

# Lattices and Their Applications to Wireless Communications

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# Sabbatical FY 2017

## **Universities visited:**

- June 2017 Technion (Israel Institute of Technology) – Eitan Yaakobi
- October 2017 Texas A&M University (USA) – Krishna Narayanan
- March 2018 Monash University (Melbourne, Australia) – Emanuele Viterbo
- March–May, July–Sept, Nov–Dec 2017 Oregon State University (USA) – Bella Bose

## **Impressions of my sabbatical:**

Excellent time to “clear the mind.” Stimulate research progress by finding new ideas.

Giving presentation at university is good way to meet people.

Visiting many universities was a chance to get many new ideas.

But, not enough time to complete a paper etc.

Continued to advise 3 PhD students and 2 masters students remotely.

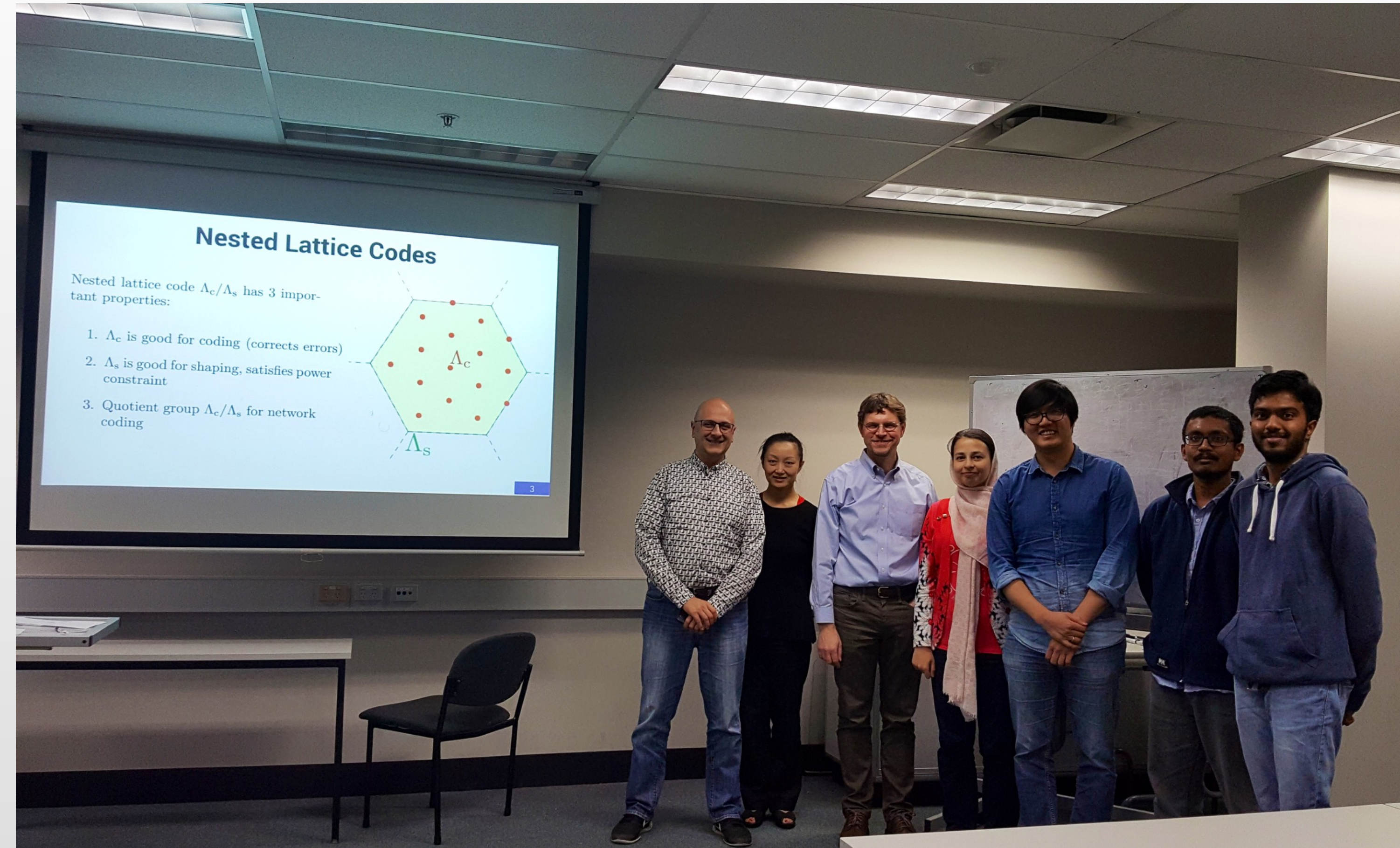


# “Introduction to Lattice Coding Theory”

At the universities I visited,  
I gave multiple lectures on  
“Lattice Coding Theory”

The target audience is first-  
year graduate students  
who have already studied  
coding theory.

Today’s presentation  
summarizes those  
lectures.



B. M. Kurkoski, “Introduction to lattice coding theory.” ECSE Seminar, Monash University, March 2018. 15, 22 and 29 March 2018.



# Contributors



Mohammad Nur  
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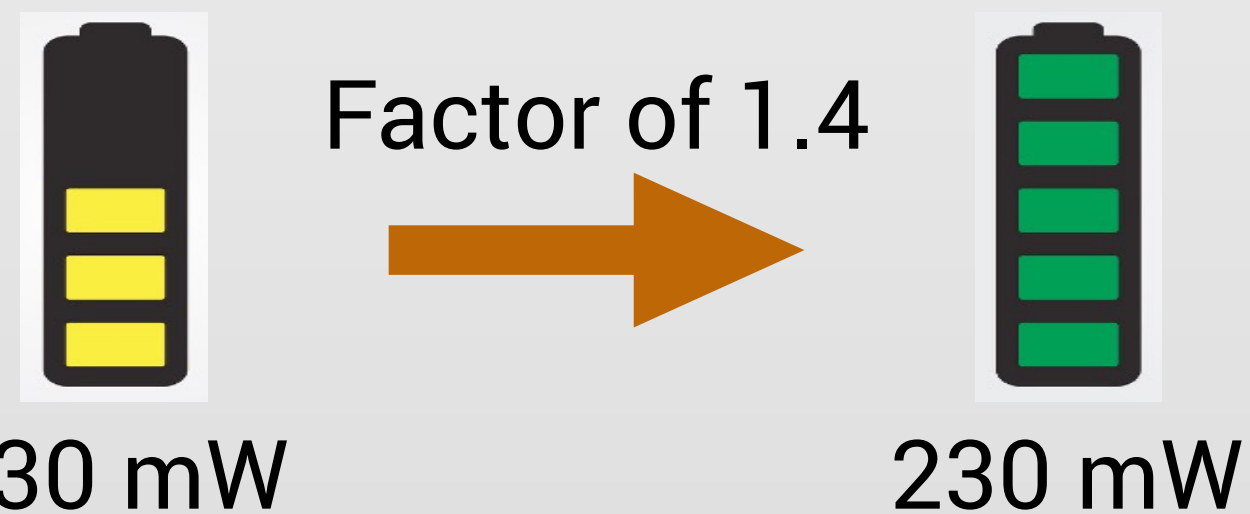


# Lattices and Their Applications to Wireless Communications

**Central question: How might lattices effectively be used in wireless communication systems?**

1. Lattice shaping is a practical way to gain 1.53 dB in SNR
2. Lattice-based physical layer network coding brings benefits of network coding to wireless communications

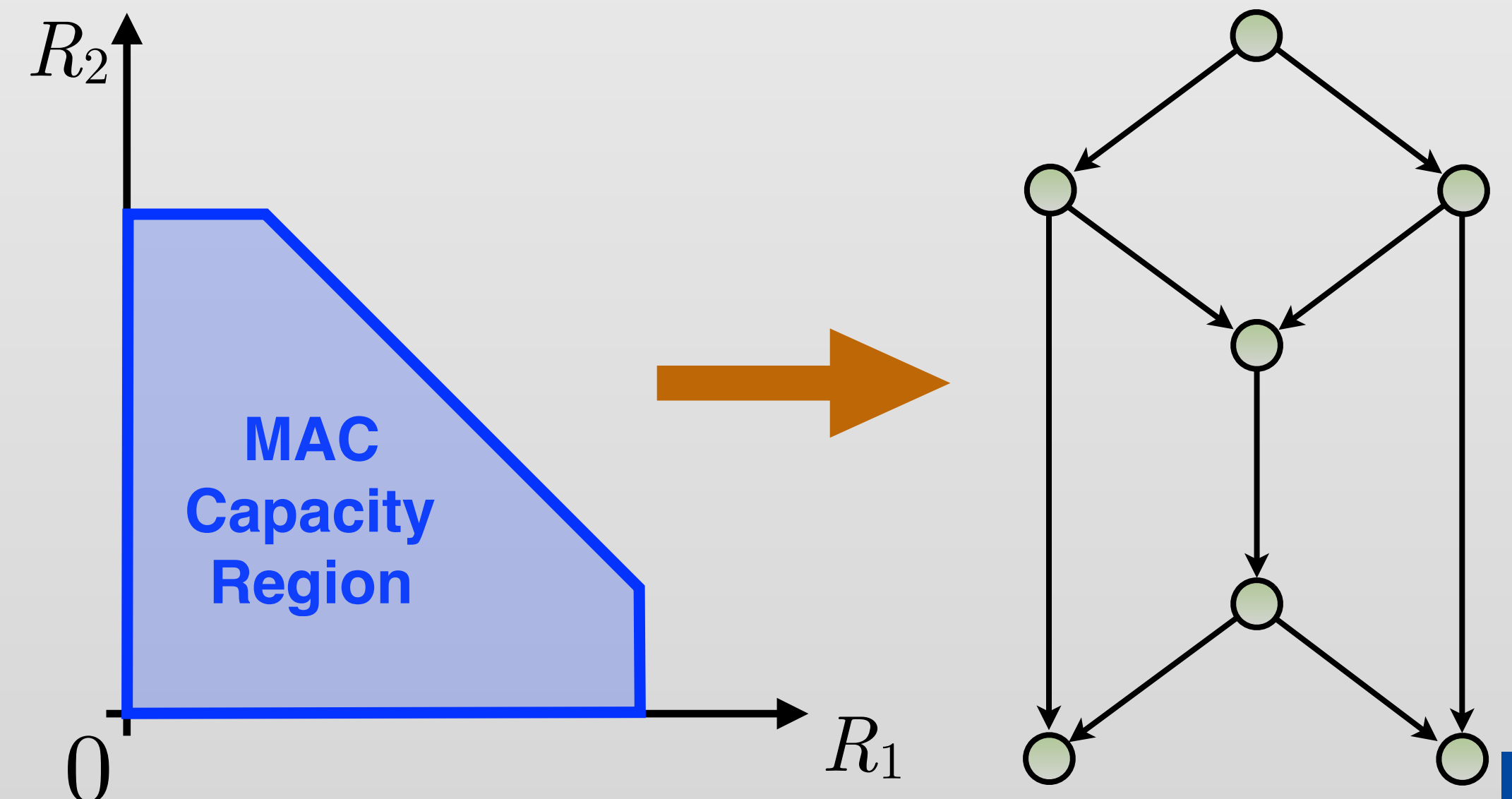
**How much is 1.53 dB?**



Significant reduction in transmit power:

- Smartphone battery lasts longer, efficient base stations
- Typical smartphone battery is 10000 mW-hour

**From MAC to Wireless Networks**





# Outline of Semi-Tutorial

## 1. Introduction to Lattices

- Tutorial and background on lattices

## 2. Lattices from Construction D and D'

- Form lattices from binary codes
- Since binary codes are well understood, promising candidate for practical lattices
- Lattices based on quasi-cyclic LDPC codes

## 3. Nested Lattices Codes for the AWGN Channel

- Classify nested lattice codes. Lattices with inflated lattice decoding achieve capacity
- Convolutional code lattices with good shaping gain

## 4. Physical Layer Network Coding

- Compute-Forward: Network coding when wireless signals add over the air
- Two channels: Bidirectional relay channel and the multiple access relay channel (MARC)



# Lattice Definition

**Definition 1** An  $n$ -dimensional *lattice*  $\Lambda$  is a discrete additive subgroup of  $\mathbb{R}^n$ .

**Intuition** A lattice is an error-correcting code defined on the real numbers (rather than a finite field)



# Lattice Definition

**Definition 1** An  $n$ -dimensional *lattice*  $\Lambda$  is a discrete additive subgroup of  $\mathbb{R}^n$ .

Vector addition in  $\mathbb{R}^n$ :

$$\mathbf{x} = [x_1 \quad , \dots , x_n \quad ]$$

$$\mathbf{y} = [y_1 \quad , \dots , y_n \quad ]$$

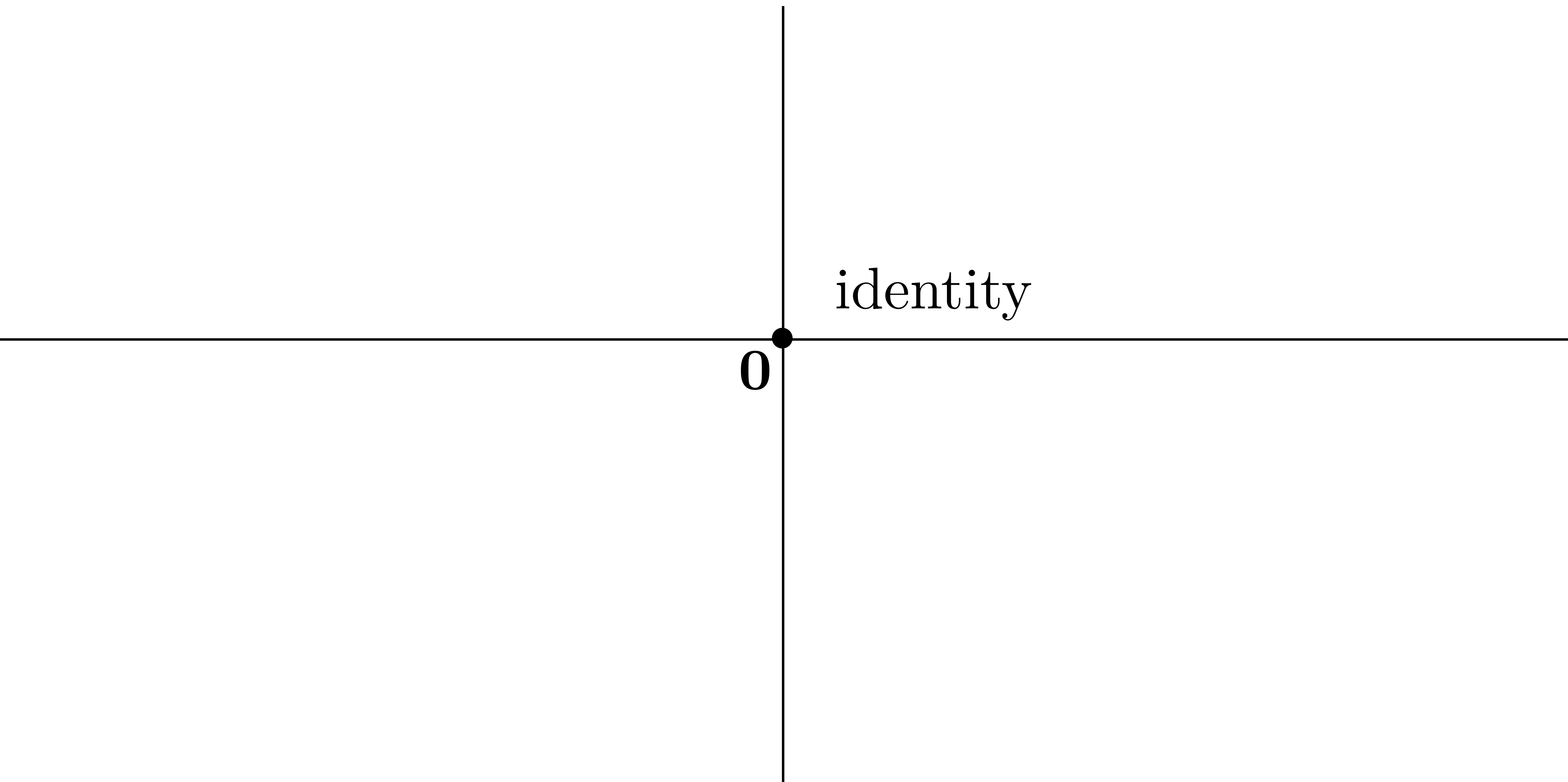
$$\mathbf{x} + \mathbf{y} = [x_1 + y_1, \dots, x_n + y_n]$$

Group properties:

- has identity
- has inverse
- associative
- closure
- (commutative)

