

# Lattice Coding I: From Theory To Application

Amin Sakzad

Dept of Electrical and Computer Systems Engineering

Monash University

amin.sakzad@monash.edu

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  - Sphere Packing Problem
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# Motivation I: Geometry of Numbers

Initiated by Minkowski and studies convex bodies and integer points in  $\mathbb{R}^n$ .

- ① Diophantine Approximation,
- ② Functional Analysis

**Examples** Approximating real numbers by rationals, sphere packing problem, covering problem, factorizing polynomials, etc.

## Motivation II: Telecommunication

- ① Channel Coding Problem,
- ② Quantization Problem

**Examples** Signal constellations, space-time coding, lattice-reduction-aided decoders, relaying protocols, etc.



## Definition

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### Definition

An infinite discrete set  $\Lambda \subseteq \mathbb{R}^n$  is called a *lattice* if  $\Lambda$  is a group under addition in  $\mathbb{R}^n$ .

Every lattice is generated by the integer combination of some linearly independent vectors  $\mathbf{g}_1, \dots, \mathbf{g}_m \in \mathbb{R}^n$ , i.e.,

$$\Lambda = \{u_1\mathbf{g}_1 + \dots + u_m\mathbf{g}_m : u_1, \dots, u_m \in \mathbb{Z}\}.$$

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### Definition

The  $m \times n$  matrix  $\mathbf{G} = (\mathbf{g}_1, \dots, \mathbf{g}_m)$  which has the generator vectors as its rows is called a **generator matrix** of  $\Lambda$ . A lattice is called **full rank** if  $m = n$ .



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Note that

$$\Lambda = \{\mathbf{x} = \mathbf{uG} : \mathbf{u} \in \mathbb{Z}^n\}.$$

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The *minimum distance* of  $\Lambda$  is defined by

$$d_{\min}(\Lambda) = \min\{\|\mathbf{x}\| : \mathbf{x} \in \Lambda \setminus \{\mathbf{0}\}\},$$

where  $\|\cdot\|$  stands for Euclidean norm.

## Definition

The *determinate* (*volume*) of an  $n$ -dimensional lattice  $\Lambda$ ,  $\det(\Lambda)$ , is defined as

$$\det[\mathbf{G}\mathbf{G}^T]^{\frac{1}{2}}.$$

## Definition

The *coding gain* of a lattice  $\Lambda$  is defined as:

$$\gamma(\Lambda) = \frac{d_{\min}^2(\Lambda)}{\det(\Lambda)^{\frac{2}{n}}}.$$

Geometrically,  $\gamma(\Lambda)$  measures the increase in the density of  $\Lambda$  over the lattice  $\mathbb{Z}^n$ .

## Definition

The set of all vectors in  $\mathbb{R}^n$  whose inner product with all elements of  $\Lambda$  is an integer form the *dual lattice*  $\Lambda^*$ .

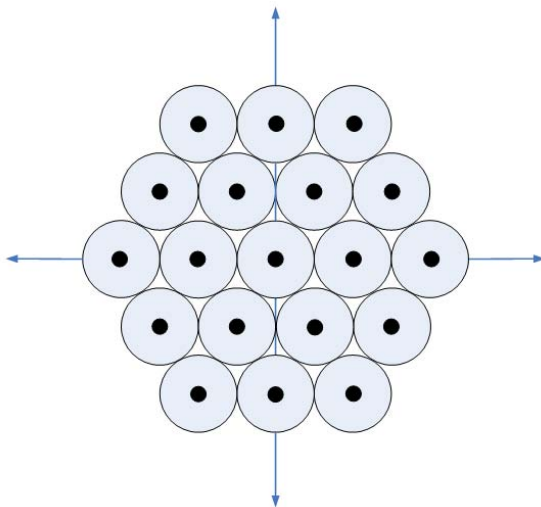
## Definition

The set of all vectors in  $\mathbb{R}^n$  whose inner product with all elements of  $\Lambda$  is an integer form the *dual lattice*  $\Lambda^*$ .

