

Lukas Helminger

Mathematical Foundations of Cryptography – WT 2019/20

SCIENCE PASSION TECHNOLOGY



## Outline

#### Vector Spaces

#### **Definition and Properties**

- Fundamental Domain
- Volume

#### Short Vectors in Lattices

- Computational Problems
- Minkowski's and Hermite's Theorem

#### Lattice Reduction Algorithms

Babai's Algorithm

## Literature

The slides are based on the following sources

- An Introduction to Mathematical Cryptography, Hoffstein, Jeffrey, Pipher, Jill, Silverman, J.H.
- A Decade of Lattice Cryptography, Chris Peikert
- The LLL Algorithm, Phong Q. Nguyen, Brigitte Vallée (Eds.)

Many graphics are based on graphics from Maria Eichlseder.

## Lattice-Based Cryptography

- Conjectured security against quantum attacks:
  One half of the 2nd round candidates for NIST Post-Quantum Cryptography Standardization are lattice-based (in the category PKE).
- Algorithmic simplicity, efficiency, and parallelism.
- Strong security guarantees from worst-case hardness.
- Construction of versatile and powerful cryptographic objects
  - Fully Homomorphic Encryption
  - Attribute-Based Encryption

## **Vector Spaces**

### **Vector Spaces**

- A vector space V is a subset of ℝ<sup>m</sup> that is closed under addition and under scalar multiplication by elements of ℝ.
- A linear combination of the vectors  $v_1, \ldots, v_k$  is any vector of the form

 $w = \alpha_1 v_1 + \dots + \alpha_k v_k$ , with  $\alpha_1, \dots, \alpha_k \in \mathbb{R}$ .

The collection of all such linear combinations is called the span of  $\{v_1, \ldots, v_k\}$ .

• A set of vectors  $v_1, \ldots, v_k \in V$  is linearly dependent

$$\alpha_1 \mathbf{v}_1 + \dots + \alpha_k \mathbf{v}_k = \mathbf{0} \Rightarrow \alpha_1 = \dots = \alpha_k = \mathbf{0}.$$

• A basis for V is a set of linearly independent vectors  $v_1, \ldots, v_k$  that span V.

## Length and Angle

• The dot product of  $v = (x_1, ..., x_m), w = (y_1, ..., y_m) \in V$  is the quantity

$$v \cdot w = x_1 y_1 + \dots + x_m y_m.$$

- v and w are orthogonal if  $v \cdot w = 0$ .
- The length, or Euclidean norm, of *v* is the quantity

$$\|\mathbf{v}\| = \sqrt{x_1^2 + \cdots + x_m^2}.$$

• A basis  $v_1, \ldots, v_n$  is an orthogonal basis if

$$v_i \cdot v_j = 0 \quad \forall i \neq j.$$

Let α be the angle between v and w, then

$$\mathbf{v} \cdot \mathbf{w} = \|\mathbf{v}\| \|\mathbf{w}\| \cos(\alpha).$$

## Gram Matrix

Let  $v_1, \ldots, v_n$  be vectors in  $\mathbb{R}^m$ . The entries of the Gram matrix are given by  $G_{ij} = v_i \cdot v_j$ . The determinant of *G* is called the Gram determinant.

- det  $G \neq 0 \Rightarrow v_1, \ldots, v_n$  linearly independent.
- $\sqrt{\det G}$  is the *n*-dimensional volume spanned by  $v_1, \ldots, v_n$ .