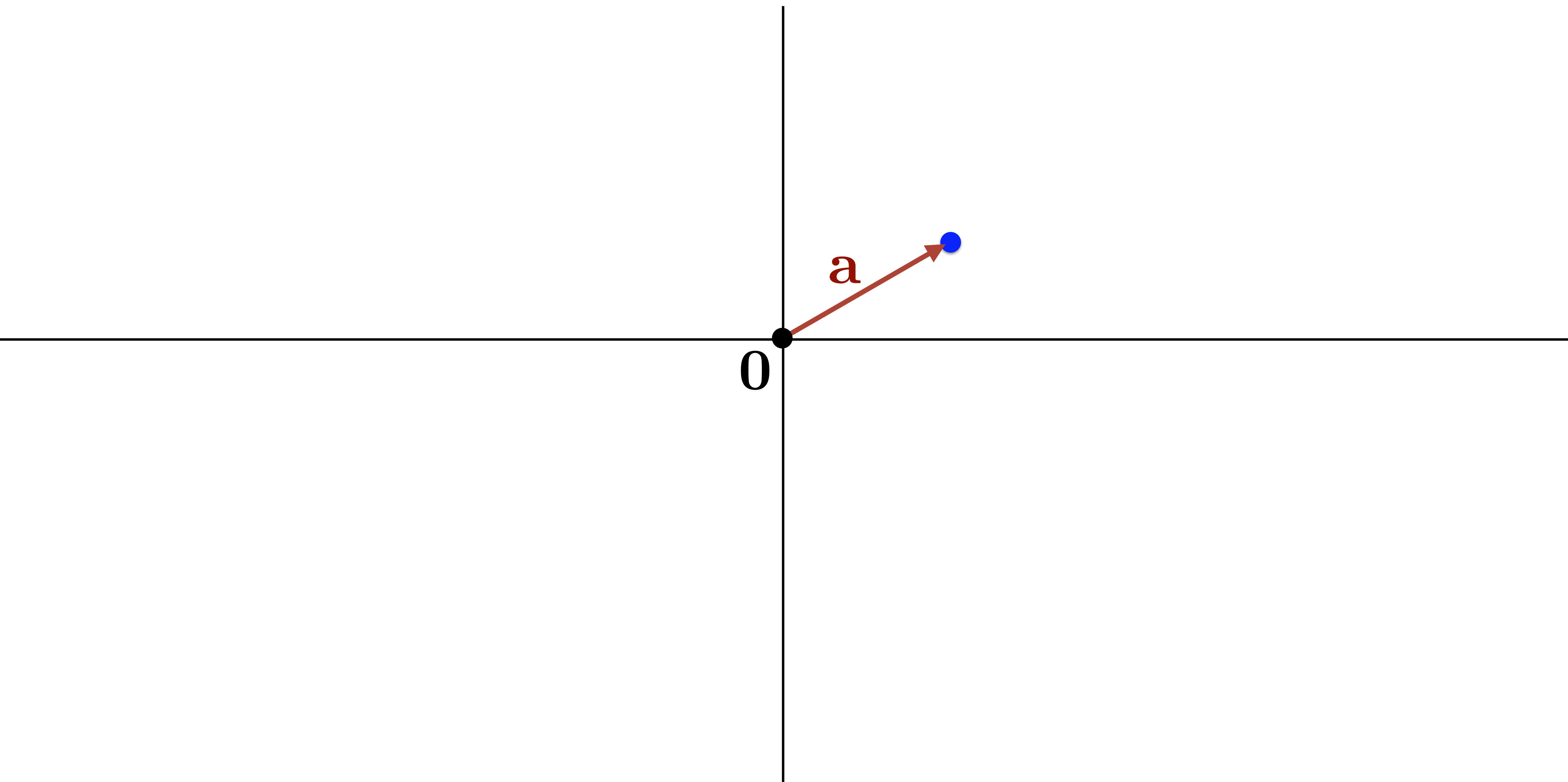


# Lattices in $\mathbb{R}^2$



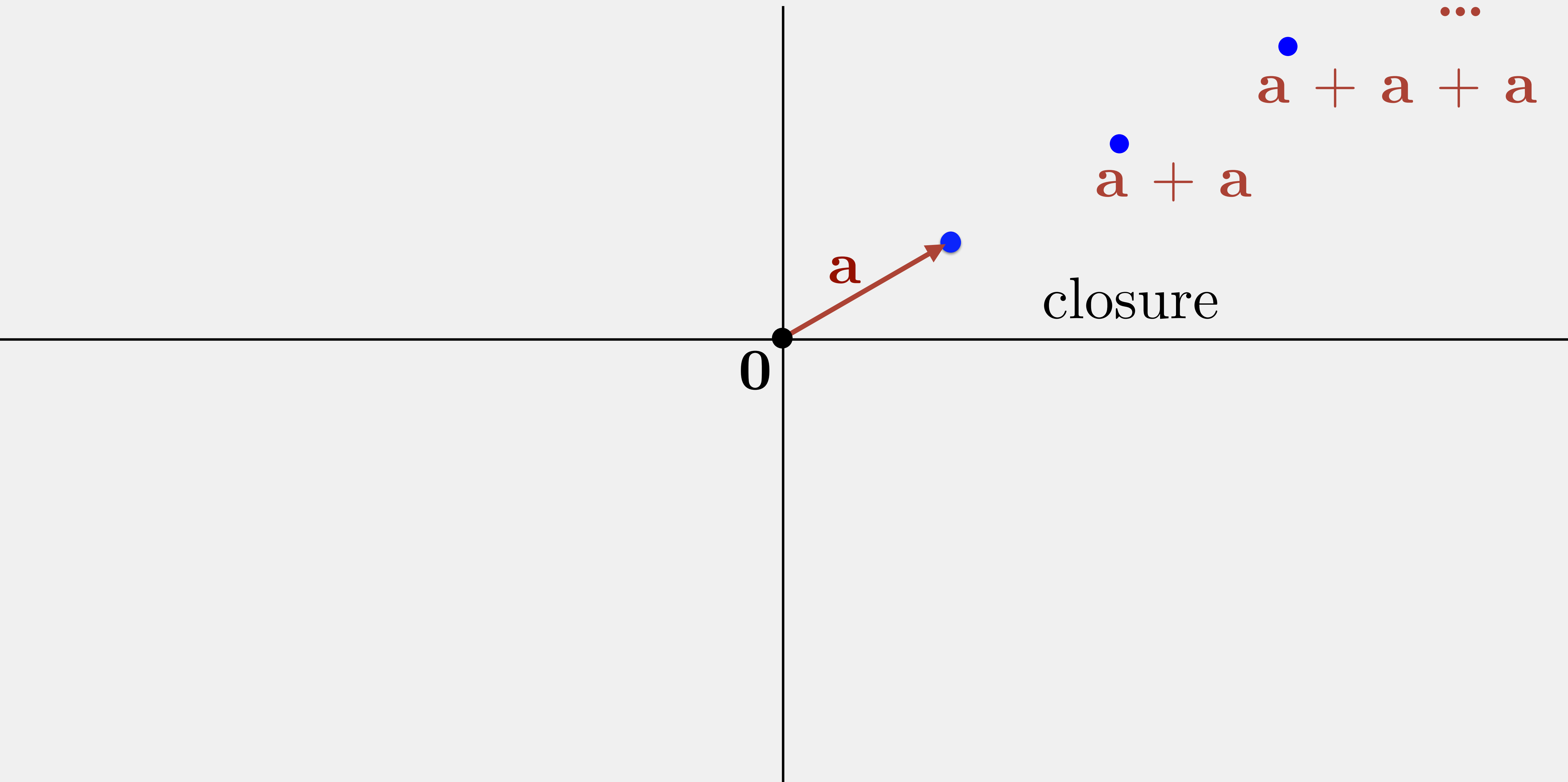


# Lattices in $\mathbb{R}^2$



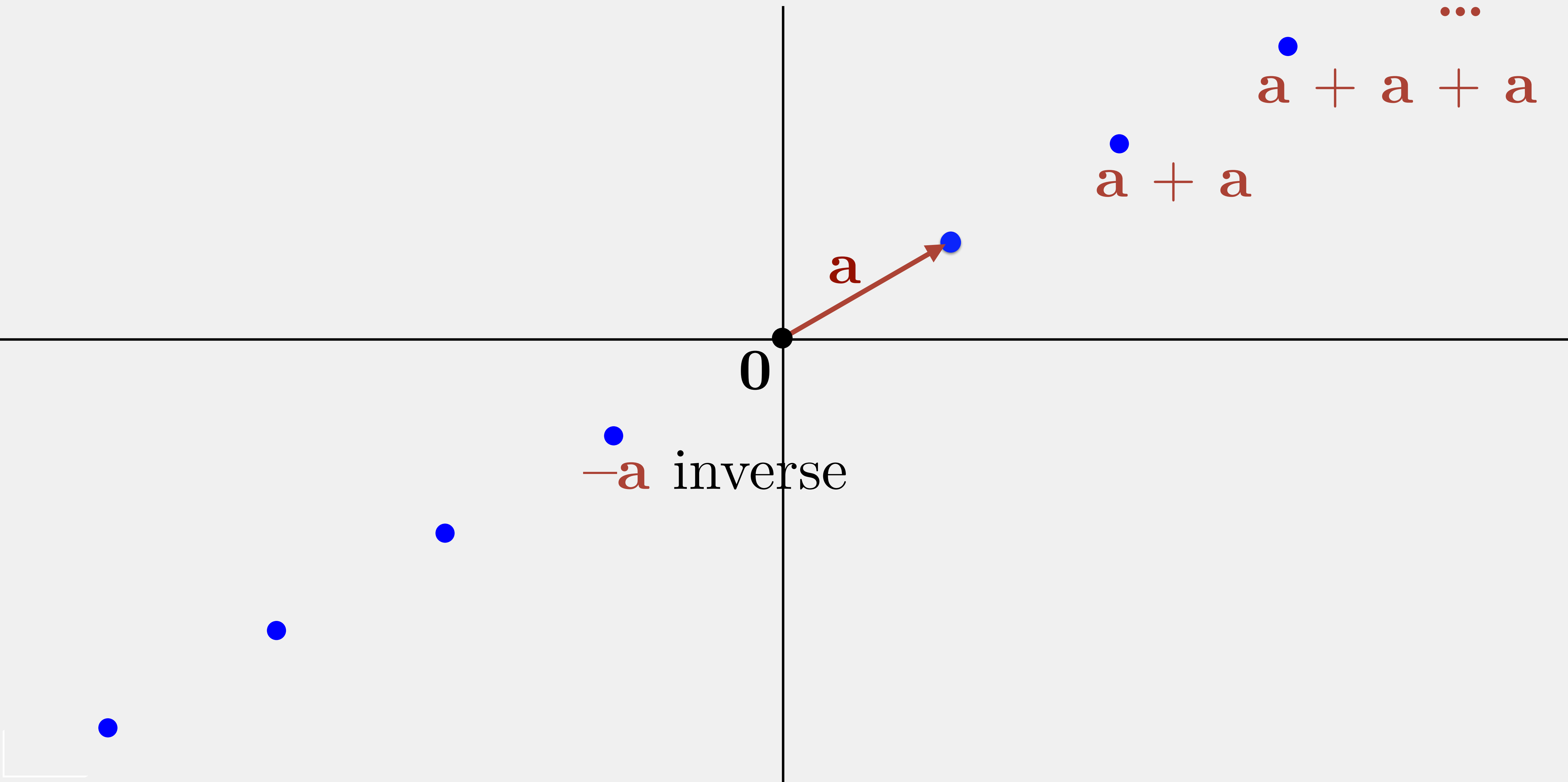


# Lattices in $\mathbb{R}^2$



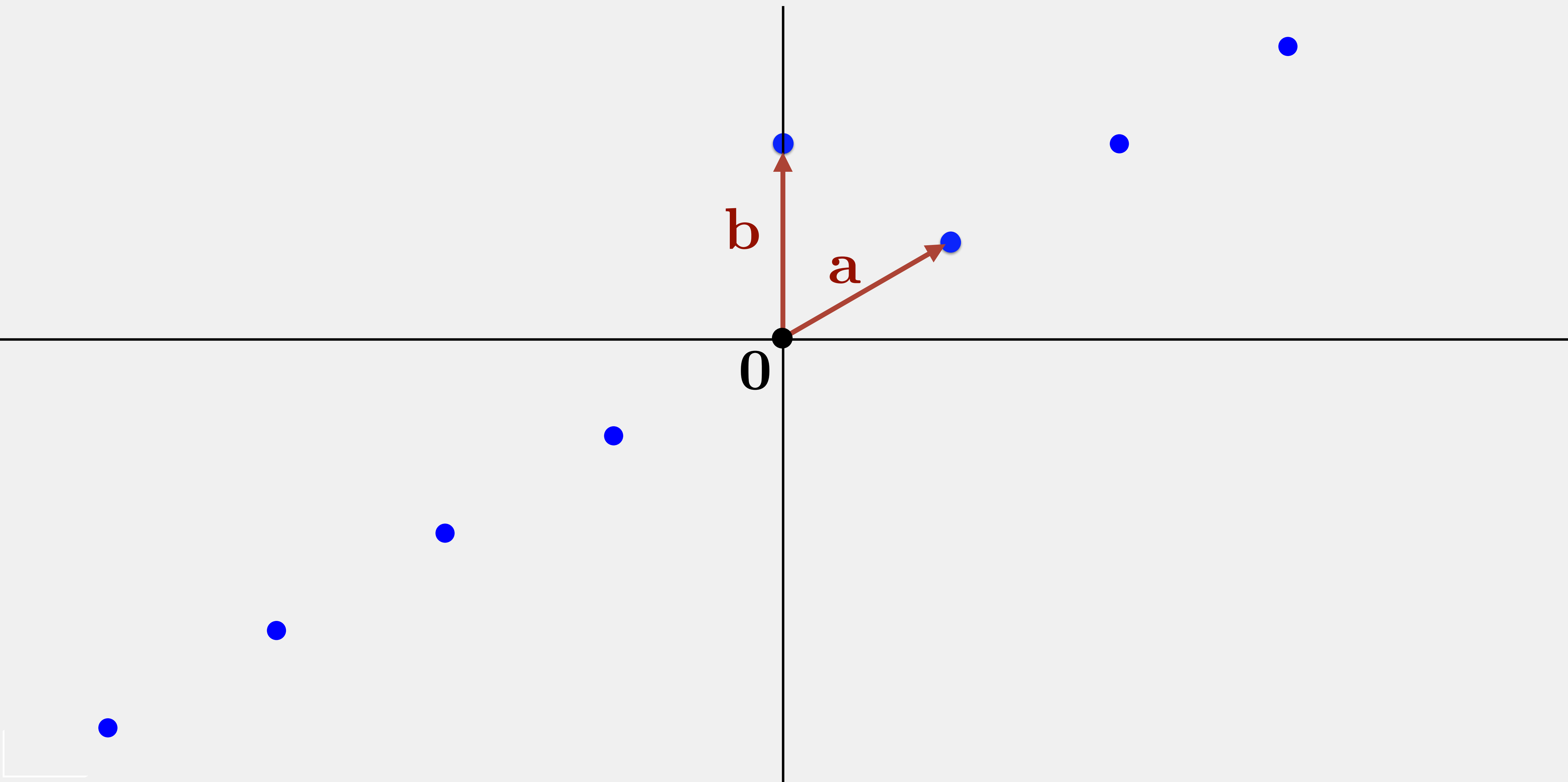


# Lattices in $\mathbb{R}^2$



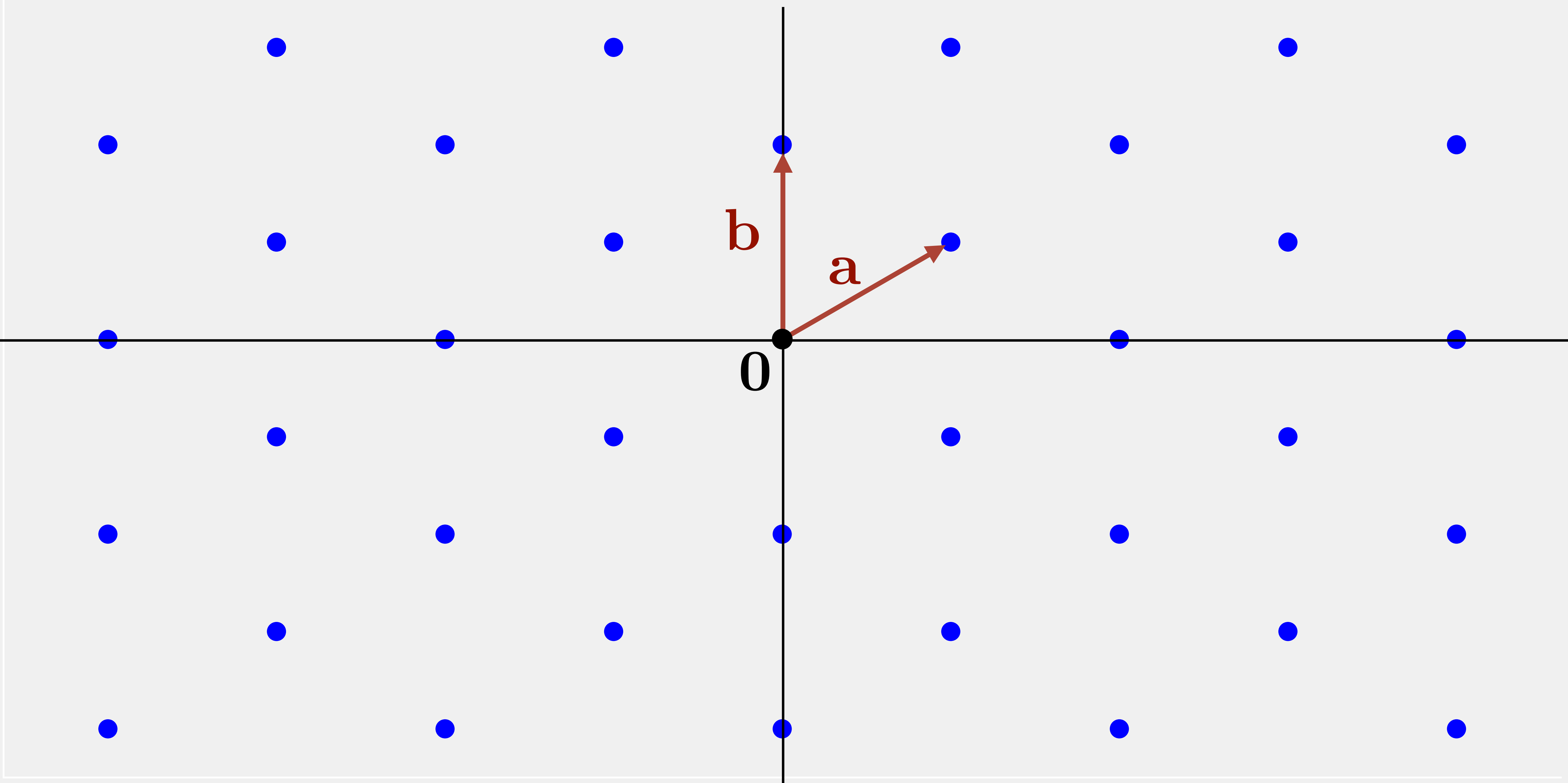


# Lattices in $\mathbb{R}^2$





# Lattices in $\mathbb{R}^2$



# Lattice Generator Matrix

The  $n$ -by- $n$  generator matrix  $G$  is:

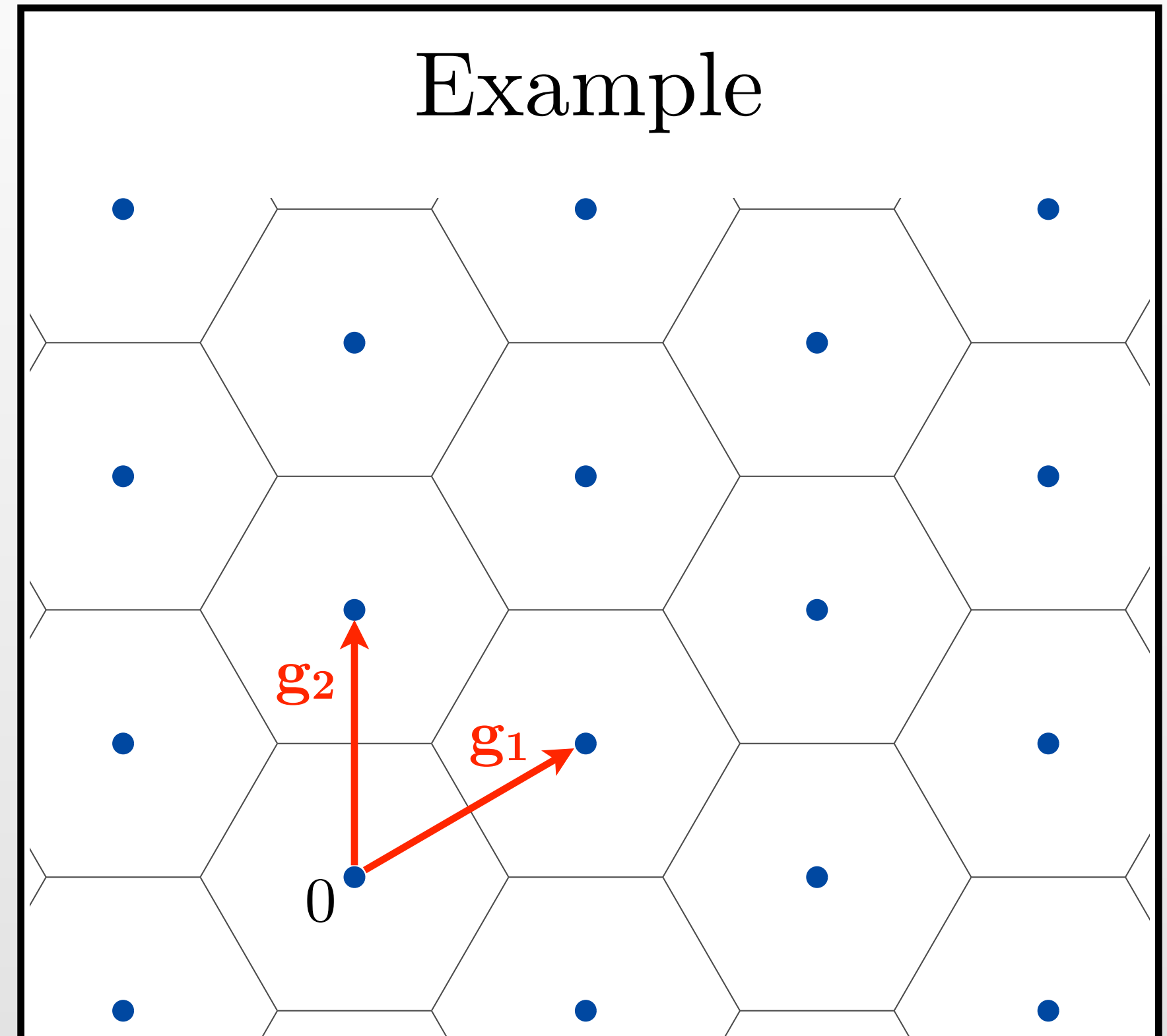
$$G = \begin{bmatrix} | & | & \cdots & | \\ \mathbf{g}_1 & \mathbf{g}_2 & & \mathbf{g}_n \\ | & | & & | \end{bmatrix}$$

so that:

$$\mathbf{x} = \mathbf{G} \cdot \mathbf{b}$$

where  $\mathbf{b} \in \mathbb{Z}^n$  is a vector of integers.

Example

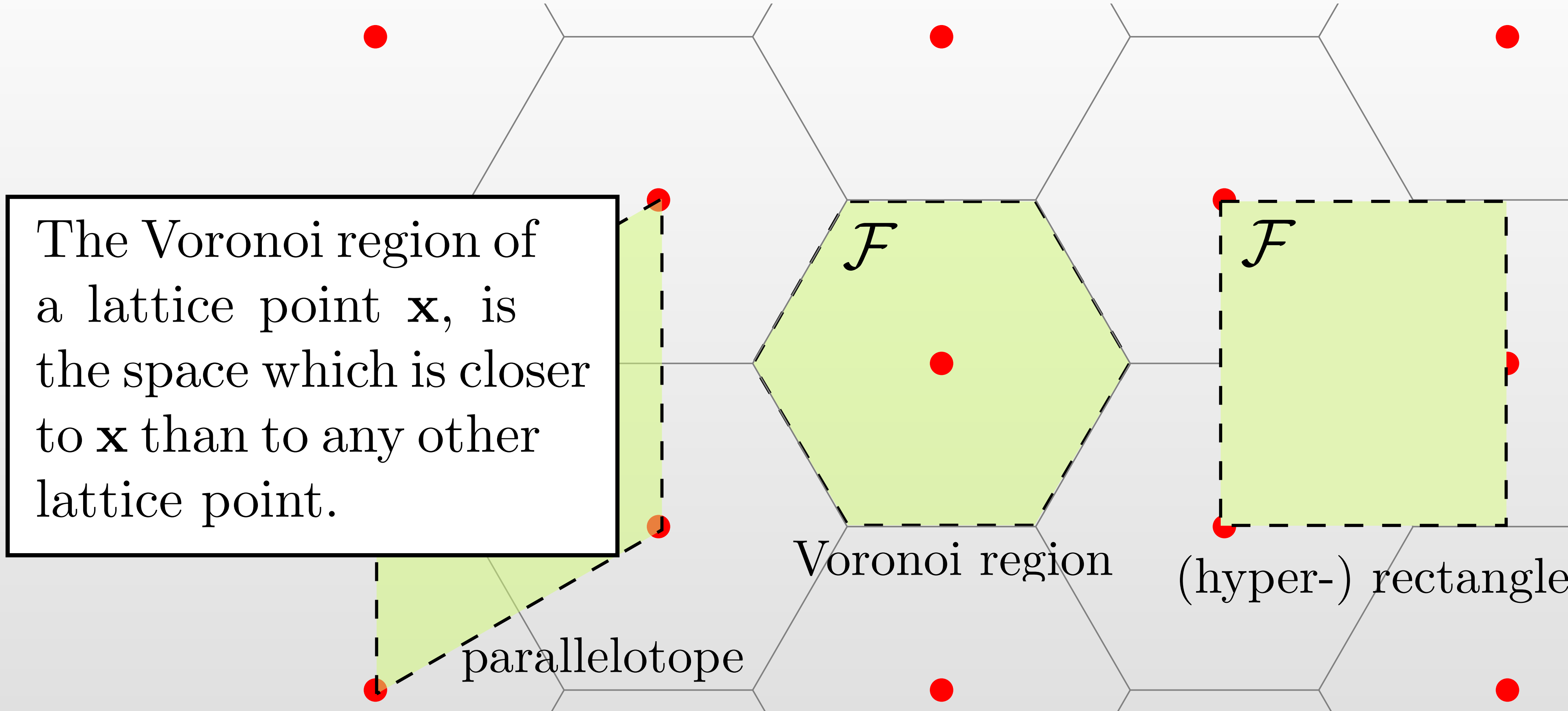


$$\mathbf{G} = \begin{bmatrix} \frac{\sqrt{3}}{2} & 0 \\ \frac{1}{2} & 1 \end{bmatrix}$$