For a lattice  $\Lambda$ , with generator matrix  $\mathbf{G}$ , the matrix  $\mathbf{G}^{-T}$  forms a basis matrix for  $\Lambda^*$ .



## Three examples





# Barens-Wall Lattices

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$$\mathbf{G} = \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix}.$$

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- A basis matrix for Barnes-Wall lattice  $\mathcal{BW}_n$ ,  $n = 2^m$ , can be formed by selecting the rows of matrices  $\mathbf{G}^{\otimes m}, \dots, 2^{\lfloor \frac{m}{2} \rfloor} \mathbf{G}^{\otimes m}$  which have a square norm equal to  $2^{m-1}$  or  $2^m$ .

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- $d_{\min}(\mathcal{BW}_n) = \sqrt{\frac{n}{2}}$  and  $\det(\mathcal{BW}_n) = (\frac{n}{2})^{\frac{n}{4}}$ , which confirms that  $\gamma(\mathcal{BW}_n) = \sqrt{\frac{n}{2}}$ .

$\mathcal{D}_n$  Lattices

- For  $n \geq 3$ ,  $\mathcal{D}_n$  can be represented by the following basis matrix:

$$\mathbf{G} = \begin{pmatrix} -1 & -1 & 0 & \cdots & 0 \\ 1 & -1 & 0 & \cdots & 0 \\ 0 & 1 & -1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & -1 \end{pmatrix}.$$

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- We have  $\det(\mathcal{D}_n) = 2$  and  $d_{\min}(\mathcal{D}_n) = \sqrt{2}$ , which result in  $\gamma(\mathcal{D}_n) = 2^{\frac{n-2}{n}}$ .

- Sphere Packing Problem,
- Covering Problem,
- Quantization,
- Channel Coding Problem.



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## Sphere Packing Problem

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## Definition

The *density* of  $\Lambda$  is defined as

$$\Delta(\Lambda) = \frac{\rho^n V_n}{\det(\Lambda)},$$

where  $V_n$  is the volume of an  $n$ -dimensional sphere with radius 1.

Note that

$$V_n = \frac{\pi^{n/2}}{(n/2)!}.$$

## Definition

The *kissing number*  $\tau(\Lambda)$  is the number of spheres that touches one sphere.

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## Definition

The *Hermite's constant*  $\gamma_n$  is the highest attainable coding gain of an  $n$ -dimensional lattice.

# Lattice Sphere Packing Problem

Find the densest lattice packing of equal nonoverlapping, solid spheres (or balls) in  $n$ -dimensional space.

# Summary of Well-Known Results

## Theorem

- *For large  $n$ 's we have*

$$\frac{1}{2\pi e} \leq \frac{\gamma_n}{n} \leq \frac{1.744}{2\pi e},$$

# Summary of Well-Known Results

## Theorem

- *For large  $n$ 's we have*

$$\frac{1}{2\pi e} \leq \frac{\gamma_n}{n} \leq \frac{1.744}{2\pi e},$$

- *The densest lattice packings are known for dimensions 1 to 8 and 12, 16, and 24.*



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### Definition

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### Definition

The *thickness* of  $\Lambda$  is defined as

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### Definition

The *normalized thickness* of  $\Lambda$  is then  $\theta(\Lambda) = \frac{\Theta}{V_n}$ .

# Lattice Covering Problem

Ask for the thinnest lattice covering of equal overlapping, solid spheres (or balls) in  $n$ -dimensional space.