**Example:** Let  $v_1 = (2,3), v_2 = (1,4)$ .

$$G = \begin{pmatrix} 2 & 3 \\ 1 & 4 \end{pmatrix} \cdot \begin{pmatrix} 2 & 1 \\ 3 & 4 \end{pmatrix} = \begin{pmatrix} 13 & 14 \\ 14 & 17 \end{pmatrix}$$

 $\operatorname{vol}(v_1, v_2) = \sqrt{\det G} = \sqrt{25} = 5$ 

## Gram-Schmidt Algorithm

#### Theorem (Gram-Schmidt Algorithm)

Let  $v_1, \ldots, v_n$  be a basis for a vector space  $V \subset \mathbb{R}^m$ . The following algorithm creates an orthogonal basis  $v_1^*, \ldots, v_n^*$  for V:

$$v_1^* \leftarrow v_1$$
  
for  $i = 2..n$  do  
for  $j = 1..i - 1$   
$$\mu_{i,j} \leftarrow \frac{v_i \cdot v_j^*}{\|v_j^*\|^2}$$
$$v_i^* = v_i - \sum_{j=1}^{i-1} \mu_{i,j} v_j^*$$

# **Definition and Properties**

## Lattices

#### Definition (Lattice)

An *n*-dimensional lattice *L* is any subset of  $\mathbb{R}^n$  that is both:

- an additive subgroup
- discrete

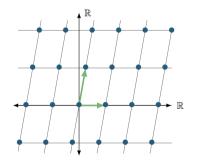
A basis for *L* is any set of independent vectors that generates *L*.

#### Lattice: Example

In other words, let  $v_1, \ldots, v_n \in \mathbb{R}^n$  be a set of linearly independent vectors. The lattice generated by  $v_1, \ldots, v_n$  is the set of linear combinations of  $v_1, \ldots, v_n$  with coefficients in  $\mathbb{Z}$ ,

$$L = \{a_1v_1 + \cdots + a_nv_n : a_1, \ldots, a_n \in \mathbb{Z}\}.$$

Example:



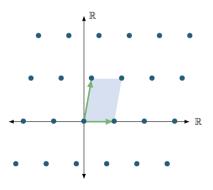
$$\boldsymbol{v}_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \, \boldsymbol{v}_2 = \begin{pmatrix} 1/4 \\ \sqrt{2} \end{pmatrix}$$

#### **Fundamental Domains**

#### Definition (Fundamental Domain)

Let *L* be a lattice of dimension *n* and let  $v_1, \ldots, v_n$  be a basis for *L*. The fundamental domain is the set

$$F = [0, 1)v_1 + \dots + [0, 1)v_n.$$



#### Volumes

#### Definition (Volume)

Let *L* be a lattice of dimension *n* and let *F* be a fundamental domain of *L*. Then the *n*-dimensional volume of *F* is called the volume of *L* (or sometimes the determinant of *L*).

**Example:** Let *L* be generated by the vectors

$$v_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, v_2 = \begin{pmatrix} 1/4 \\ \sqrt{2} \end{pmatrix}.$$

First, compute Gram matrix:

$$G = \begin{pmatrix} 1 & 0 \\ \frac{1}{4} & \sqrt{2} \end{pmatrix} \cdot \begin{pmatrix} 1 & \frac{1}{4} \\ 0 & \sqrt{2} \end{pmatrix} = \begin{pmatrix} 1 & \frac{1}{4} \\ \frac{1}{4} & \frac{33}{16} \end{pmatrix}$$

Therefore,

$$\operatorname{vol}(L) = \sqrt{\det G} = \sqrt{2}$$

## Same Lattice?

 $\boldsymbol{v}_1 = \begin{pmatrix} 3 \\ 0 \end{pmatrix}, \boldsymbol{v}_2 = \begin{pmatrix} 2 \\ 2 \end{pmatrix}$  $\boldsymbol{v}_1' = \begin{pmatrix} 8 \\ 2 \end{pmatrix}, \boldsymbol{v}_2' = \begin{pmatrix} 5 \\ 2 \end{pmatrix}$ 

#### Volume: Task

**Task:** Compute the volumes V resp. V' of the fundamental domains corresponding to  $v_1, v_2$  respectively  $v'_1, v'_2$ .

$$G = \begin{pmatrix} 3 & 0 \\ 2 & 2 \end{pmatrix} \begin{pmatrix} 3 & 2 \\ 0 & 2 \end{pmatrix} = \begin{pmatrix} 9 & 6 \\ 6 & 8 \end{pmatrix}.$$
$$G' = \begin{pmatrix} 8 & 2 \\ 5 & 2 \end{pmatrix} \begin{pmatrix} 8 & 5 \\ 2 & 2 \end{pmatrix} = \begin{pmatrix} 68 & 44 \\ 44 & 29 \end{pmatrix}.$$
Therefore  $V = \sqrt{G} = \sqrt{36} = 6 = \sqrt{36} = \sqrt{G'} = V'.$ 

#### Proposition

т

Every fundamental domain for a given lattice *L* has the same volume.