$\mathbf{g}_1$        $\mathbf{g}_2$

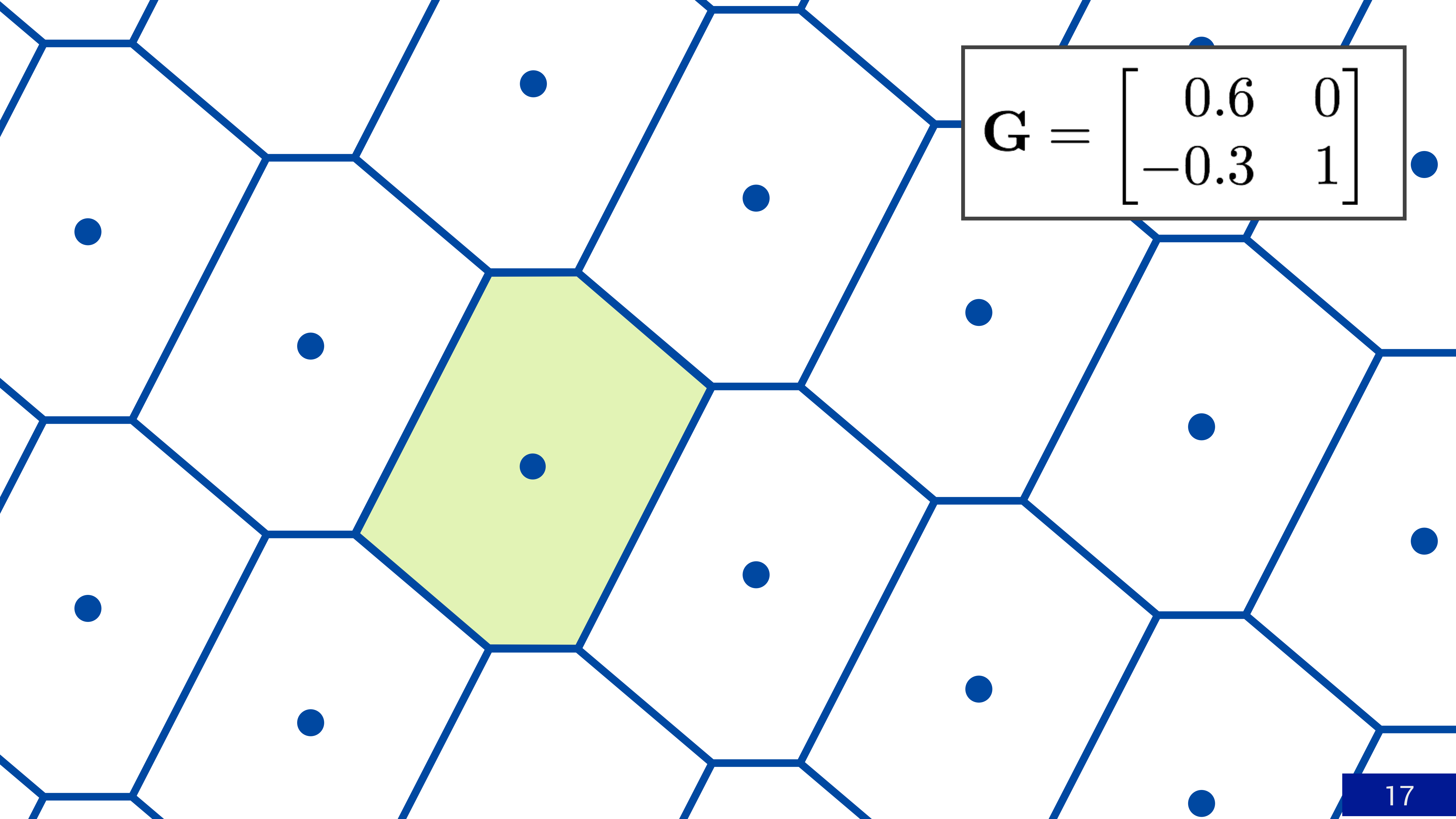


# Fundamental Region



A fundamental region  $\mathcal{F} \subset \mathbb{R}^n$  is a shape that, if shifted by each lattice point, will exactly cover the whole real space.

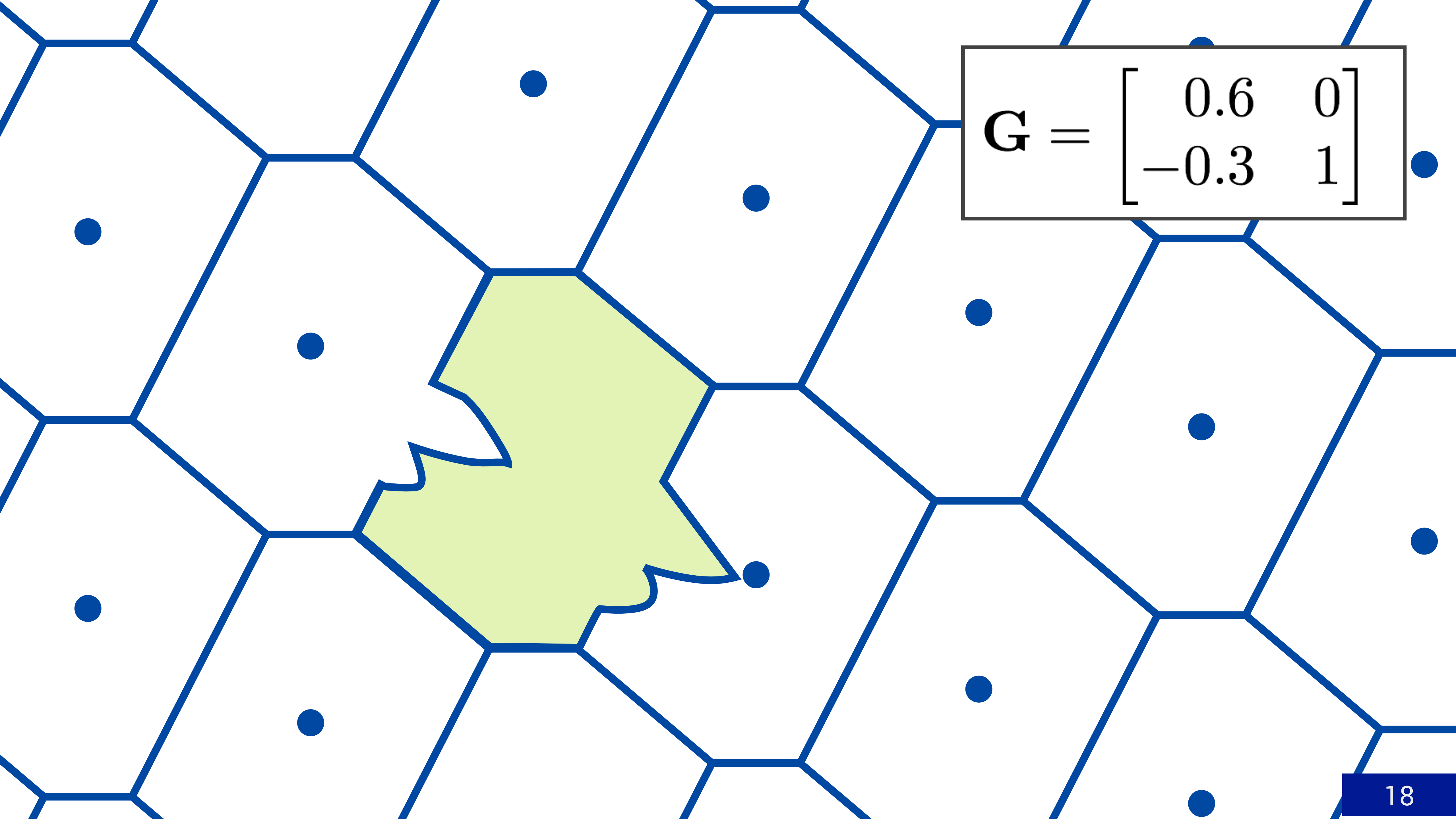
Volume of  $\mathcal{F}$  is  $V(\Lambda) = |\det \mathbf{G}|$ , and is a constant.



The diagram shows a hexagonal lattice of cells. The central cell is highlighted in light green. A box in the top right contains the matrix  $\mathbf{G}$ , and a small blue box in the bottom right contains the number 17.

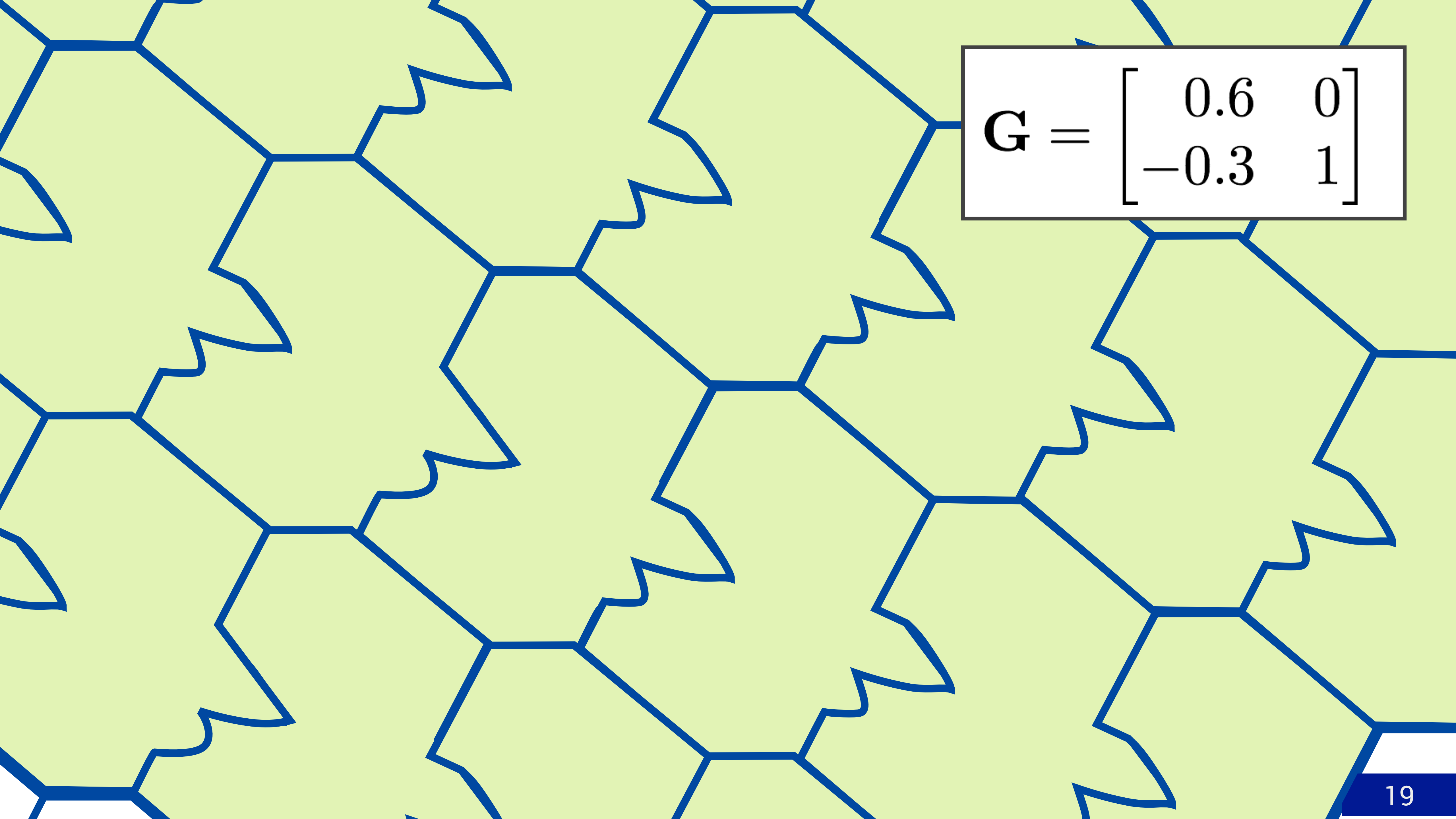
$$\mathbf{G} = \begin{bmatrix} 0.6 & 0 \\ -0.3 & 1 \end{bmatrix}$$



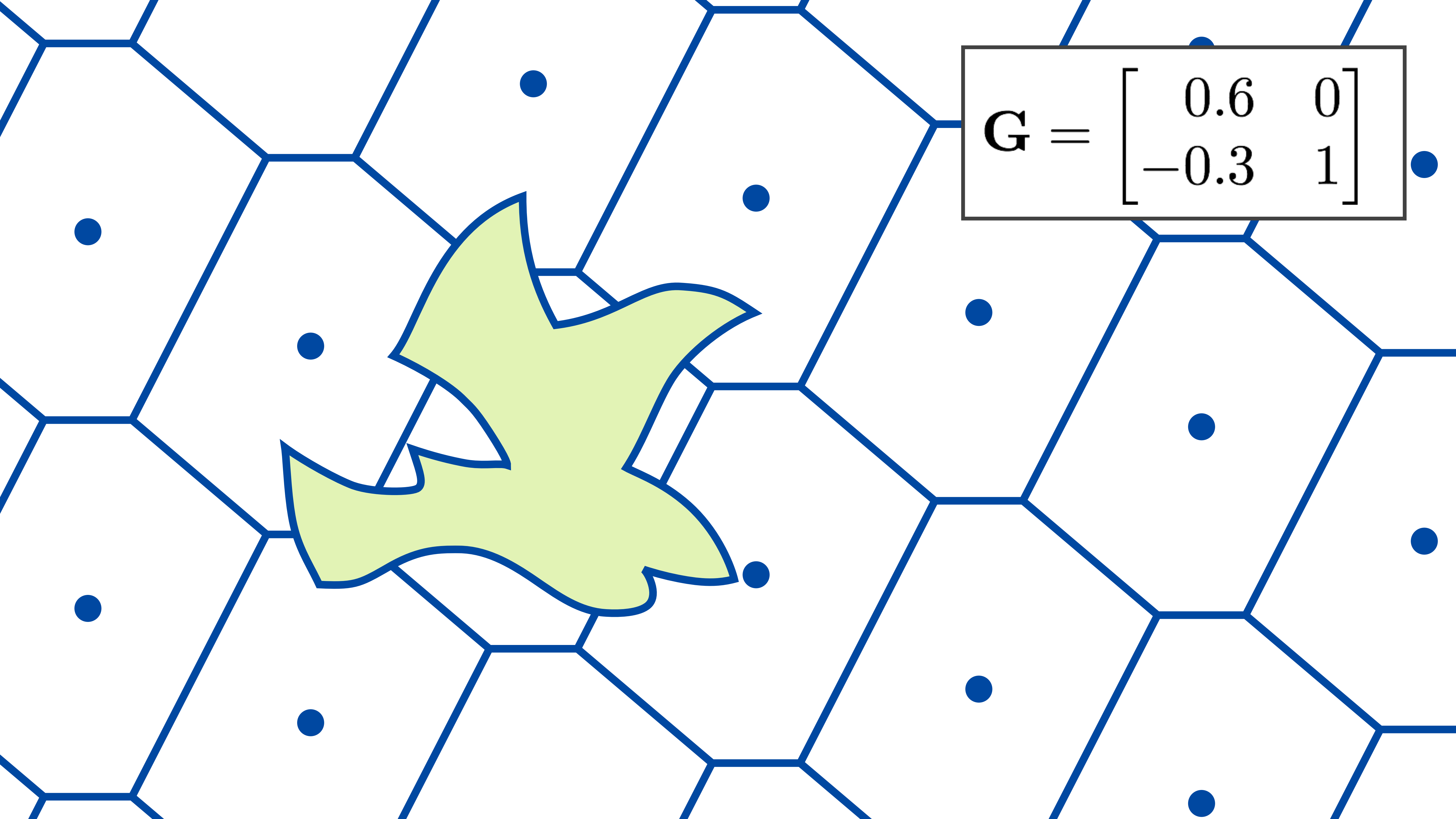


The diagram shows a hexagonal lattice of blue lines. A central region is shaded in light green. Several blue dots are placed at various lattice points. A box in the upper right contains the transformation matrix  $G$ .

$$\mathbf{G} = \begin{bmatrix} 0.6 & 0 \\ -0.3 & 1 \end{bmatrix}$$


$$\mathbf{G} = \begin{bmatrix} 0.6 & 0 \\ -0.3 & 1 \end{bmatrix}$$





The diagram illustrates a hexagonal lattice structure with blue dots at the vertices. A green shaded region is located in the center-left. A box in the top right corner contains the matrix  $\mathbf{G}$ .

$$\mathbf{G} = \begin{bmatrix} 0.6 & 0 \\ -0.3 & 1 \end{bmatrix}$$

$$\mathbf{G} = \begin{bmatrix} 0.6 & 0 \\ -0.3 & 1 \end{bmatrix}$$





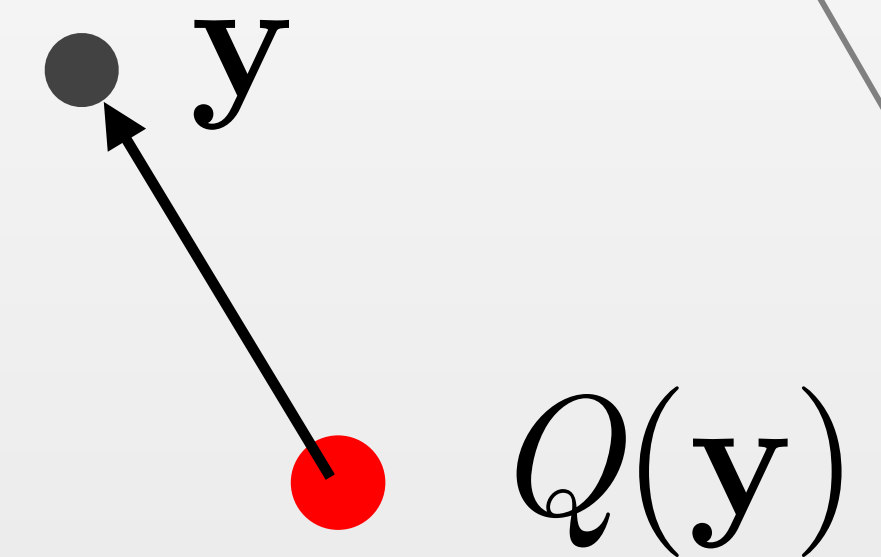
M C Escher  
マウリッツ・エッシャー



# Quantization and Modulo

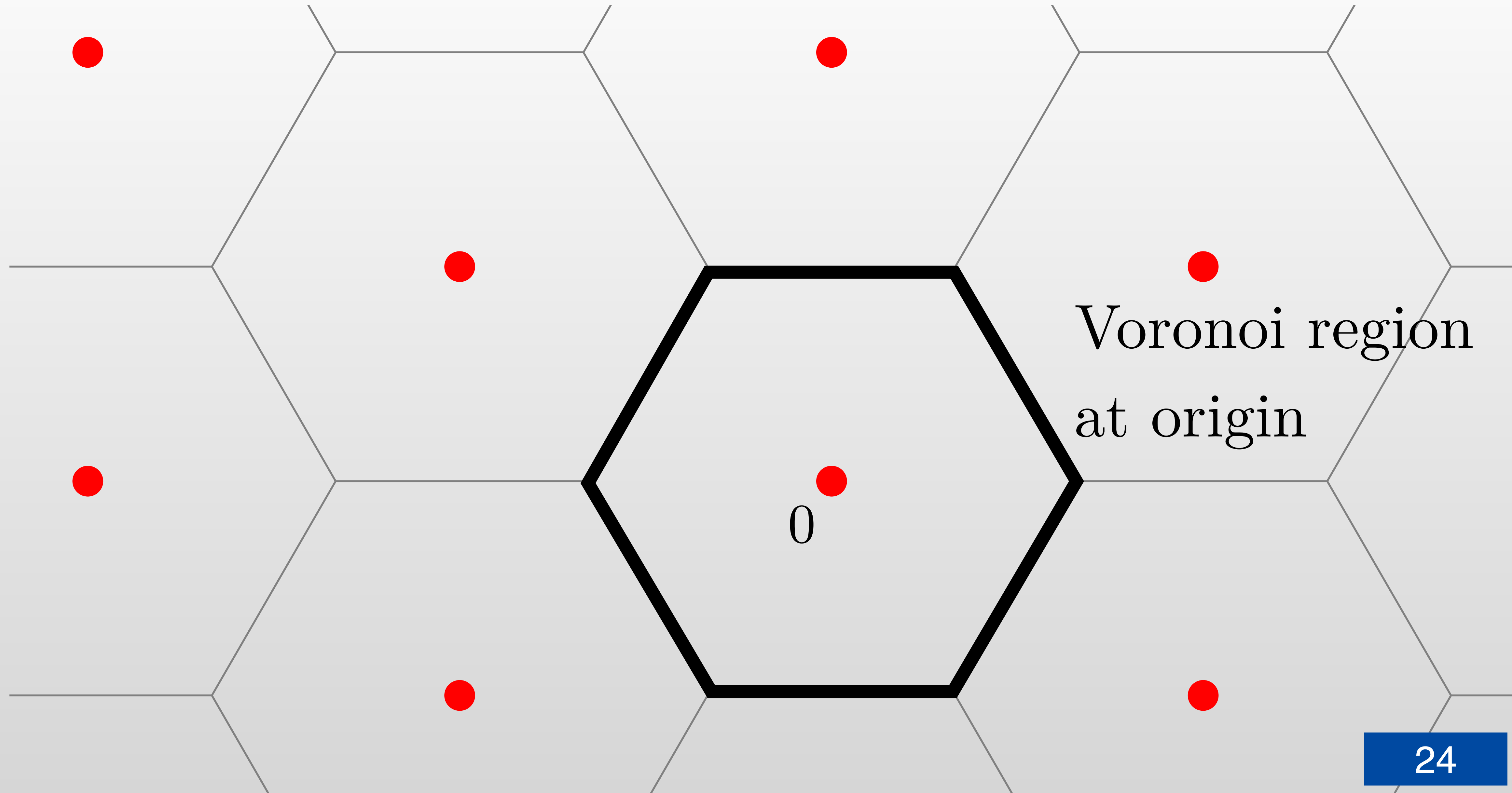
*Quantization* Closest point in  $\Lambda_s$ :

$$Q_{\Lambda_s}(\mathbf{y}) = \arg \min_{\mathbf{x} \in \Lambda_s} \|\mathbf{x} - \mathbf{y}\|^2$$