# Summary of Well-Known Results

## Theorem

- *The thinnest lattice coverings are known for dimensions 1 to 5, (all  $\mathcal{A}_n^*$ ).*
- *Davenport's Construction of thin lattice coverings, (thinner than  $\mathcal{A}_n^*$  for  $n \leq 200$ ).*

## Definition

For any point  $\mathbf{x}$  in a constellation  $\mathcal{A}$  the *Voronoi cell*  $\nu(\mathbf{x})$  is defined by the set of points that are at least as close to  $\mathbf{x}$  as to any other point  $\mathbf{y} \in \mathcal{A}$ , i.e.,

$$\nu(\mathbf{x}) = \{\mathbf{v} \in \mathbb{R}^n : \|\mathbf{v} - \mathbf{x}\| \leq \|\mathbf{v} - \mathbf{y}\|, \forall \mathbf{y} \in \mathcal{A}\}.$$

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We simply denote  $\nu(\mathbf{0})$  by  $\nu$ .

## Definition

An  $n$ -dimensional *quantizer* is a set of points chosen in  $\mathbb{R}^n$ . The input  $\mathbf{x}$  is an arbitrary point of  $\mathbb{R}^n$ ; the output is the closest point to  $\mathbf{x}$ .

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A good quantizer attempts to minimize the *mean squared error* of quantization.



# Lattice Quantizer Problem

finds and  $n$ -dimensional lattice  $\Lambda$  for which

$$G(\nu) = \frac{\frac{1}{n} \int_{\nu} \mathbf{x} \cdot \mathbf{x} d\mathbf{x}}{\det(\nu)^{1+\frac{2}{n}}},$$

is a minimum.

# Summary of Well-Known Results

## Theorem

- *The optimum lattice quantizers are only known for dimensions 1 to 3.*
- *As  $n \rightarrow \infty$ , we have*

$$G_n \rightarrow \frac{1}{2\pi e}.$$

# Summary of Well-Known Results

## Theorem

- *The optimum lattice quantizers are only known for dimensions 1 to 3.*
- *As  $n \rightarrow \infty$ , we have*

$$G_n \rightarrow \frac{1}{2\pi e}.$$

It is worth remarking that the best  $n$ -dimensional quantizers presently known are always the duals of the best packings known.

## Definition

For two points  $\mathbf{x}$  and  $\mathbf{y}$  in  $\mathbb{F}_q^n$  the *Hamming distance* is defined as

$$d(\mathbf{x}, \mathbf{y}) = \|\{i: \mathbf{x}_i \neq \mathbf{y}_i\}\|.$$

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## Definition

A  $q$ -ary  $(n, M, d_{\min})$  code  $\mathcal{C}$  is a subset of  $M$  points in  $\mathbb{F}_q^n$ , with minimum distance

$$d_{\min}(\mathcal{C}) = \min_{\mathbf{x} \neq \mathbf{y} \in \mathcal{C}} d(\mathbf{x}, \mathbf{y}).$$

# Performance Measures I

- Suppose that  $\mathbf{x}$ , which is in a constellation  $\mathcal{A}$ , is sent,
- $\mathbf{y} = \mathbf{x} + \mathbf{z}$  is received, where the components of  $\mathbf{z}$  are i.i.d. based on  $\mathcal{N}(0, \sigma^2)$ ,
- The **probability of error** is defined as

$$P_e(\mathcal{A}, \sigma^2) = 1 - \frac{1}{(\sqrt{2\pi}\sigma)^n} \int_{\nu} \exp\left(\frac{-\|\mathbf{x}\|^2}{2\sigma^2}\right) d\mathbf{x}.$$

# Performance Measures II

## Rate

### Definition

The *rate*  $\mathfrak{r}$  of an  $(n, M, d_{\min})$  code  $\mathcal{C}$  is

$$\mathfrak{r} = \frac{\log_2(M)}{n}.$$

# Performance Measures II

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The power of a transmission has a close relation with the rate of the code.



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The power of a transmission has a close relation with the rate of the code.

