/lattices_fall_2009/)
- Daniele Micciancio's course notes "Lattices Algorithms and Applications" (from 2010) from the University of California San Diego that are accessible via the link: <http://cseweb.ucsd.edu/classes/wi10/cse206a/>
- Chi's, Choi's, Kim's and Kim's lecture notes "Lattice Based Cryptography for Beginners" accessible via the link: <https://eprint.iacr.org/2015/938.pdf>
- Steven D. Galbraith's book "Mathematics of public key cryptography".