0

# Quantization and Modulo

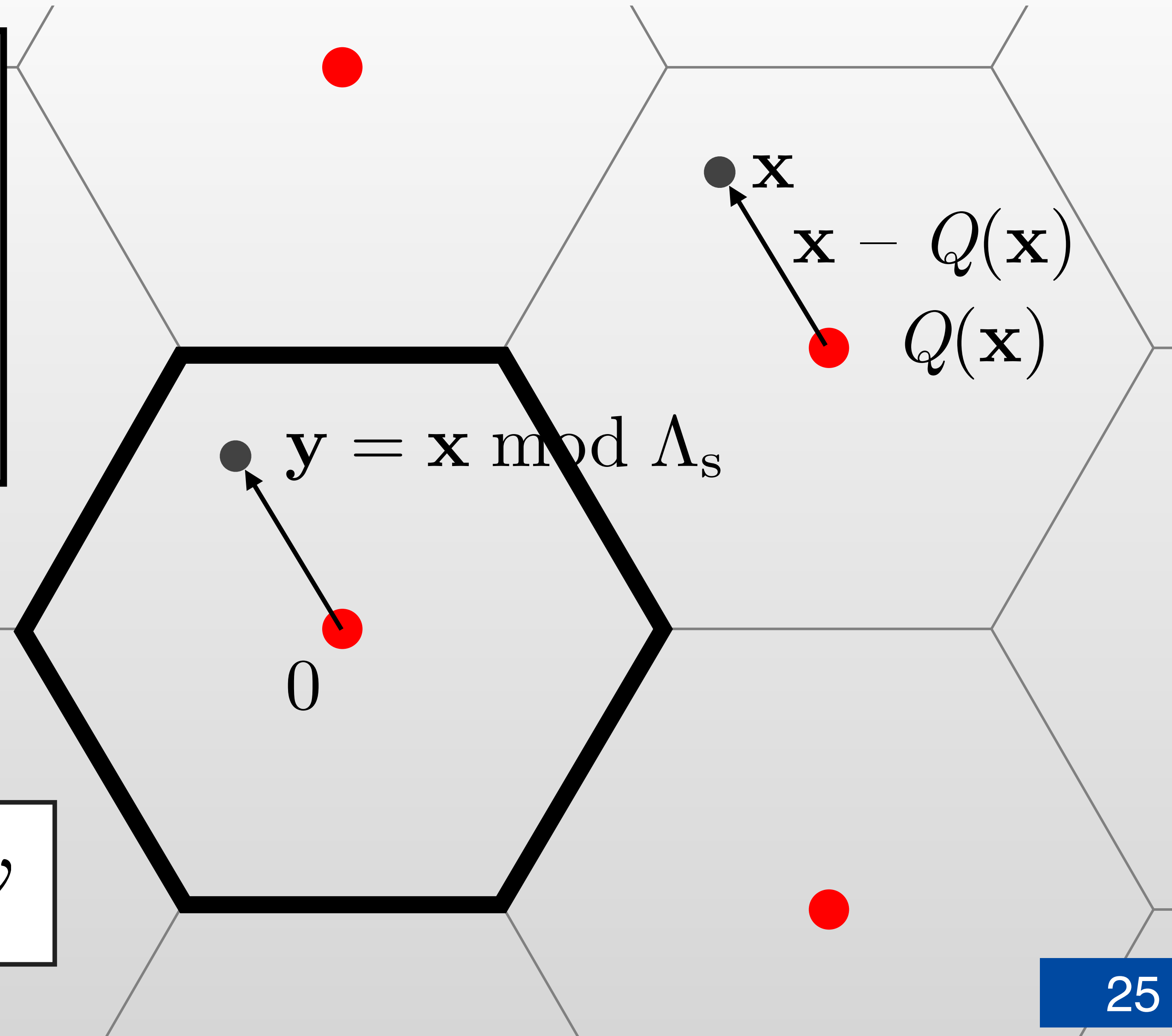




# Quantization and Modulo

Modulo operation:

$$\begin{aligned} \mathbf{y} &= \mathbf{x} \bmod \Lambda_s \\ &= \mathbf{x} - Q_{\Lambda_s}(\mathbf{x}) \end{aligned}$$



find the image of  $\mathbf{x}$  in  $\mathcal{V}$

# Construction D and Construction D'

Construction D and D' are methods to construct lattices from binary codes

Many binary codes have lattice counterpart through Construction D or D':

- Barnes-Wall lattice (from Reed-Muller code)
- LDPC code lattices
- Polar code lattices
- Turbo code lattices

Construction D: Uses binary code's generator matrix

Construction D': Uses binary code's parity-check matrix

**Because binary codes are very well studied, Construction D/D' are the most promising method to construct practical lattices**



# A Tale of Construction D

## Chapter 1 Early Days

Once upon a time, Barnes and Sloane made lattices from binary codes, which they called “Construction D” [CJM 1985]

Soon after that, Forney created the Code Formula construction, to show special lattices can be written as coset codes [IT 1988]

## Chapter 2 Glory Days

Many years pass. Invigorated by Zamir’s lattices, Forney shows that the Code Formula Construction achieves capacity & gives multilevel decoding [IT 2000].

Excited by Code Formula decoding, several researchers create new codes from LDPC, turbo and & codes (2006, 2011, 2013). Multilevel decoding is excellent.

All seems well in the kingdom, until...



## Chapter 3 Dark Days

It is a dark time for Construction  $D/D'$ . Kositwattanakar and Oggier show that Construction  $D/D'$  and the Code Formula Construction agree only in some special cases [DCC 2014].

Code Formula Construction is not a lattice, generally.

In some papers, LDPC “lattices”, turbo “lattices”, polar “lattices” are valid structures, but multilevel decoding is their Code Formula version.



# How to Decode Construction D?



Krishna Narayanan,  
Texas A&M Univ

Those previous “lattices” were decoded as Code Formula, not lattices. How to decode Construction D/D’ lattices?

How to decode Construction D is known.

Actually, *you* showed that.

Not clear yet how to decode Construction D’.  
(at that time)



## Chapter 3 Dark Days

It is a dark time for Construction D/D'. Kositwattananarerk and Oggier show that Construction D/D' and the Code Formula Construction agree only in some special cases [DCC 2014].

Code Formula Construction is not a lattice, generally.

LDPC “lattices”, turbo “lattices”, polar “lattices” are valid structures, but multilevel decoding is their Code Formula version.

## Chapter 4 A New Beginning

Vem, Huang, Narayanan, Pfister make a decoder for Construction D (but not for Construction D') [ISIT 2014]

Finally a decoder for Construction D! Branco da Silva and Silva show how to decode lattice based on binary LDPC codes. [ISIT 2018]

And the lattices lived happily ever after.



# Construction D': LDPC-like Example

LDPC check matrix  $9 \times 12$  for two nested codes

$$\tilde{\mathbf{H}}_0 = \begin{bmatrix} 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 \end{bmatrix}$$

$\mathcal{C}_0$   
 $\mathcal{C}_1$

# Construction D': LDPC-like Example

Lattice check matrix  $12 \times 12$

$$\mathbf{H} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1/2 & 0 & 1/2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1/2 & 0 & 1/2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1/2 & 0 & 0 & 0 & 0 & 1/2 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1/2 & 0 & 1/2 & 0 & 1/2 & 0 & 0 & 0 & 0 & 0 \\ 1/2 & 0 & 0 & 0 & 0 & 1/2 & 0 & 1/2 & 0 & 0 & 0 & 0 \\ 0 & 1/2 & 0 & 1/2 & 0 & 0 & 0 & 0 & 1/2 & 0 & 0 & 0 \\ 1/4 & 0 & 0 & 1/4 & 0 & 0 & 1/4 & 0 & 0 & 1/4 & 0 & 0 \\ 0 & 1/4 & 0 & 0 & 1/4 & 0 & 0 & 1/4 & 0 & 0 & 1/4 & 0 \\ 0 & 0 & 1/4 & 0 & 0 & 1/4 & 0 & 0 & 1/4 & 0 & 0 & 1/4 \end{bmatrix}$$

$\begin{array}{c} \uparrow \\ \mathcal{C}_0 \\ \downarrow \end{array}$ 
 $\begin{array}{c} \uparrow \\ \mathcal{C}_1 \\ \downarrow \end{array}$

# Two Methods for LDPC Lattice Construction

$$\tilde{\mathbf{H}}_0 = \begin{bmatrix} 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 \end{bmatrix}$$

$\mathcal{C}_0$

$\mathcal{C}_1$

Problem: Both Code C0 and C1 should have column weight 3. Code C0 should have higher row weight than Code C1.  
(assuming regular codes)

**Solution 1: Check node splitting** Design code C0 such that linear combination of two rows has no overlaps, and can be used to form rows of higher degree for code C1. Designed using PEG algorithm and extensive simulations [Branco da Silva and Silva]

**Solution 2: Minimum distance design** Code C1 should have  $d_{\min} = 4$ . Code C0 should have  $d_{\min} = 16$ . C1 is a product code of single-parity check codes. C0 is a quasi-cyclic LDPC code from IEEE 802.16e with  $d_{\min} \approx 16$  [Chen, K, Rosnes]

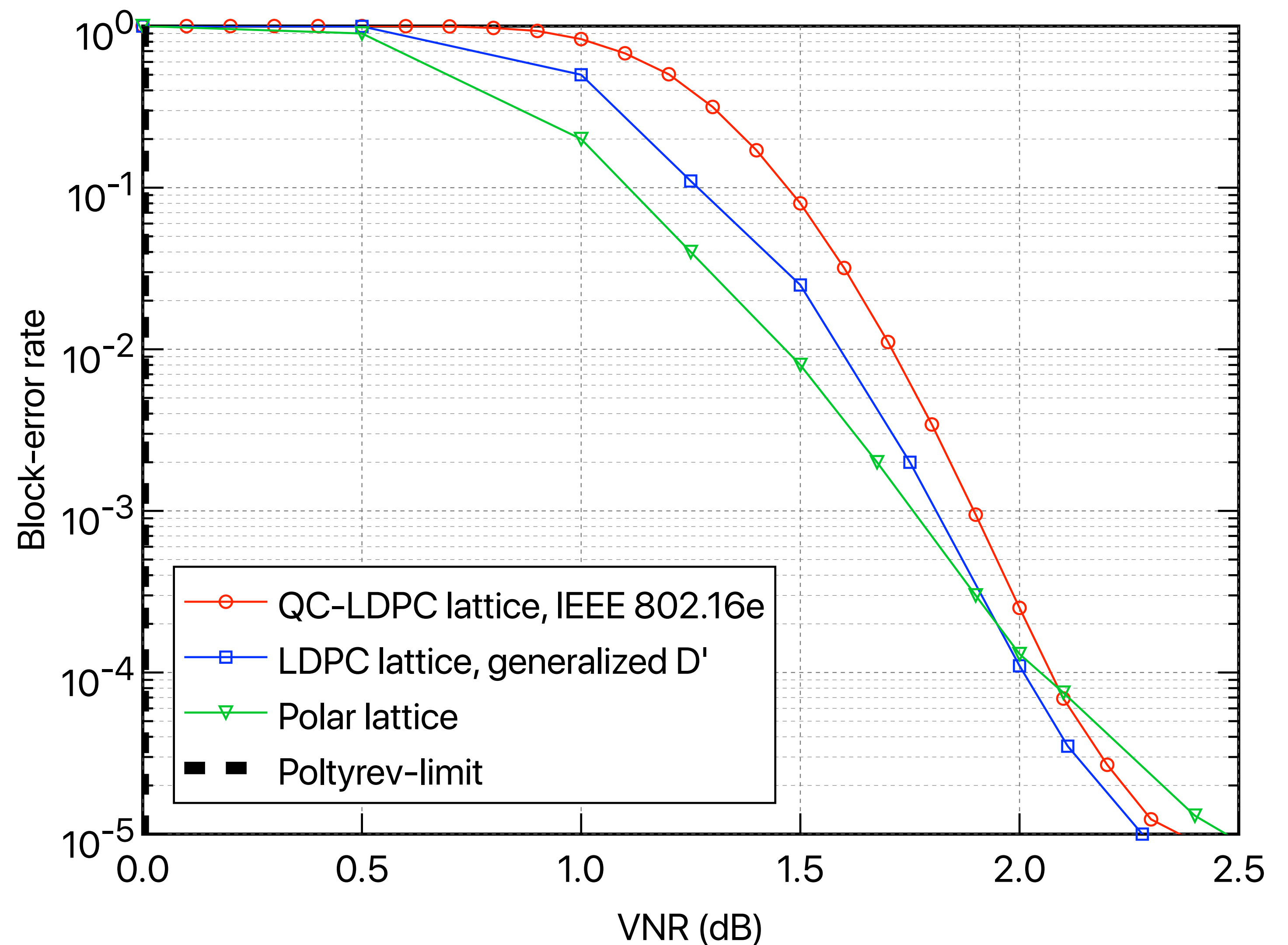


# Error Rate for LDPC Code Lattices

Proposed QC-LDPC code lattices  
loose about 0.1 dB w.r.t PEG

Minimum distance design rule is a  
more systematic design approach  
than PEG/simulations

QC-LDPC codes are widely used in  
practice. If lattices are to be used  
in practice, construction D' with  
QC-LDPC codes are a likely  
candidate.



# Nested Lattice Codes

## (Voronoi Codes, Voronoi Constellations)

**Definition 1.1.** Let  $\Lambda_c$  and  $\Lambda_s$  be two lattices with  $\Lambda_s \subseteq \Lambda_c$ . Let  $\mathcal{F}$  be a fundamental region for  $\Lambda_s$ . Then:

$$\mathcal{C} = \Lambda_c \cap \mathcal{F} \tag{1.1}$$

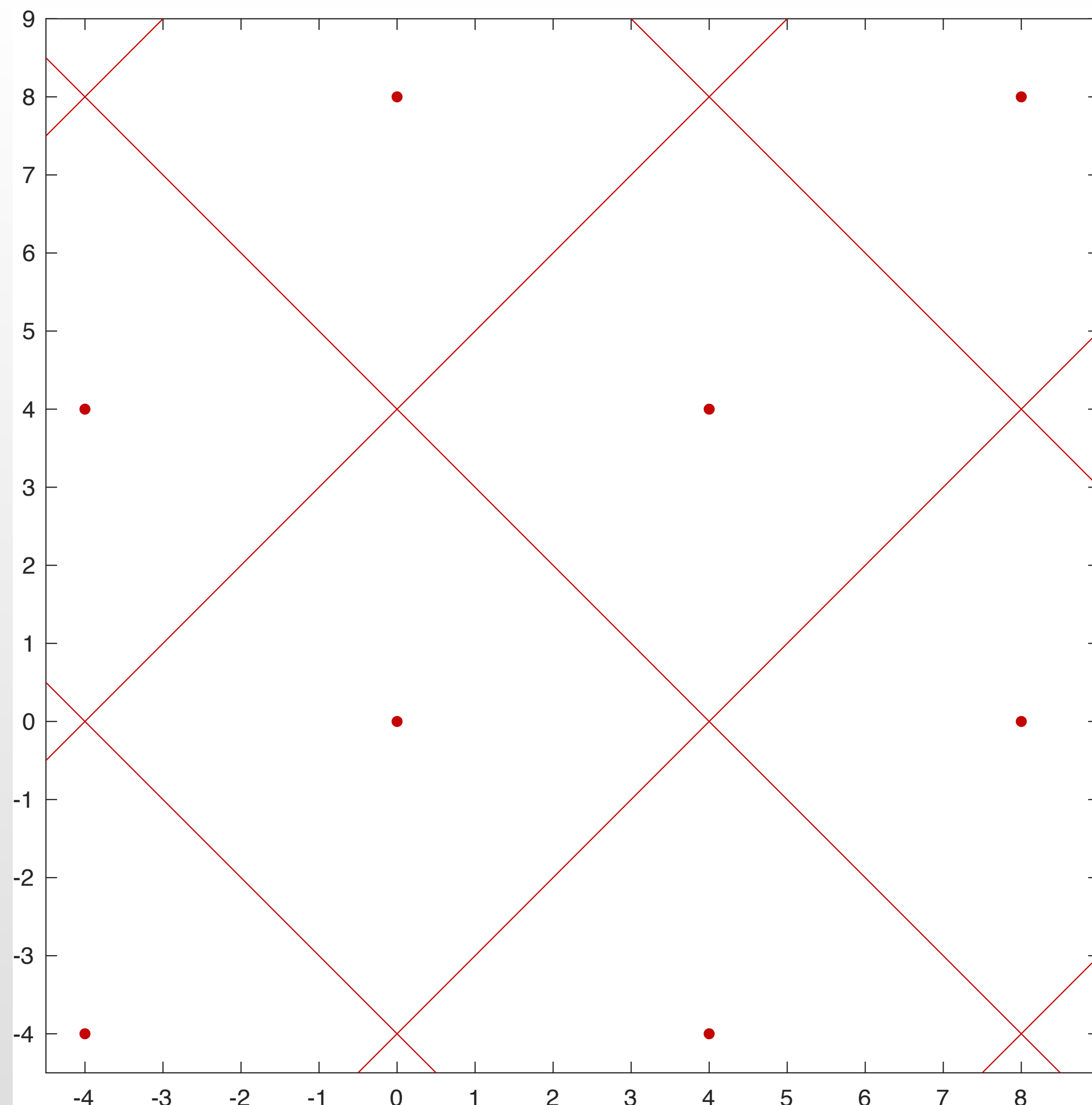
is a *nested lattice code*.

$\Lambda_c$  is called the coding lattice,  $\Lambda_s$  is called the shaping lattice.

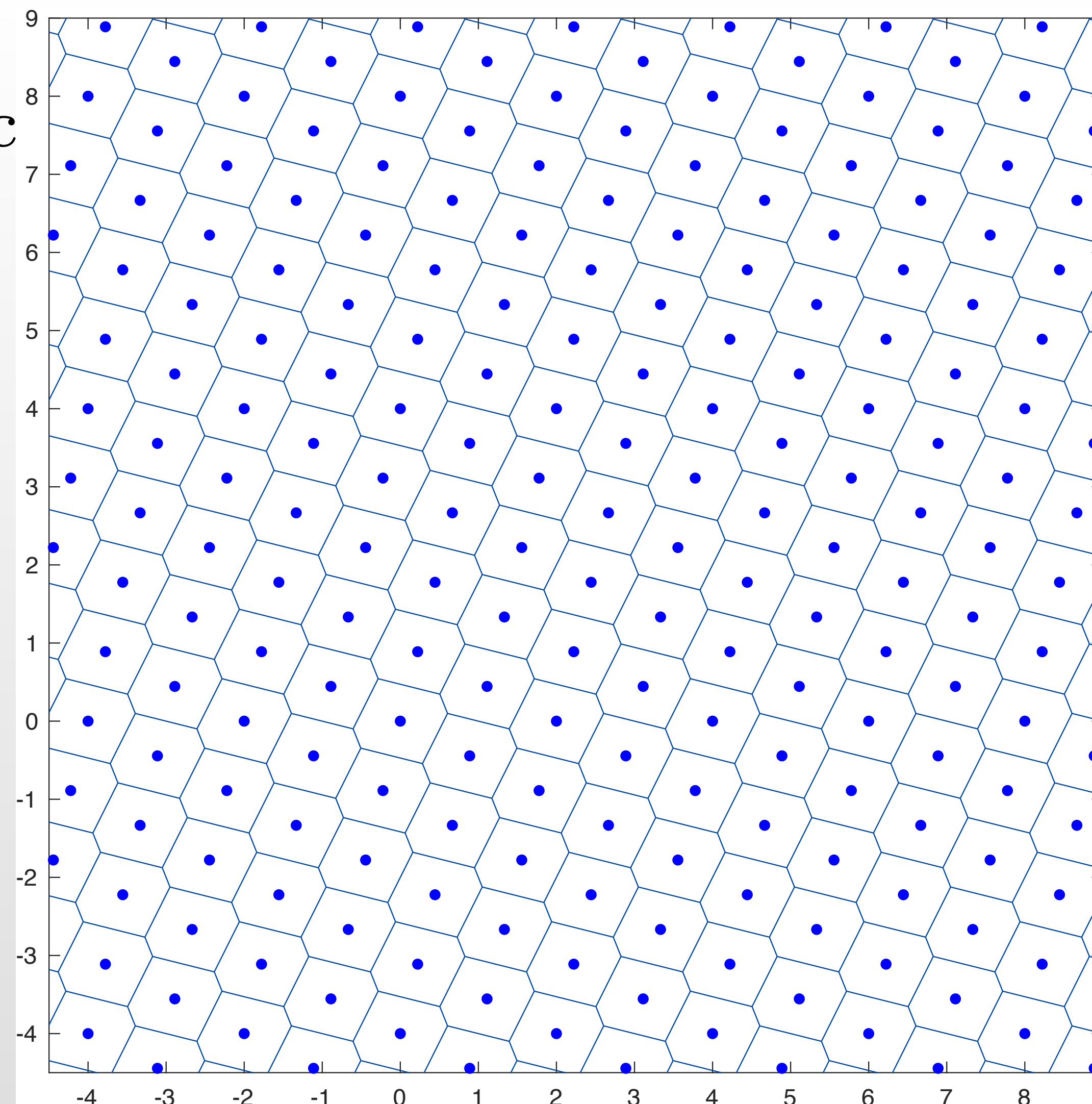
The code rate of a nested lattice code is:

$$R = \frac{1}{n} \log \frac{V(\Lambda_s)}{V(\Lambda_c)} = \frac{1}{n} \log \frac{|\det(\mathbf{G}_s)|}{|\det(\mathbf{G}_c)|}.$$