## Normalized Logarithmic Density

### Definition

The *normalized logarithmic density (NLD)* of an  $n$ -dimensional lattice  $\Lambda$  is

$$\frac{1}{n} \log \left( \frac{1}{\det(\Lambda)} \right).$$

# Performance Measures III

## Capacity

### Definition

The *capacity* of an AWGN channel with noise variance  $\sigma^2$  is

$$C = \frac{1}{2} \log \left( 1 + \frac{P}{\sigma^2} \right),$$

where  $\frac{P}{\sigma^2}$  is called the *signal-to-noise ratio*.

# Performance Measures III

## Capacity

### Definition

The *capacity* of an AWGN channel with noise variance  $\sigma^2$  is

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where  $\frac{P}{\sigma^2}$  is called the *signal-to-noise ratio*.

## Generalized Capacity

### Definition

The *capacity* of an “unconstrained” AWGN channel with noise variance  $\sigma^2$  is

$$C_\infty = \frac{1}{2} \ln \left( \frac{1}{2\pi e \sigma^2} \right).$$

# Approaching Capacity

## Capacity-Achieving Codes

### Definition

A  $(n, M, d_{\min})$  code  $\mathcal{C}$  is called *capacity-achieving* for the AWGN channel with noise variance  $\sigma^2$ , if  $\tau = C$  when  $P_e(\mathcal{C}, \sigma^2) \approx 0$ .

## Sphere-Bound-Achieving Lattices

### Definition

An  $n$ -dimensional lattice  $\Lambda$  is called *capacity-achieving* for the unconstrained AWGN channel with noise variance  $\sigma^2$ , if  $NLD(\Lambda) = C_\infty$  when  $P_e(\Lambda, \sigma^2) \approx 0$ .

## Definition

The *volume-to-noise ratio* of a lattice  $\Lambda$  over an unconstrained AWGN channel with noise variance  $\sigma^2$  is defined as

$$\alpha^2(\Lambda, \sigma^2) = \frac{\det(\Lambda)^{\frac{2}{n}}}{2\pi e \sigma^2}.$$

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The *volume-to-noise ratio* of a lattice  $\Lambda$  over an unconstrained AWGN channel with noise variance  $\sigma^2$  is defined as

$$\alpha^2(\Lambda, \sigma^2) = \frac{\det(\Lambda)^{\frac{2}{n}}}{2\pi e \sigma^2}.$$

Note that  $\alpha^2(\Lambda, \sigma^2) = 1$  is equivalent to  $\text{NLD}(\Lambda) = C_\infty$ .

# Union Bound Estimate

Using the formula of coding gain and  $\alpha^2(\Lambda, \sigma^2)$ , we obtain an estimate upper bound for the probability of error for a maximum-likelihood decoder:

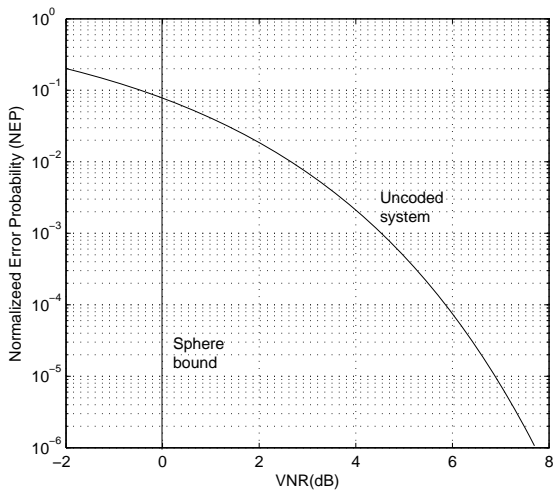
$$P_e(\Lambda, \sigma^2) \leq \frac{\tau(\Lambda)}{2} \operatorname{erfc} \left( \sqrt{\frac{\pi e}{4} \gamma(\Lambda) \alpha^2(\Lambda, \sigma^2)} \right),$$

where

$$\operatorname{erfc}(t) = \frac{2}{\sqrt{\pi}} \int_t^{\infty} \exp(-t^2) dt.$$



## Probability of Error versus VNR







Thanks for your attention! Friday 18 Oct. Building 72, Room 132.

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