## Short Vectors in Lattices

## **Computational Problems**

 $\lambda_1(L)$ ... length of shortest nonzero vector in *L*.

- Shortest Vector Problem (SVP): Find a shortest nonzero vector v in L, i.e.  $||v|| = \lambda_1(L)$ .
- **Closest Vector Problem (CVP):** Given a vector *w*, find closest vector to *w* in *L*.

**Example:** Given the lattice generated by  $v_1, v_2$ 

$$v_1 = \begin{pmatrix} 8 \\ 2 \end{pmatrix}, v_2 = \begin{pmatrix} 5 \\ 2 \end{pmatrix}$$

and given the vector  $w = (-1, 3)^T$ . What is a shortest nonzero vector of L? Which vector is closest to w?

$$\begin{pmatrix} -1 \\ 2 \end{pmatrix}$$
 and  $\begin{pmatrix} -1 \\ 2 \end{pmatrix}$ 

## How long is the shortest vector?

#### Theorem (Minkowski's Theorem)

Let  $L \subset \mathbb{R}^n$  be a lattice of dimension n. Let  $S \subset \mathbb{R}^n$  be convex, closed and symmetric. Suppose that  $vol(S) \ge 2^n vol(L)$ , then

 $S \cap L \supseteq \{0\}.$ 

S... hypercube in  $\mathbb{R}^n$  centered at 0 with length  $2 \operatorname{vol}(L)^{1/n}$ , then  $\operatorname{vol}(S) = 2^n \operatorname{vol}(L)$ . Applying Minkowski's theorem leads to:

Corollary (Hermite's Theorem)

Every lattice L of dimension n contains a nonzero  $v \in L$  satisfying

 $\|v\| \leq \sqrt{n} \operatorname{vol}(L)^{\frac{1}{n}}.$ 

# Lattice Reduction Algorithms

## Babai's Closest Vertex Algorithm

**Input:** Basis  $v_1, \ldots, v_n$  and  $w \in \mathbb{R}^n$ .

- 1. Write  $w = t_1v_1 + \cdots, t_nv_n$ , with  $t_1, \ldots, t_n \in \mathbb{R}$ .
- 2. Set  $a_i = \lfloor t_i \rfloor$  for i = 1, ..., n.
- 3. Return  $v = a_1v_1 + \dots + a_nv_n$ .

Try out the algorithm for

$$v_1 = \begin{pmatrix} 3 \\ 0 \end{pmatrix}, v_2 = \begin{pmatrix} 2 \\ 2 \end{pmatrix}, w = \begin{pmatrix} -1 \\ 3 \end{pmatrix}.$$

### **Orthogonality Defects**

#### Definition (Hadamard Ratio)

We define the Hadamard ratio of the basis  $B = \{v_1, \ldots, v_n\}$ . to be the quantity

$$H(B) = \left(\frac{\operatorname{vol}(L)}{\|v_1\| \cdots \|v_n\|}\right)^{\frac{1}{n}} \in (0, 1].$$

(the closer to 1, the more orthogonal)

Example: 
$$v_1 = \begin{pmatrix} 3 \\ 0 \end{pmatrix}, v_2 = \begin{pmatrix} 2 \\ 2 \end{pmatrix}$$
  
$$H(B) = \left(\frac{6}{\sqrt{9\sqrt{8}}}\right)^{\frac{1}{2}} \approx 0.84.$$