$\Lambda_s$



$\Lambda_c$

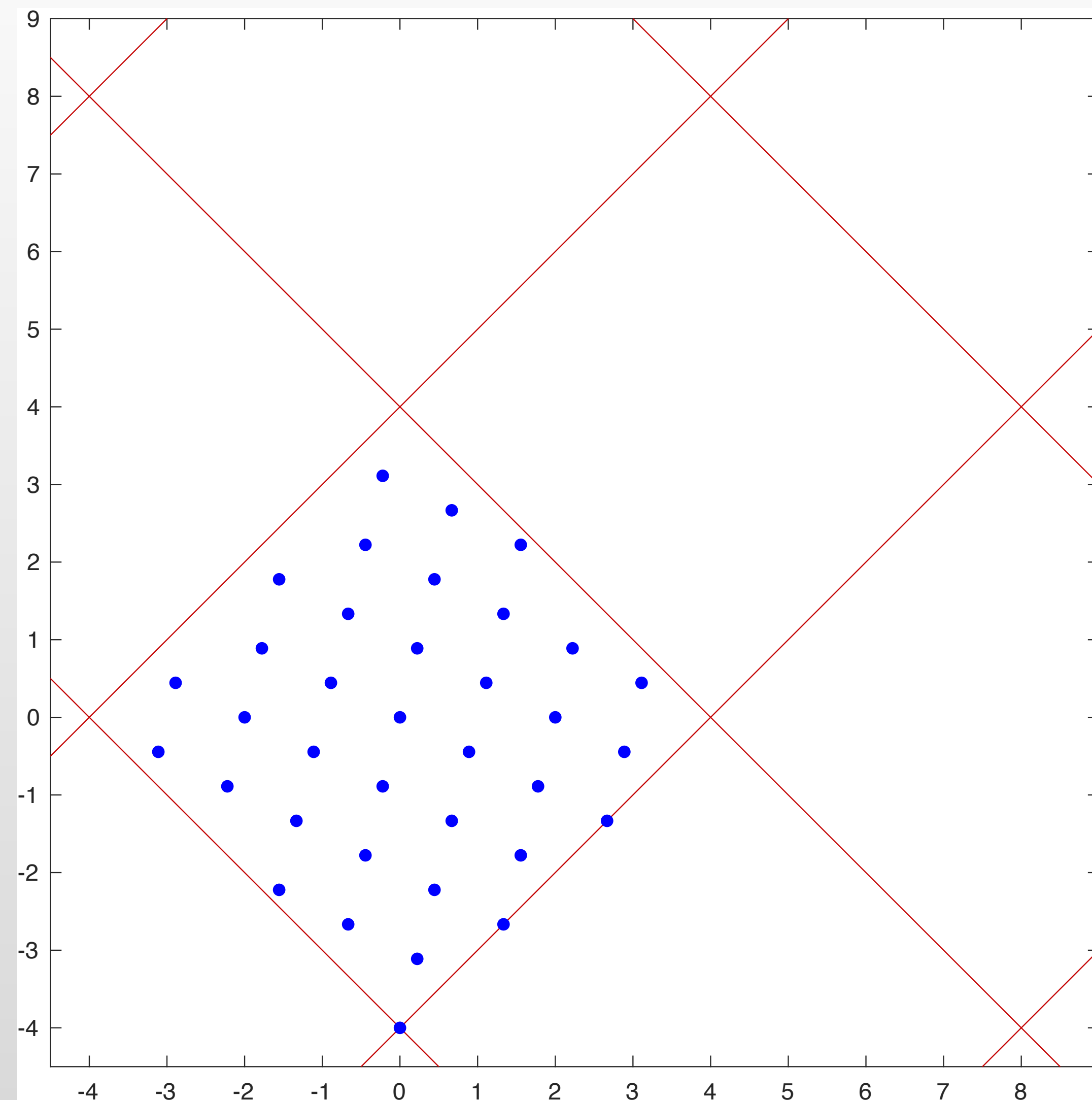
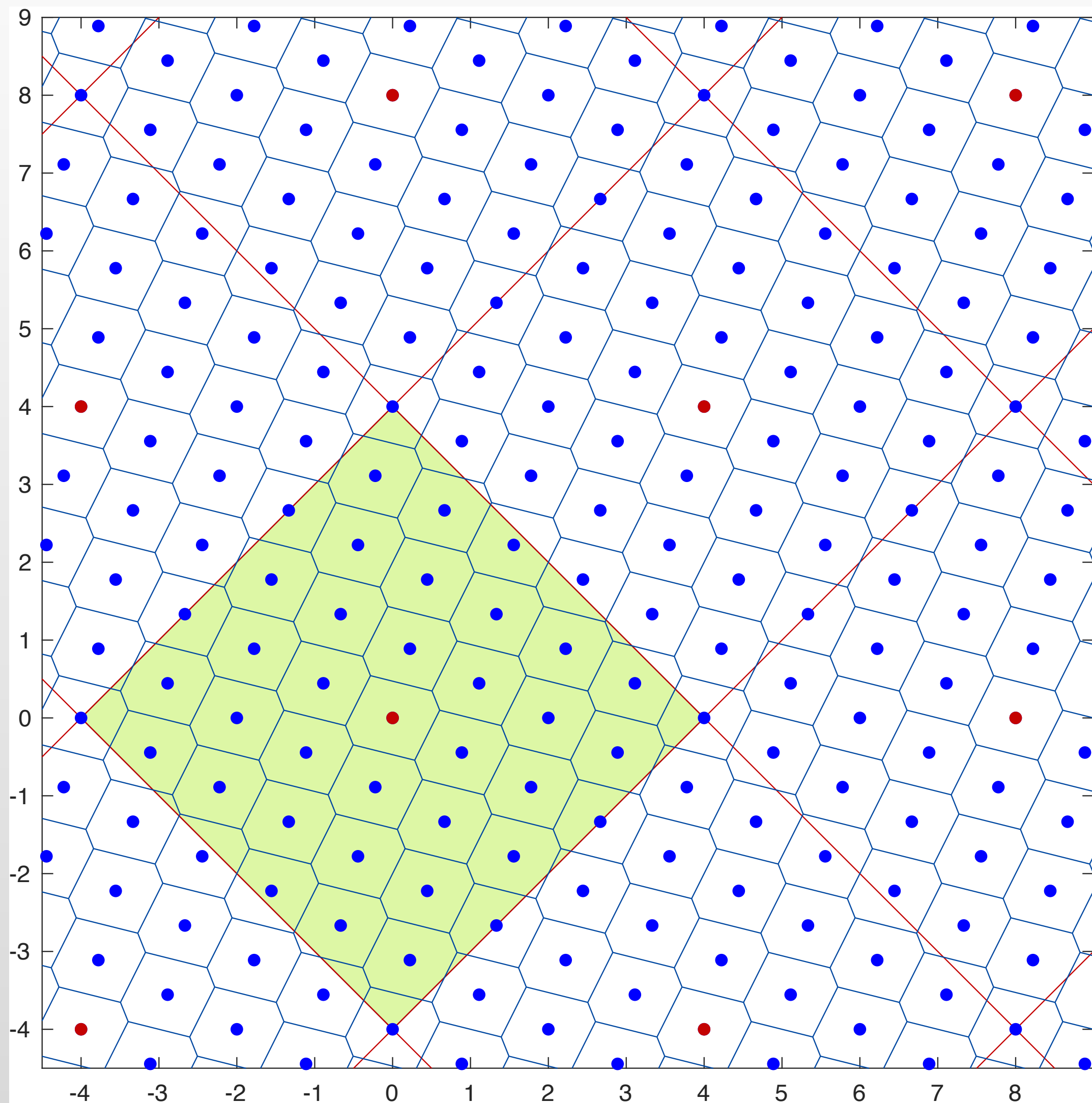


Example

$$\mathbf{G}_s = \begin{bmatrix} 4 & 0 \\ 4 & 8 \end{bmatrix}$$

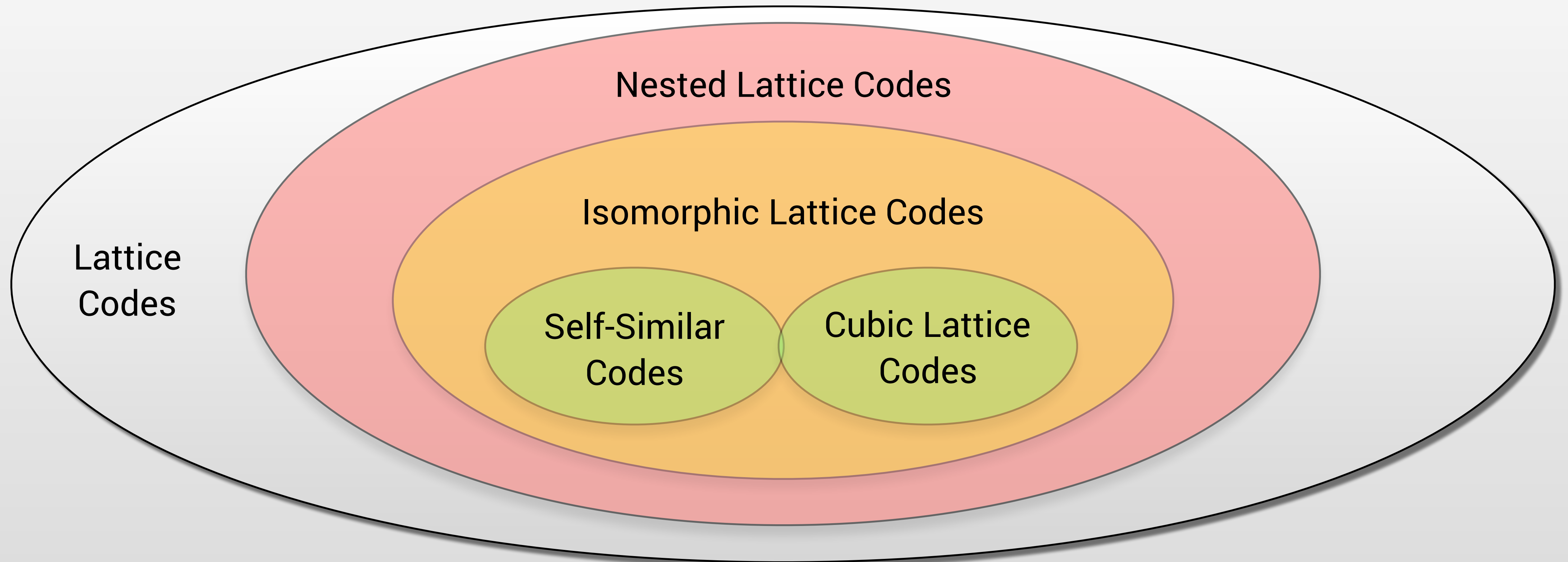
$$\mathbf{G}_c = \begin{bmatrix} \underline{4} & \underline{2} \\ \underline{3} & \underline{9} \\ \underline{4} & \underline{8} \\ \underline{3} & \underline{9} \end{bmatrix}$$







# Classification of Lattice Codes



Isomorphism is important for compute-and-forward.

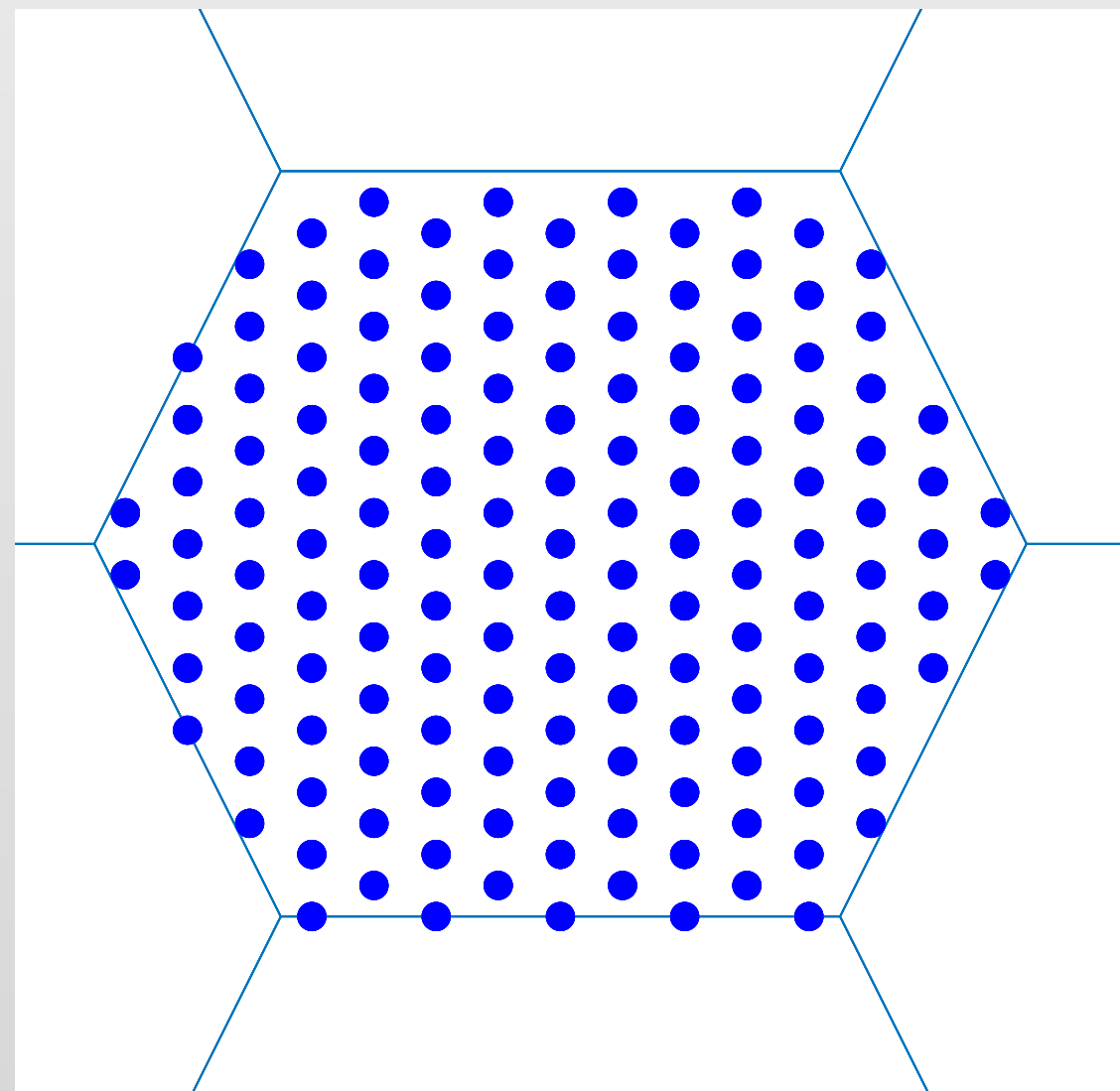


# Self-Similar & Cubic Lattice Codes

## Self-Similar Lattice Code

Shaping lattice is scaled version of coding lattice

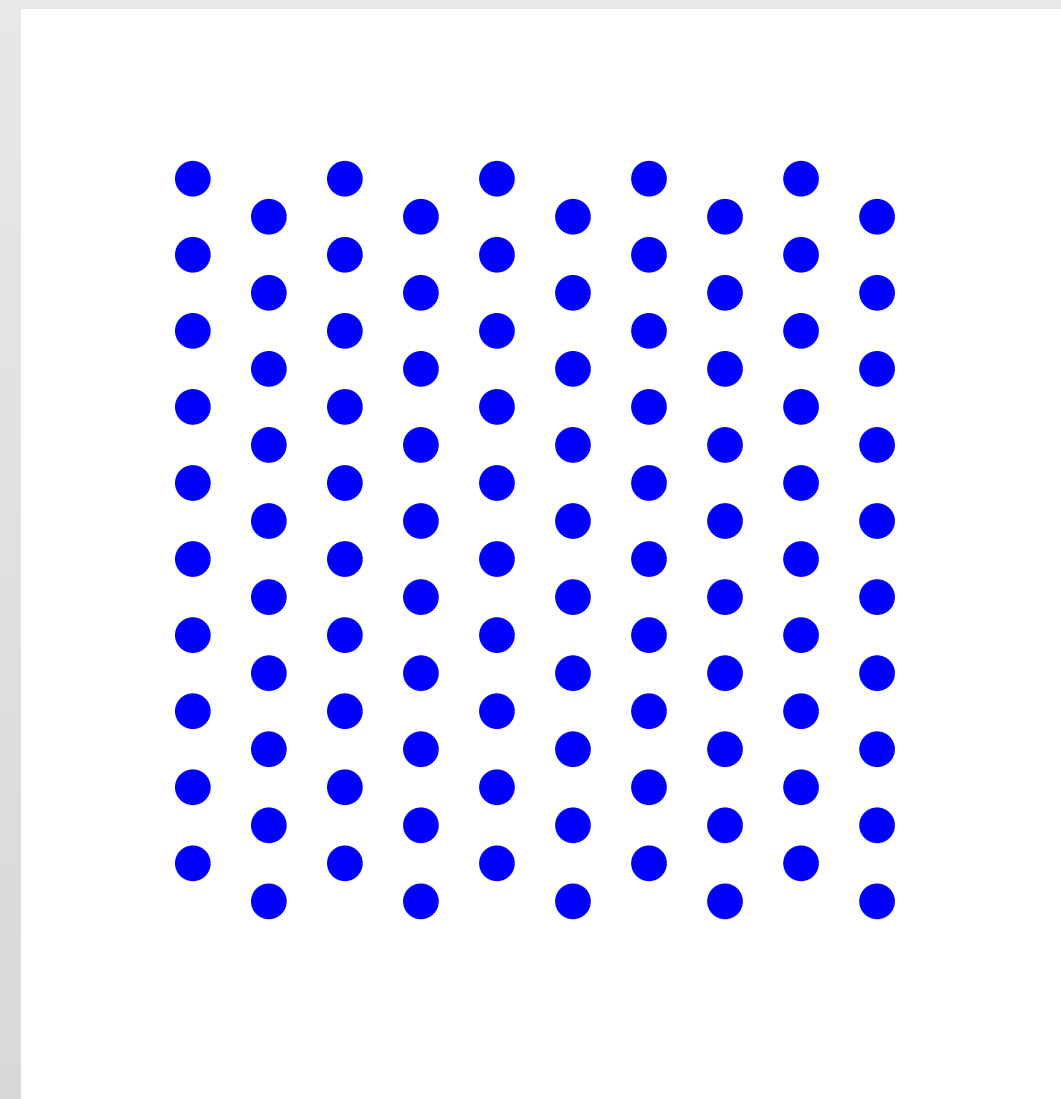
- ✓ Good shaping gain
- ✓ Group isomorphism
- ✗ High encoding complexity



## Cubic Lattice Code

Shaping lattice is a cube

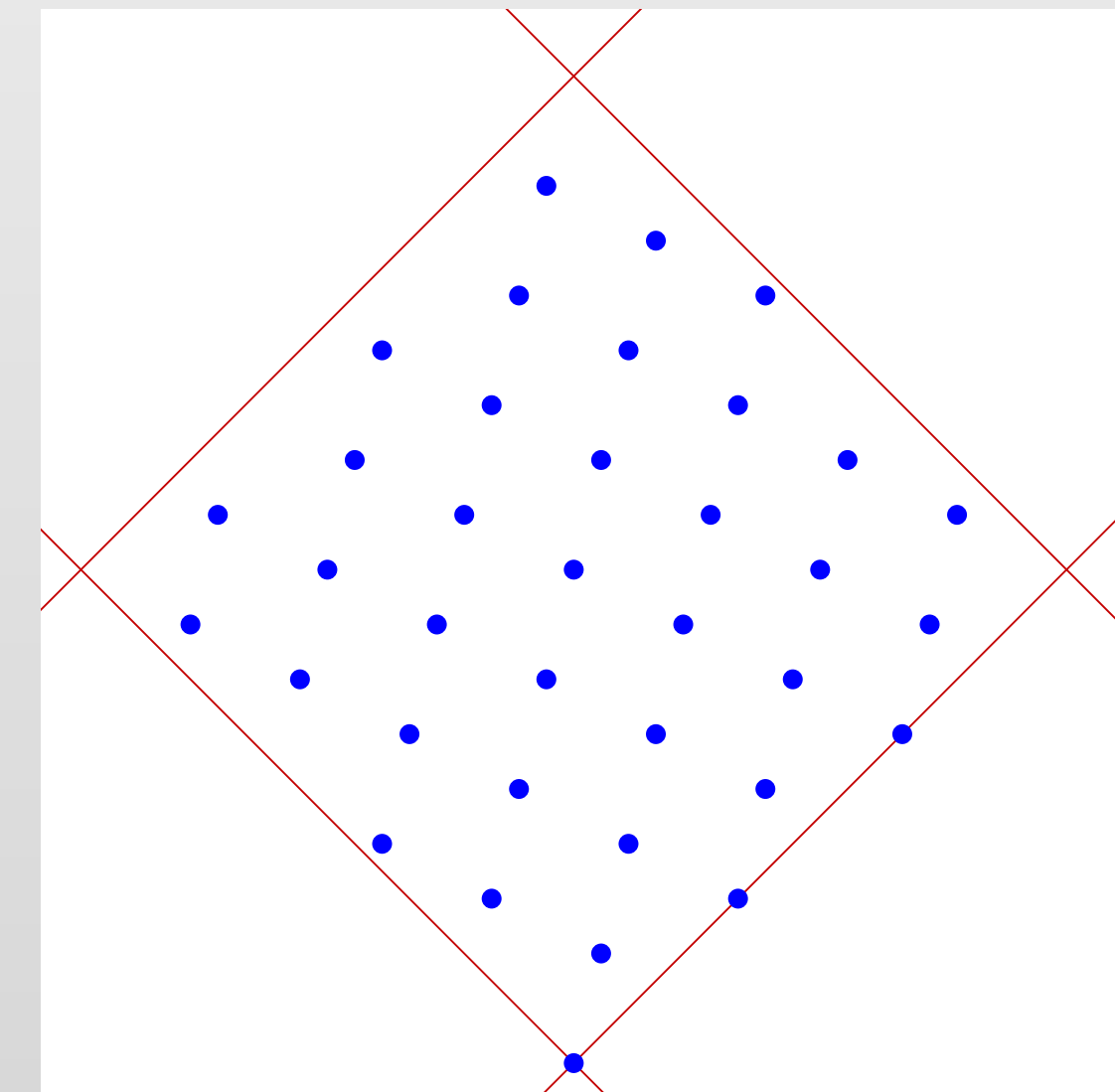
- ✗ No shaping gain
- ✓ Group isomorphism
- ✓ Low encoding complexity



## General Nested Lattice Code

Shaping lattice is sub lattice of coding lattice

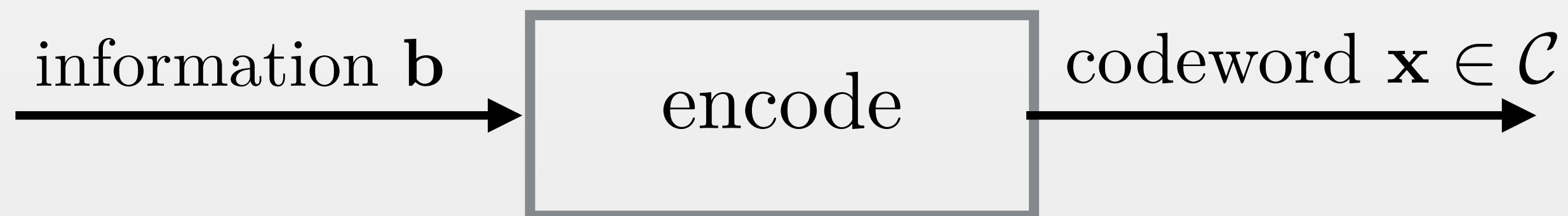
- ✓ Good shaping gain
- ✗ No gr. isomorphism (in general)
- ✓ Low encoding complexity





# Encoding and Indexing

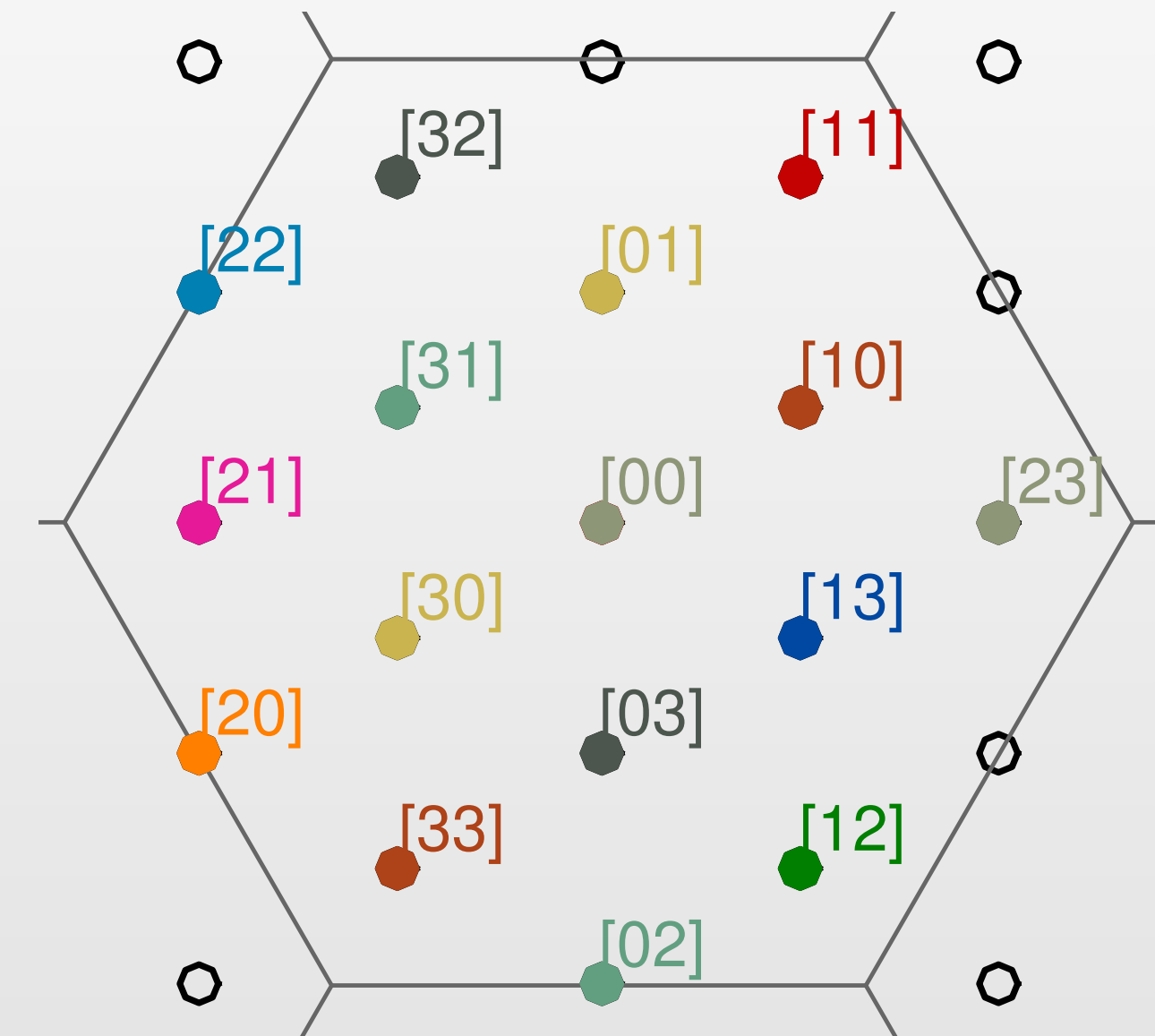
An encoding function maps information (indices)  $\mathbf{b}$  to codewords  $\mathbf{x} \in \mathcal{C}$



Indexing is the inverse of encoding maps, codewords  $\mathbf{x} \in \mathcal{C}$  to information (indices)  $\mathbf{b}$ .



Not decoding: there is no noise.



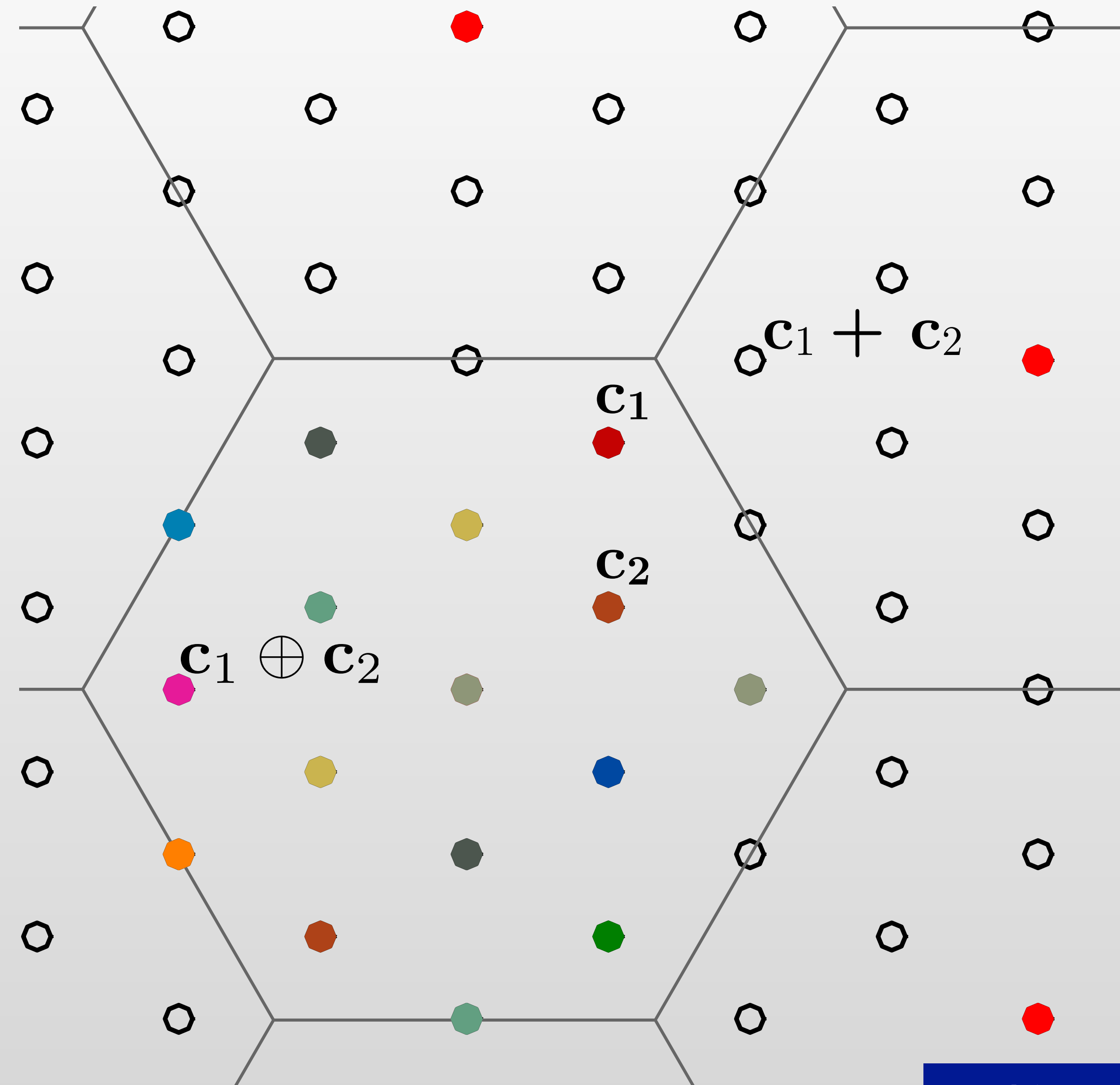
**Main result** encoding and indexing is possible if generator matrices are both in triangular form.



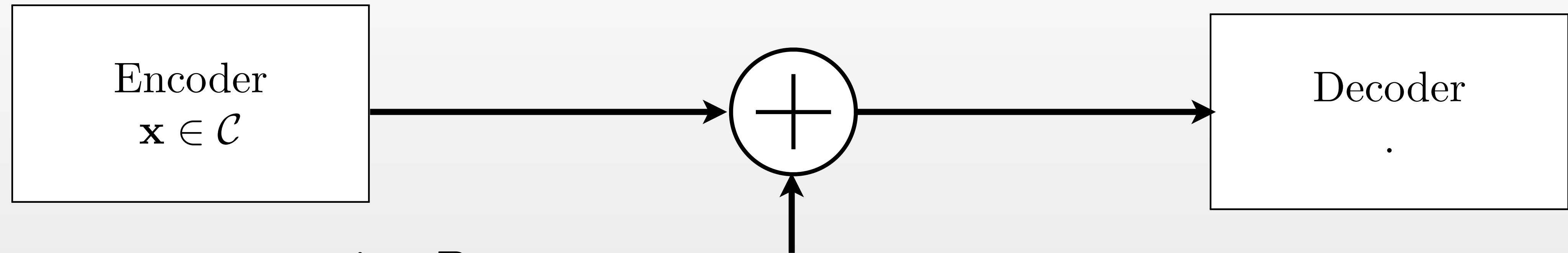
# A Nested Lattice Code is a Group

- Lattice  $\Lambda$  is a group:  $\mathbf{a}, \mathbf{b} \in \Lambda \Rightarrow \mathbf{a} + \mathbf{b} \in \Lambda$
- $\Lambda_s \subseteq \Lambda_c$ . Thus  $\Lambda_s$  is a subgroup of  $\Lambda_c$ .
- The quotient group is  $\Lambda_c/\Lambda_s$ , and is the set of all cosets of  $\Lambda_s$  in  $\Lambda_c$ .
- Group operation. Let  $\mathbf{c}_1, \mathbf{c}_2 \in \Lambda_c/\Lambda_s$ , then:

$$\mathbf{c}_1 \oplus \mathbf{c}_2 = (\mathbf{c}_1 + \mathbf{c}_2) \bmod \Lambda_s$$



# AWGN Channel Capacity

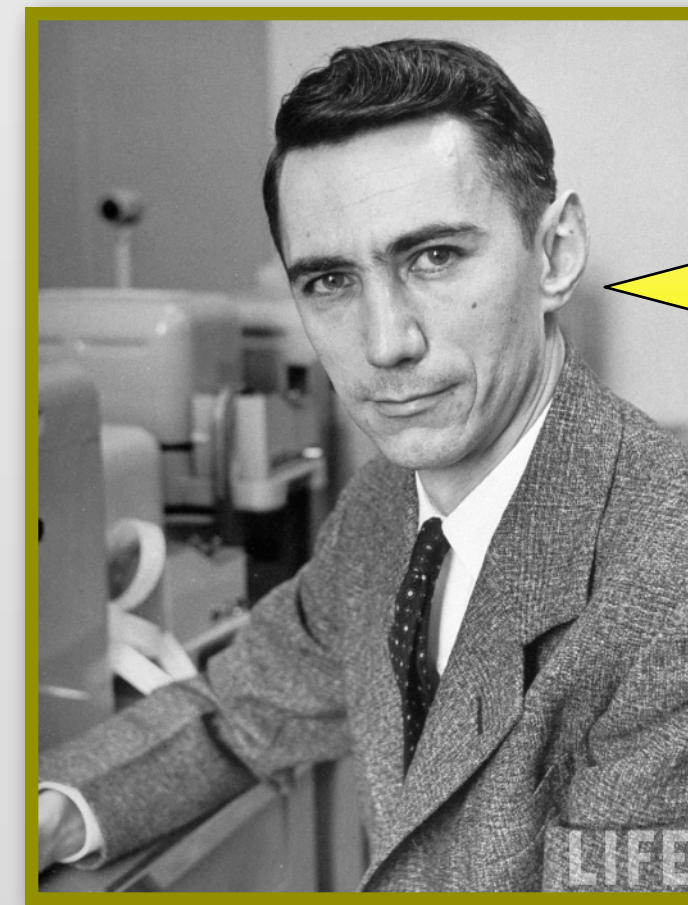


Input power constraint  $P$  :

$$\frac{1}{n} ||\mathbf{x}||^2 \leq P$$

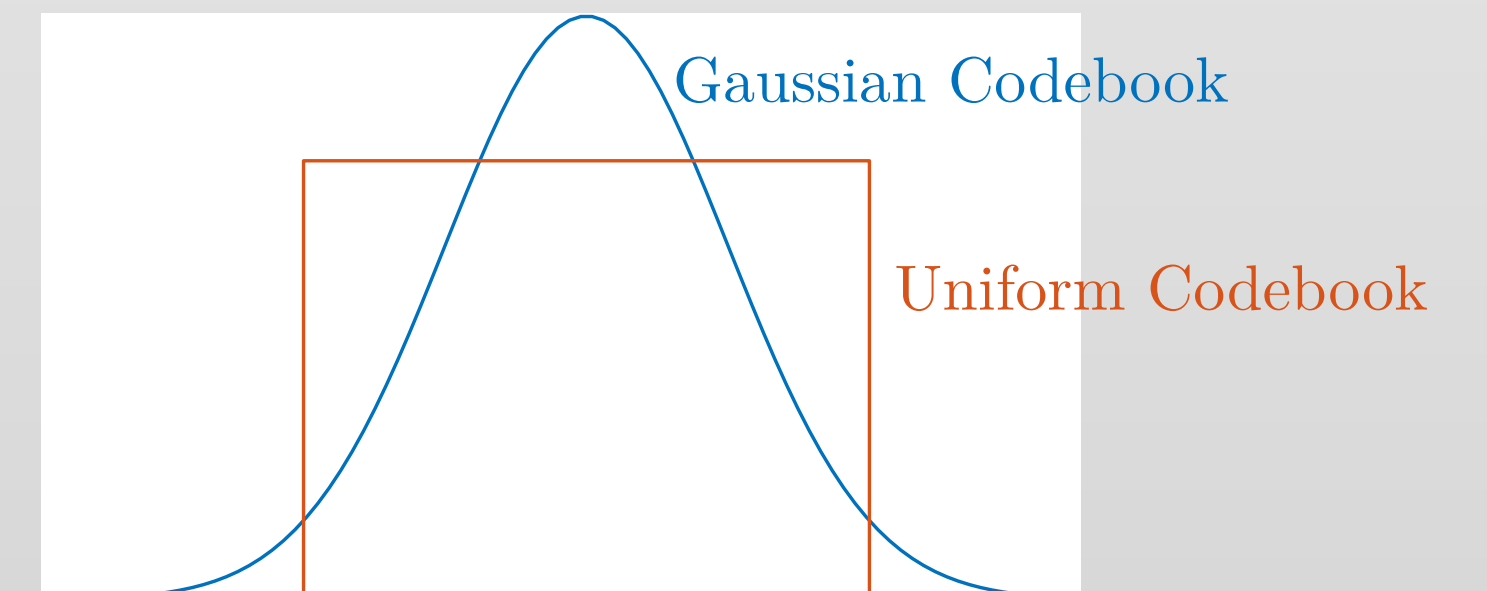
Capacity is:

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



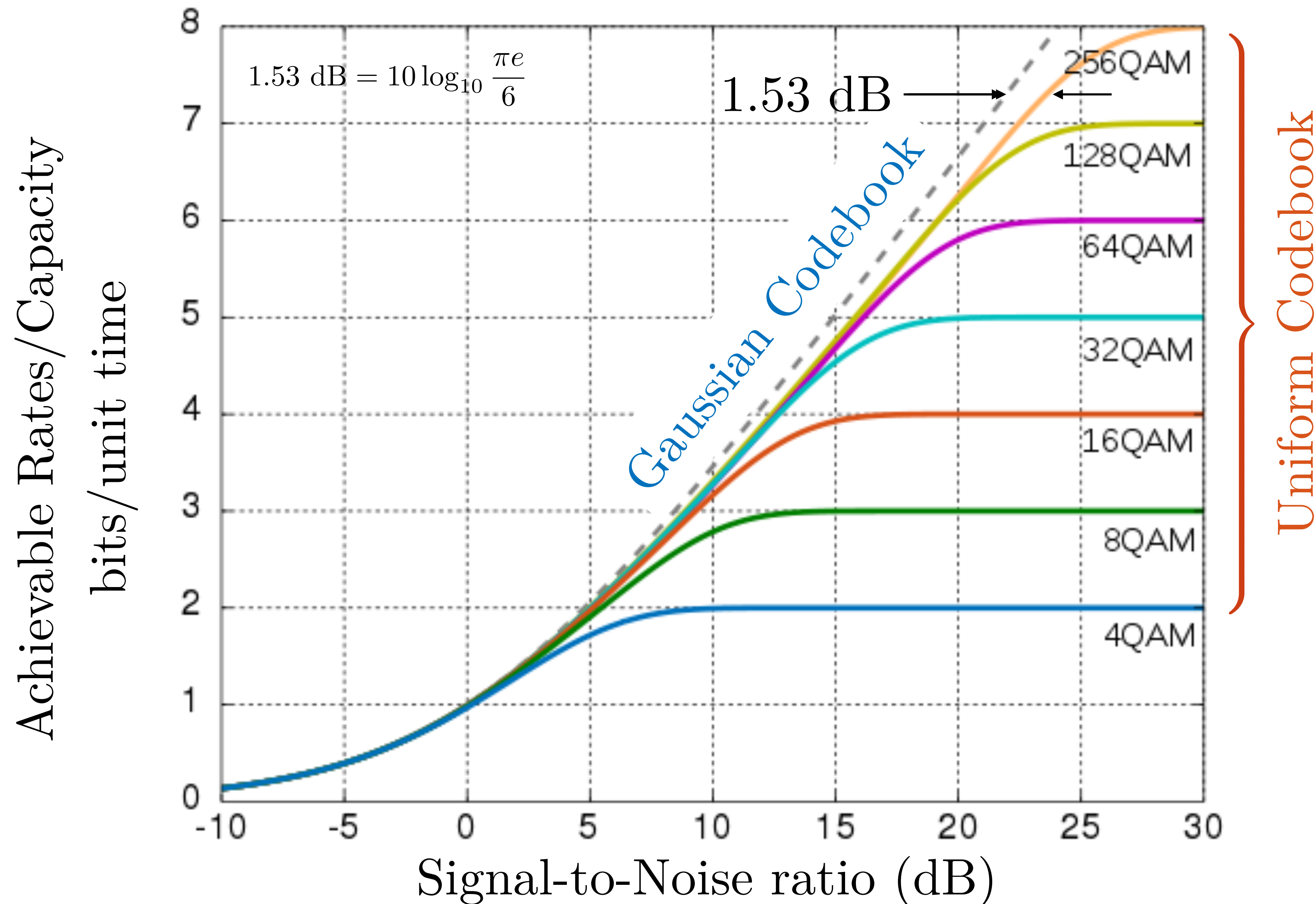
Claude Shannon  
father of information theory

Gaussian codebook maximizes capacity, uniform codebook (QAM) cannot





# Gaussian Codebook vs QAM (Uniform)



At high SNR, high rates, using a Gaussian codebook (sphere-like) gain 1.53 dB

No special benefit to using Gaussian codebook at low rates/low SNR

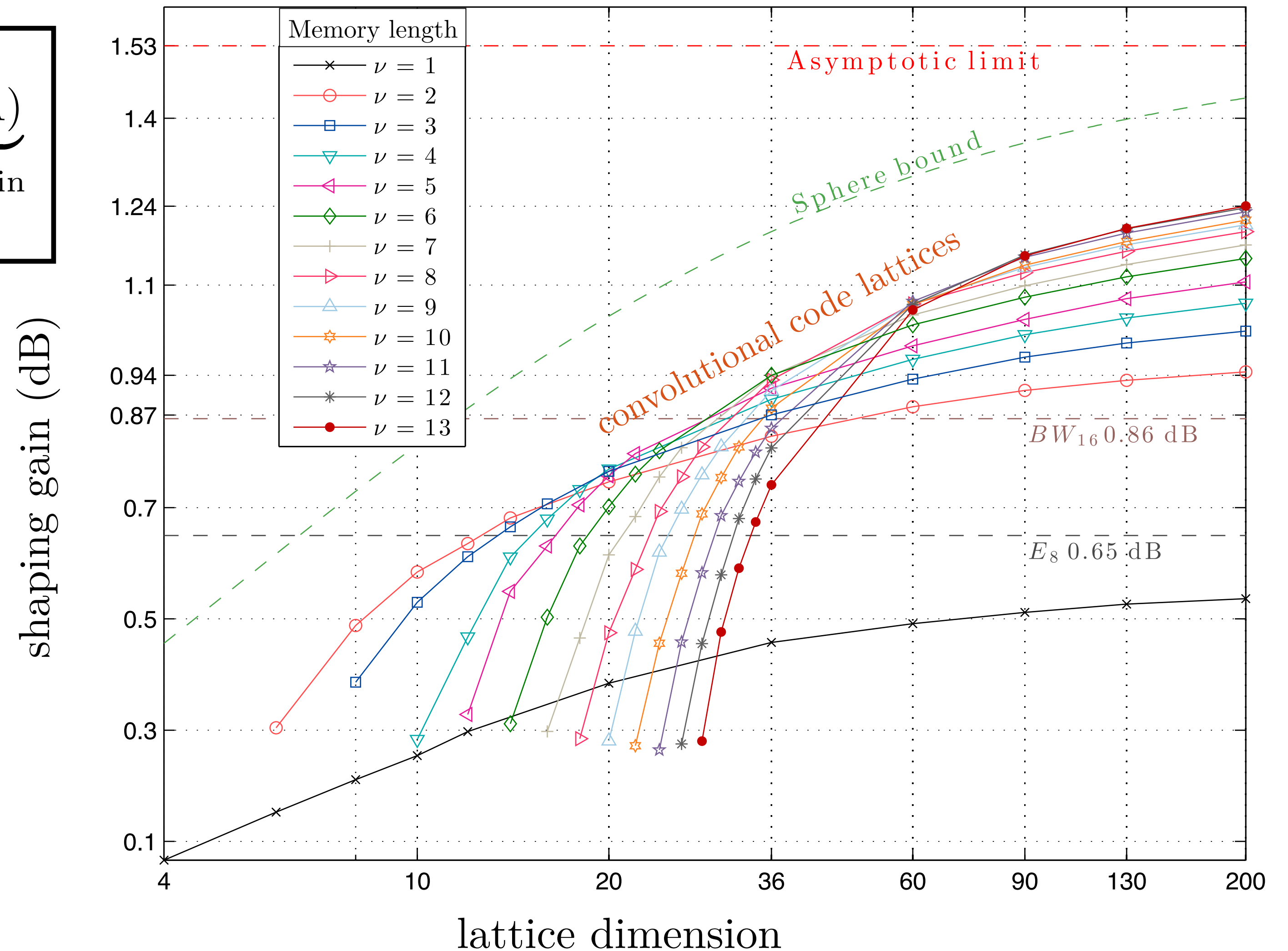
# Shaping Gain is Reduction of Transmit Power

$$\text{Average Power} \approx \underbrace{\frac{\int_B ||\mathbf{x}||^2 d\mathbf{x}}{nV(B)^{\frac{2}{n}+1}}}_{\text{shaping gain } G(B)} \cdot \underbrace{M^n V(\Lambda)}_{\text{coding gain}}$$

A spherical codebook has a Gaussian input distribution, as  $n$  to infinity.

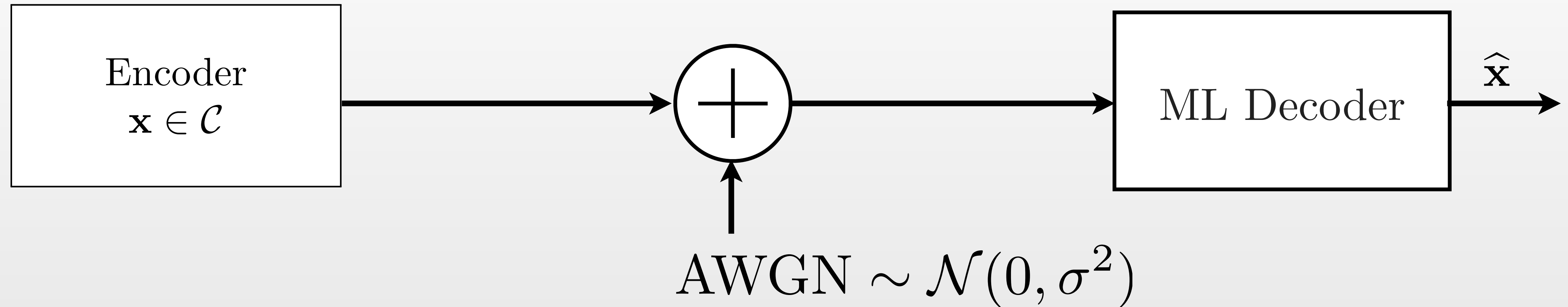
The shaping gain of various lattices is shown at the left

Proposed convolutional code lattices have excellent performance-complexity tradeoff [ZK17]





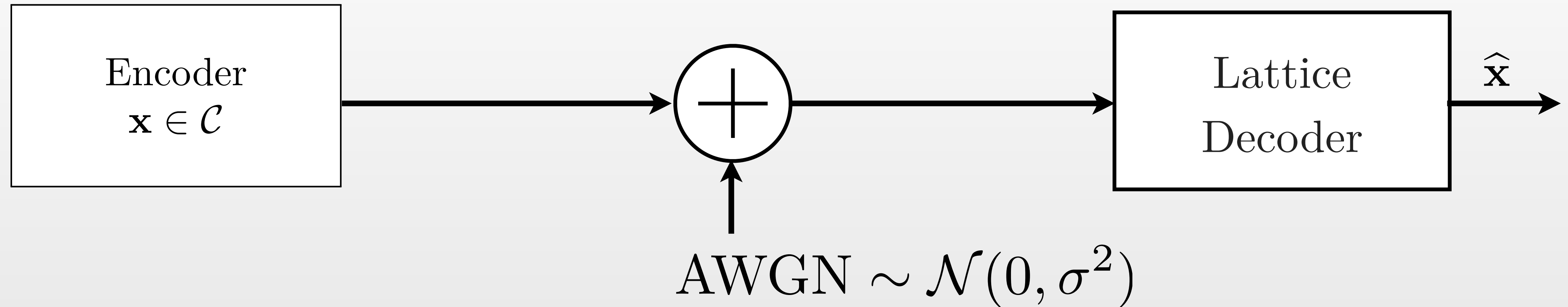
# Lattice Code ML Decoding Achieves Capacity



Lattice decoding approaches:

- Maximum likelihood decoding achieves capacity  $C = \frac{1}{2} \log(1 + P/\sigma^2)$  [de Buda. Urbanke and Rimoldi]. But this is not practical.

# Lattice Codes with Lattice Decoding

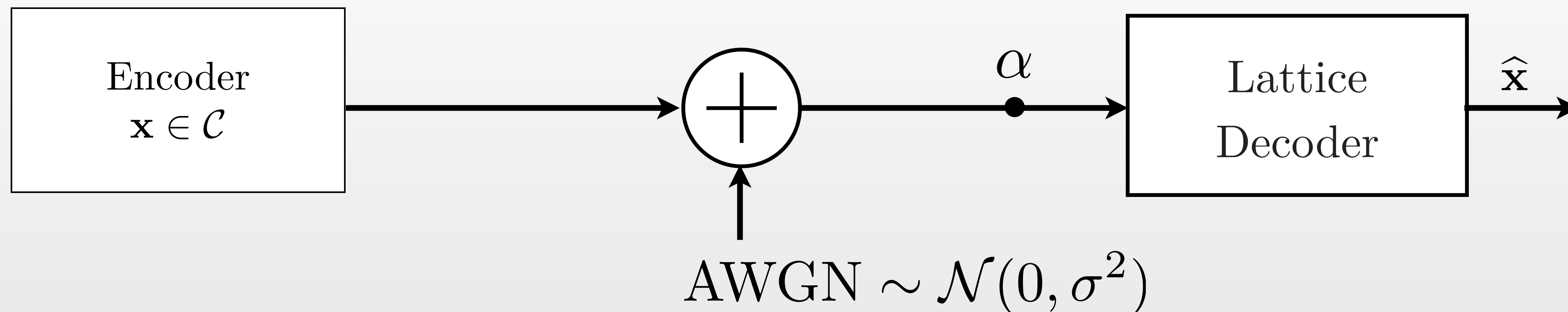


Lattice decoding approaches:

- Maximum likelihood decoding achieves capacity  $C = \frac{1}{2} \log(1 + P/\sigma^2)$  [de Buda. Urbanke and Rimoldi]. But this is not practical.
- Lattice decoding only achieves  $R < \frac{1}{2} \log(P/\sigma^2)$  [Loeliger]. Practical, but “lattice decoding” ignores the codebook boundaries.



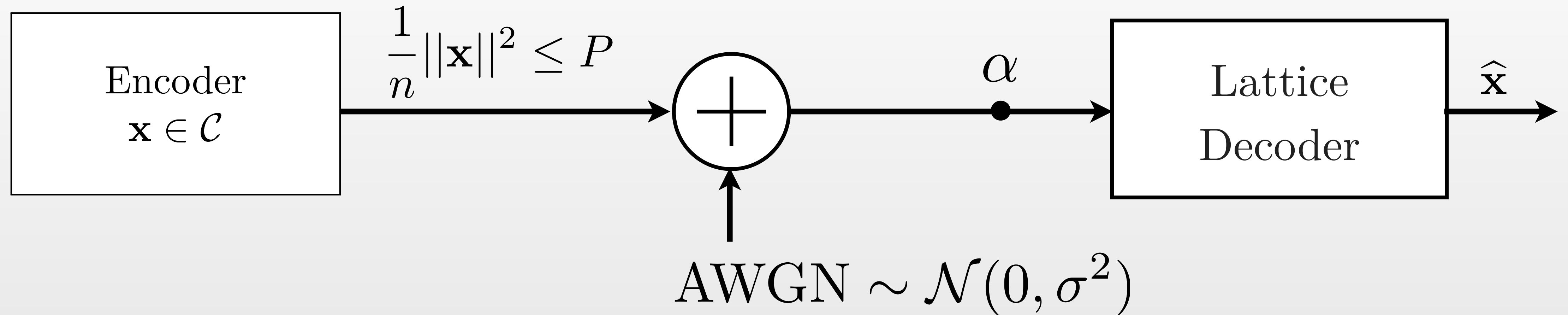
# Lattice Codes with Inflated Lattice Decoding



Lattice decoding approaches:

- Maximum likelihood decoding achieves capacity  $C = \frac{1}{2} \log(1 + P/\sigma^2)$  [de Buda, Urbanke and Rimoldi]. But this is not practical.
- Lattice decoding only achieves  $R < \frac{1}{2} \log(P/\sigma^2)$  [Loeliger]. Practical, but “lattice decoding” ignores the codebook boundaries.
- Lattice decoding with lattice inflation achieves  $C = \frac{1}{2} \log(1 + P/\sigma^2)$  [Erez and Zamir] Amazing!

# Decoding Nested Lattice Codes



$$\alpha = \frac{P}{P + \sigma^2}$$

MMSE coefficient

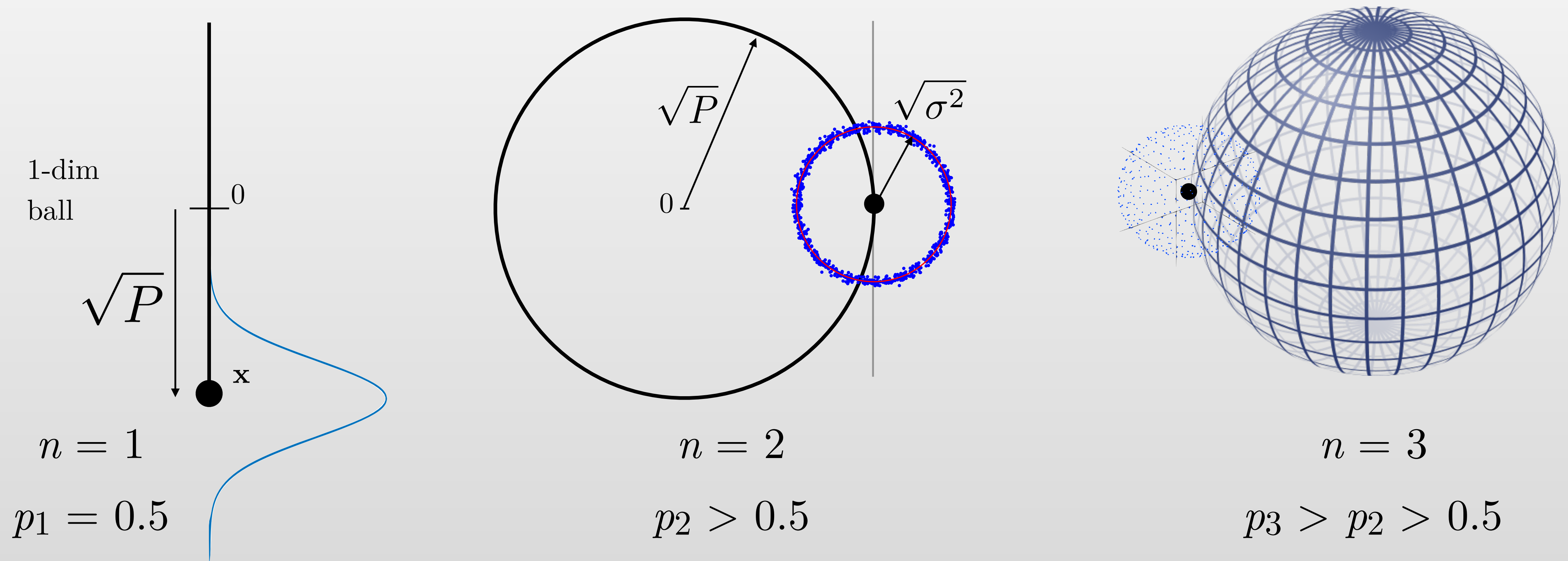
“inflates” lattice by  $\alpha^{-1}$



# Intuition for Lattice Inflation

Assume codeword  $\mathbf{c}$  is on the surface of  $n$ -ball. Noise is added to get  $\mathbf{y}$

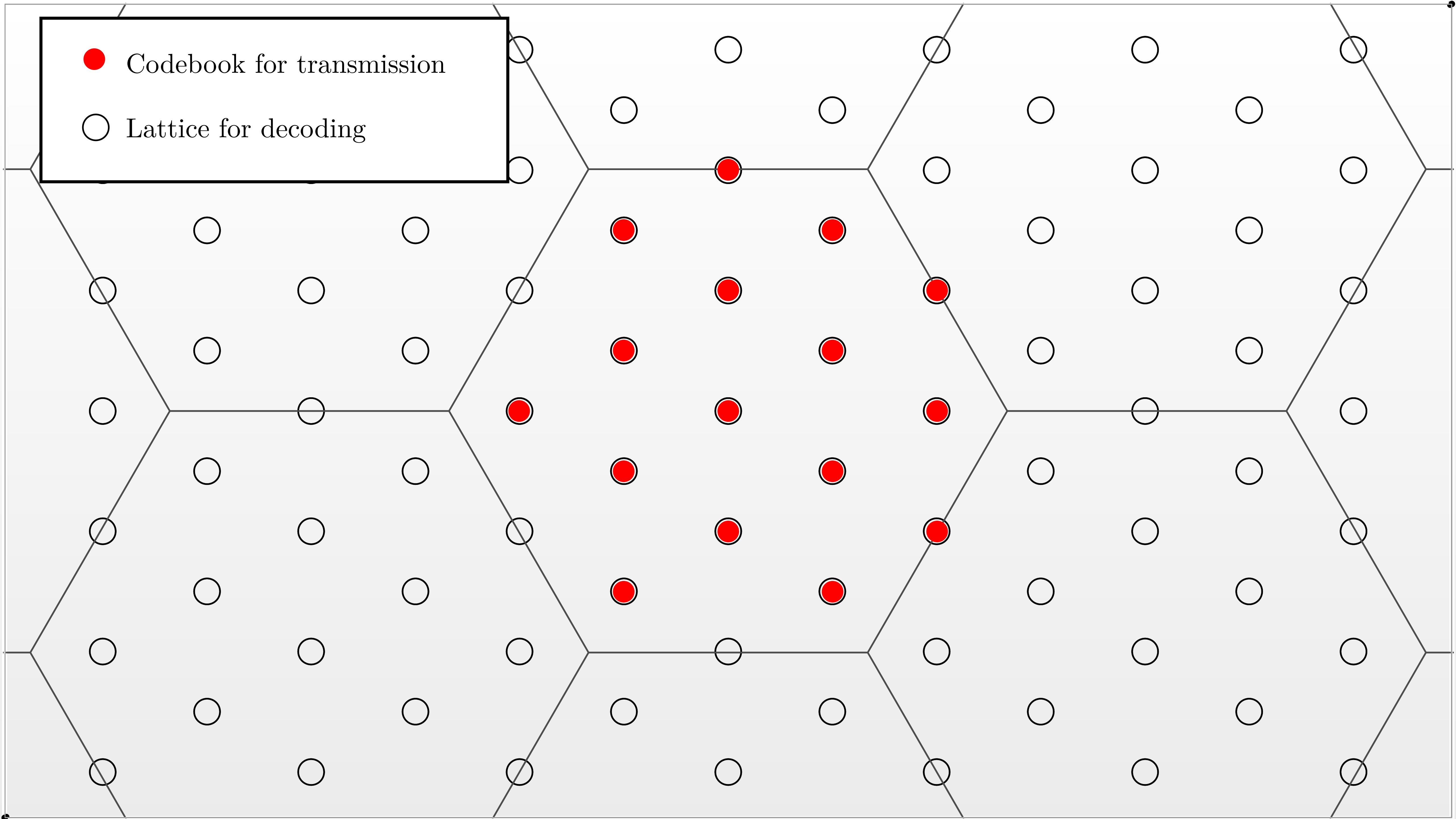
What is the probability  $p_n$  that  $\mathbf{y}$  is outside of the ball?



As  $n \rightarrow \infty$  the noise tends to be outside of the ball

● Codebook for transmission

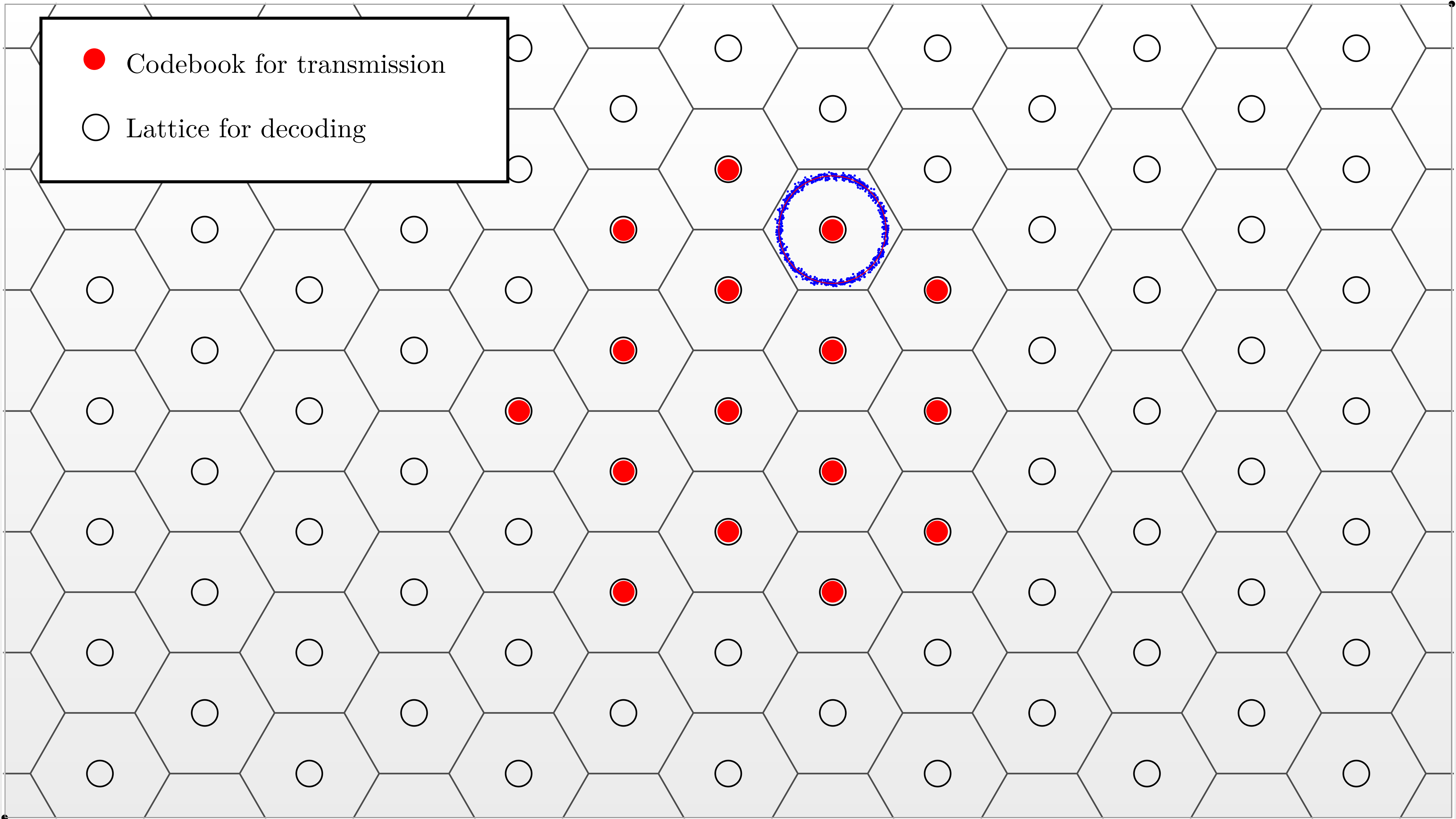
○ Lattice for decoding

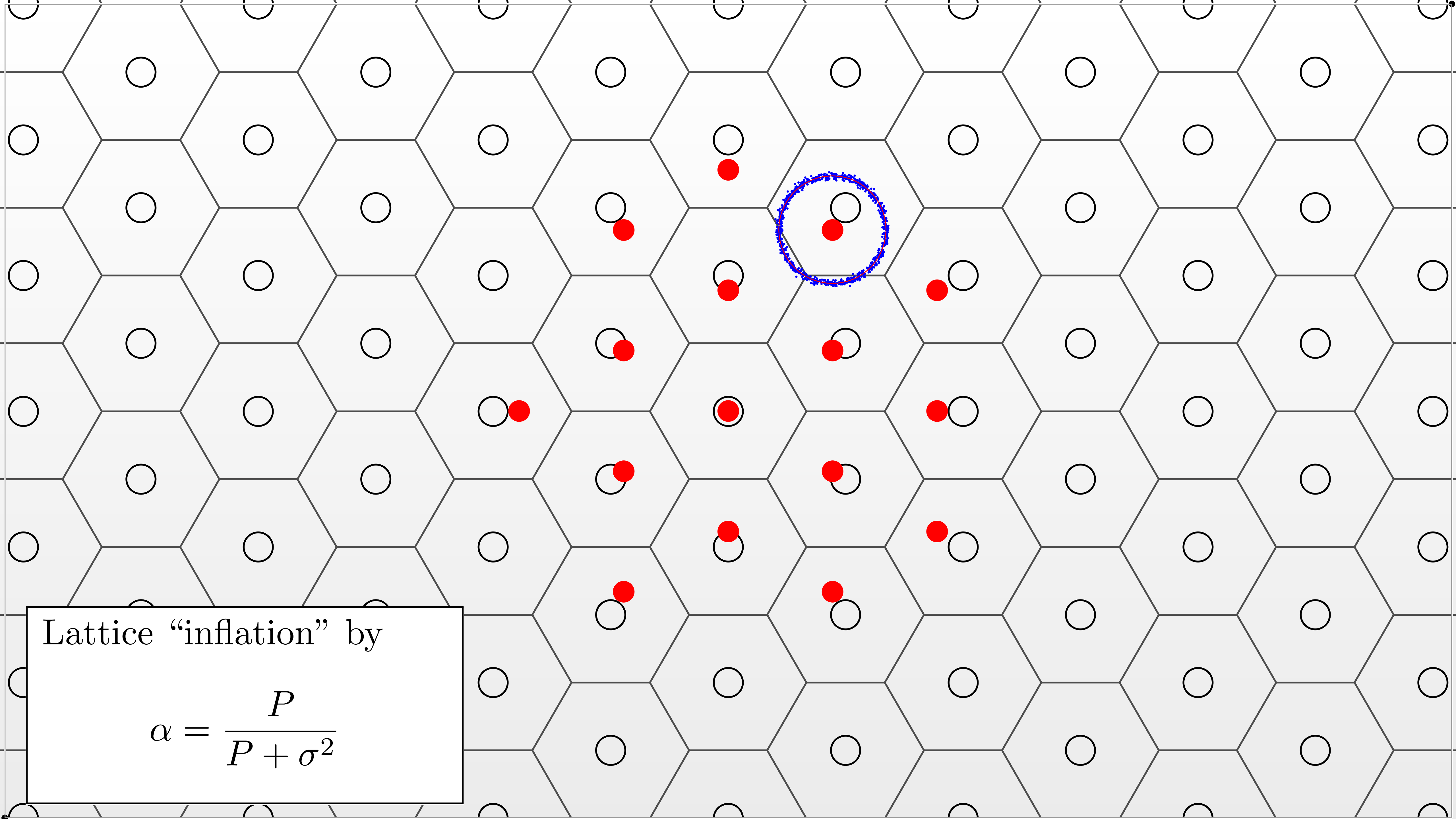




● Codebook for transmission

○ Lattice for decoding



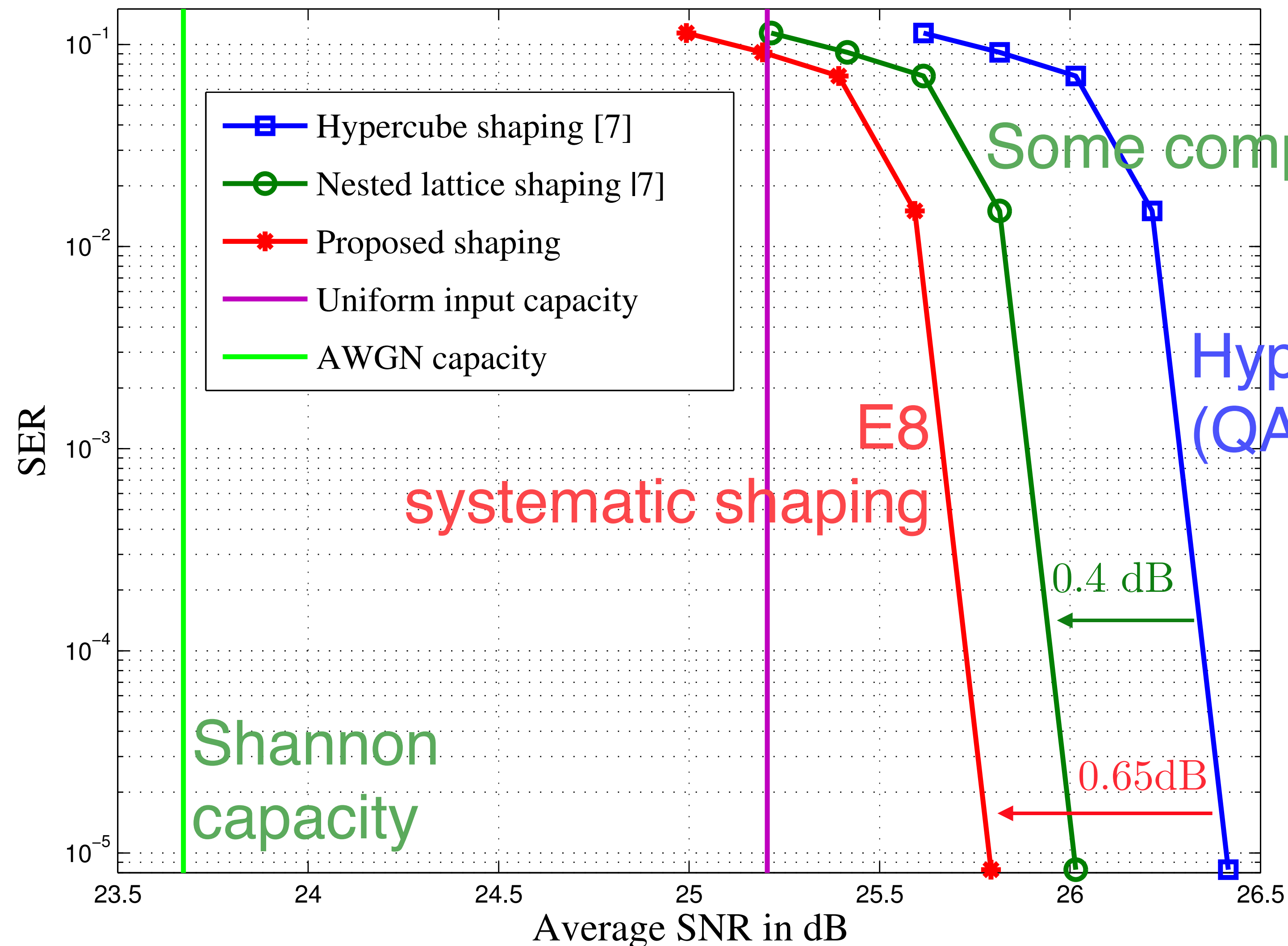


Lattice “inflation” by

$$\alpha = \frac{P}{P + \sigma^2}$$



# Shaping LDLC using E8 Lattice



Some competing algorithm

Hypercube shaping  
(QAM-like constellation)

E8  
systematic shaping

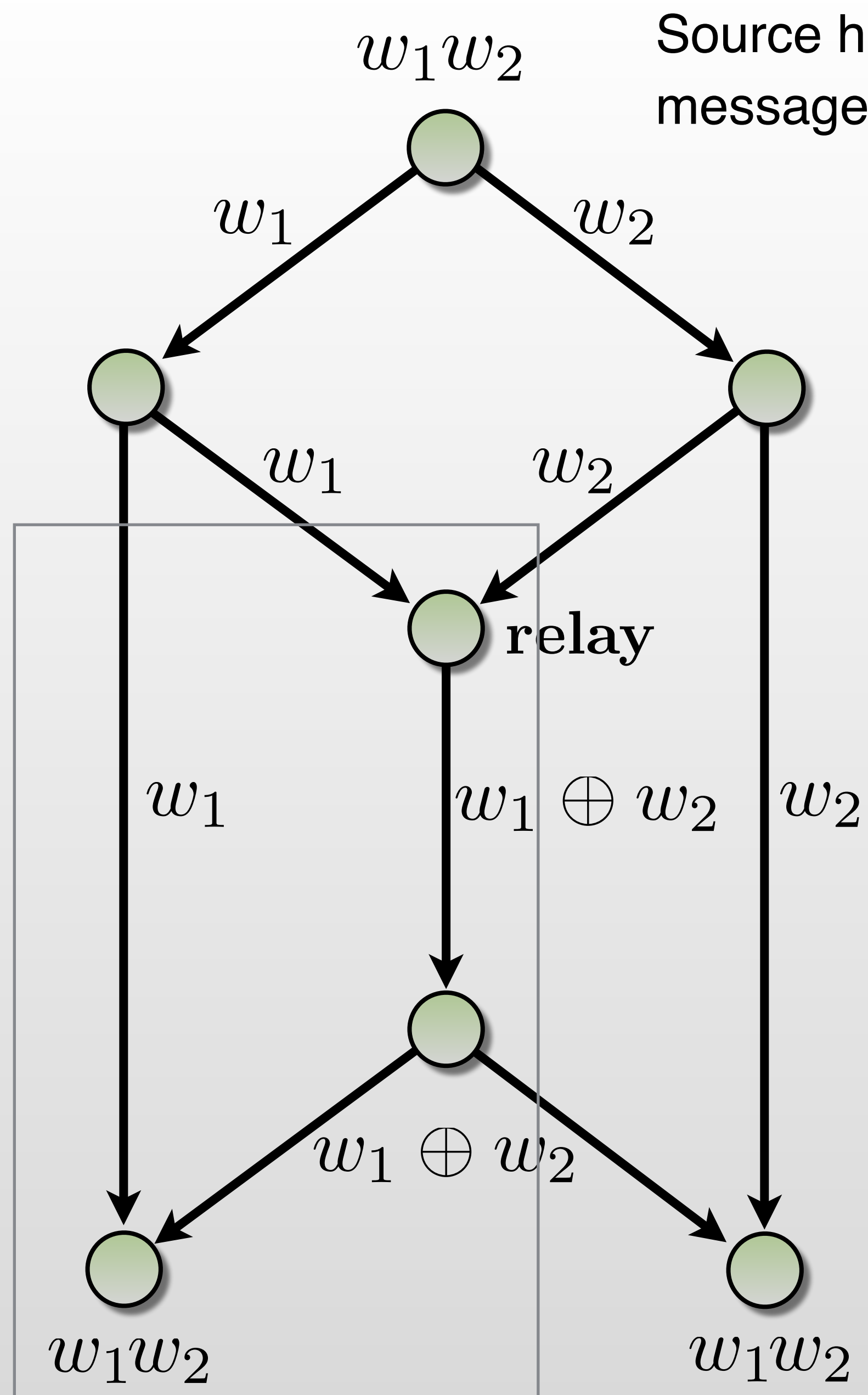
0.4 dB

0.65 dB

Shannon  
capacity

Reduction in transmit power  
by 0.65 dB. More reduction by  
using more powerful lattices.

# Routing vs. Network Coding



Source has  
messages  $w_1 w_2 w_3$

Capacity: max rate from source to destination

Routing

- Internal nodes only forward one incoming packet
- Capacity =  $3/2$

Network Coding

- Internal nodes perform linear operations
- Capacity = 2

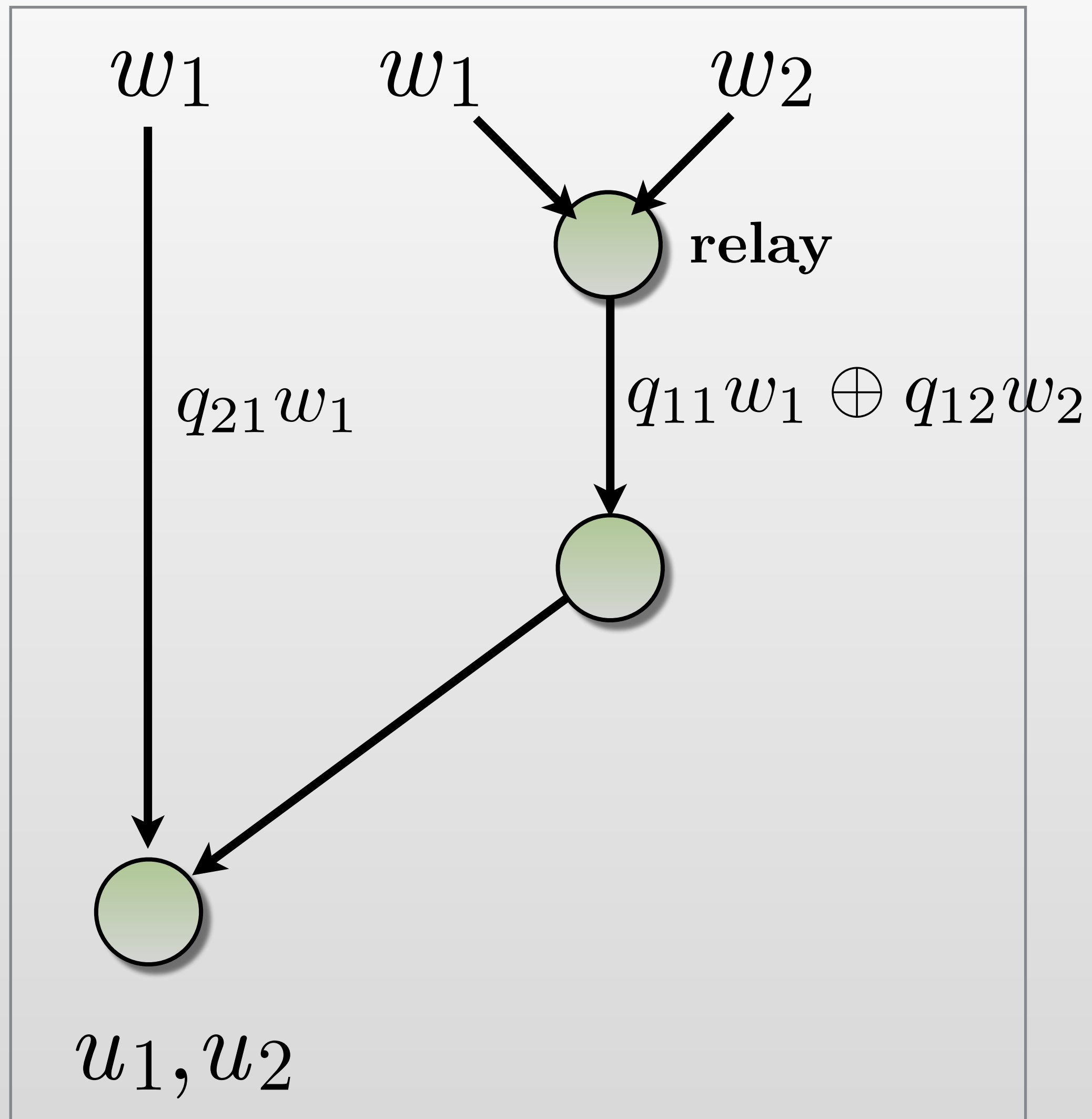
Forwarding combinations of messages can increase capacity

matrix form...

Destinations wants messages  $w_1 w_2 w_3$



# Matrix Form Recovery of Messages



$w, u, q$  in a field. Allow relay to multiply by  $q$   
 2 received messages and 2 desired messages:

$$\begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \underbrace{\begin{bmatrix} q_{11} & q_{12} \\ q_{21} & q_{22} \end{bmatrix}}_{\mathbf{Q}} \cdot \begin{bmatrix} w_1 \\ w_2 \end{bmatrix}$$

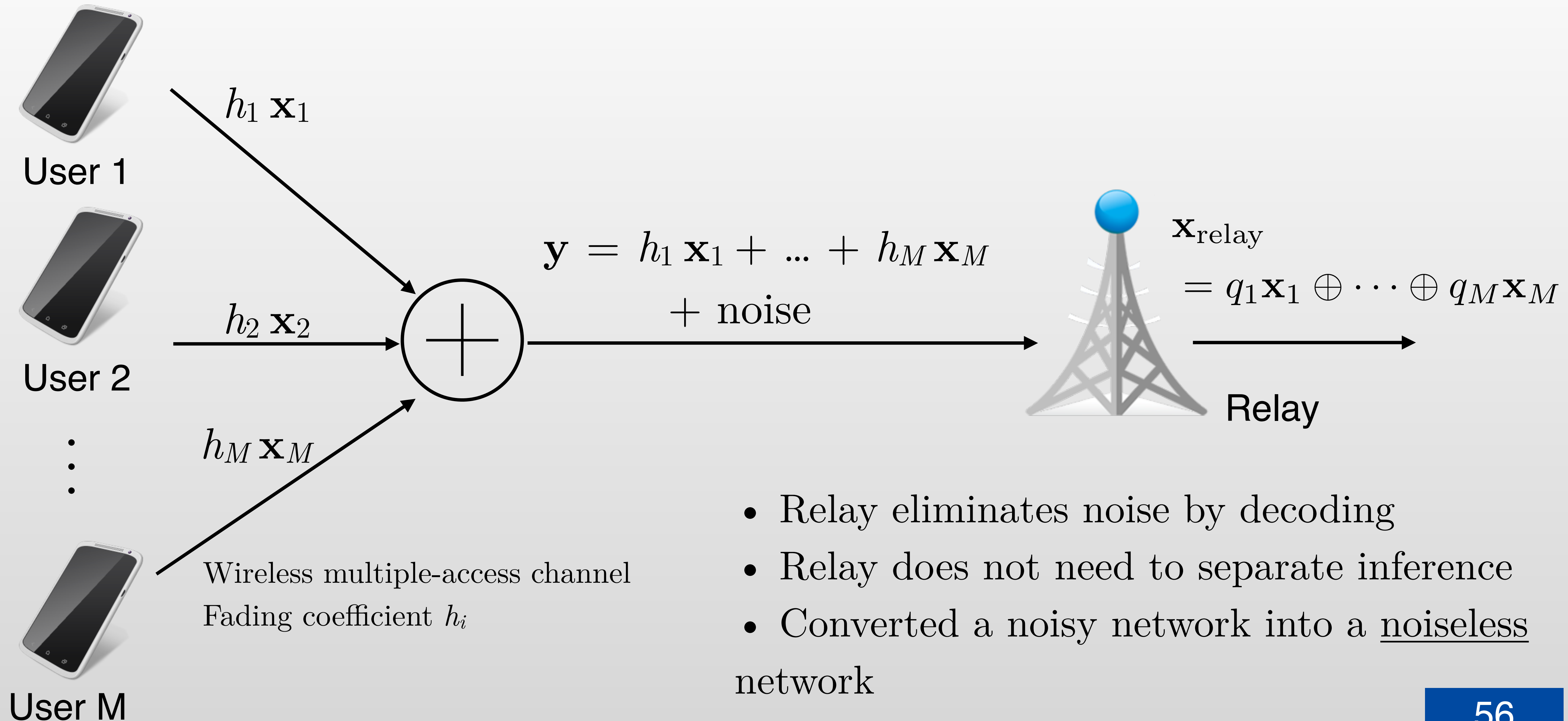
$$\begin{bmatrix} q_{11} & q_{12} \\ q_{21} & q_{22} \end{bmatrix}^{-1} \cdot \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} w_1 \\ w_2 \end{bmatrix}$$

↑ received messages
 ↑ desired messages

Destination should receive sufficient linear combinations such that  $\mathbf{Q}$  is invertible

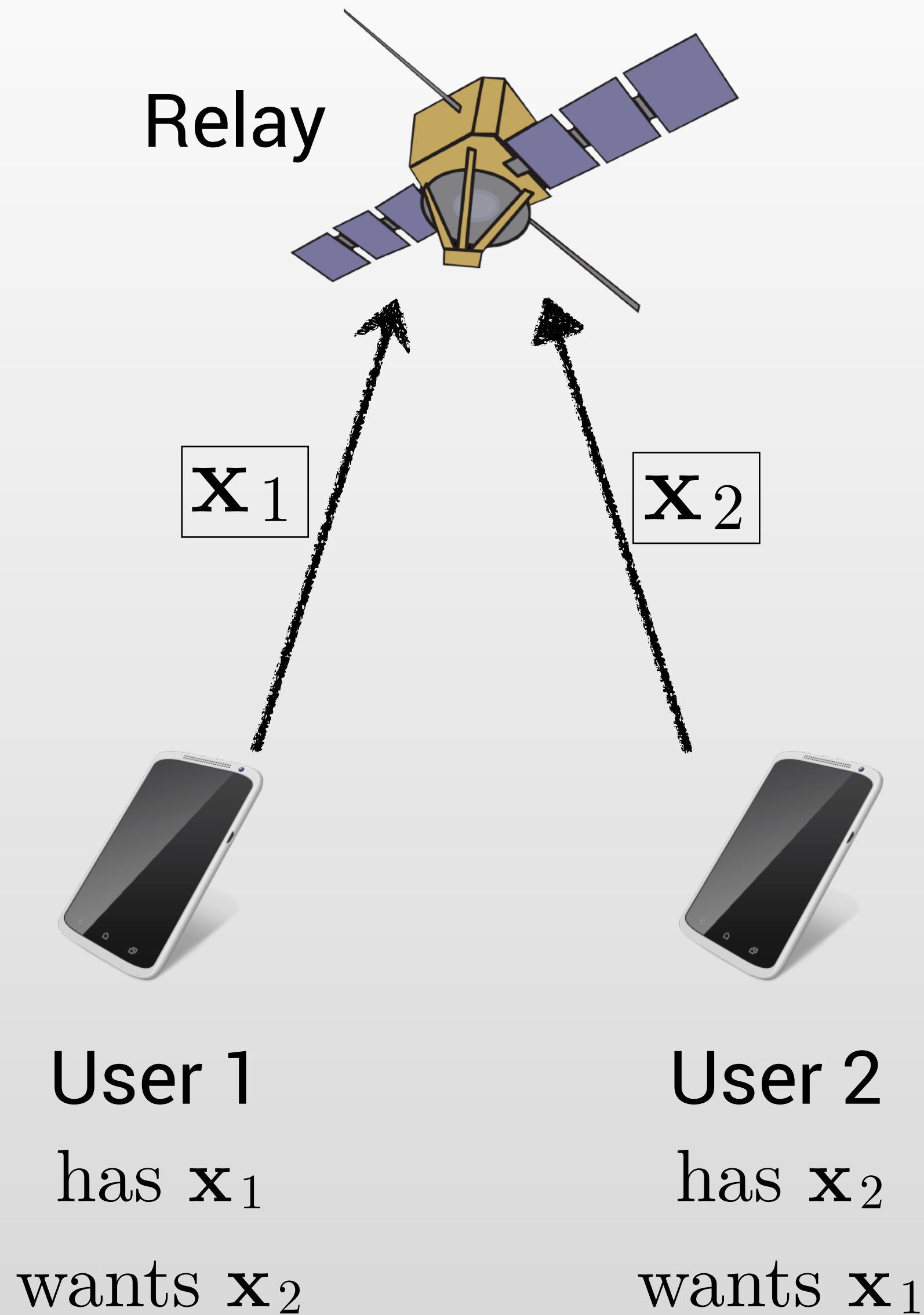
# PLNC = Physical Layer Network Coding

Addition occurs over the air



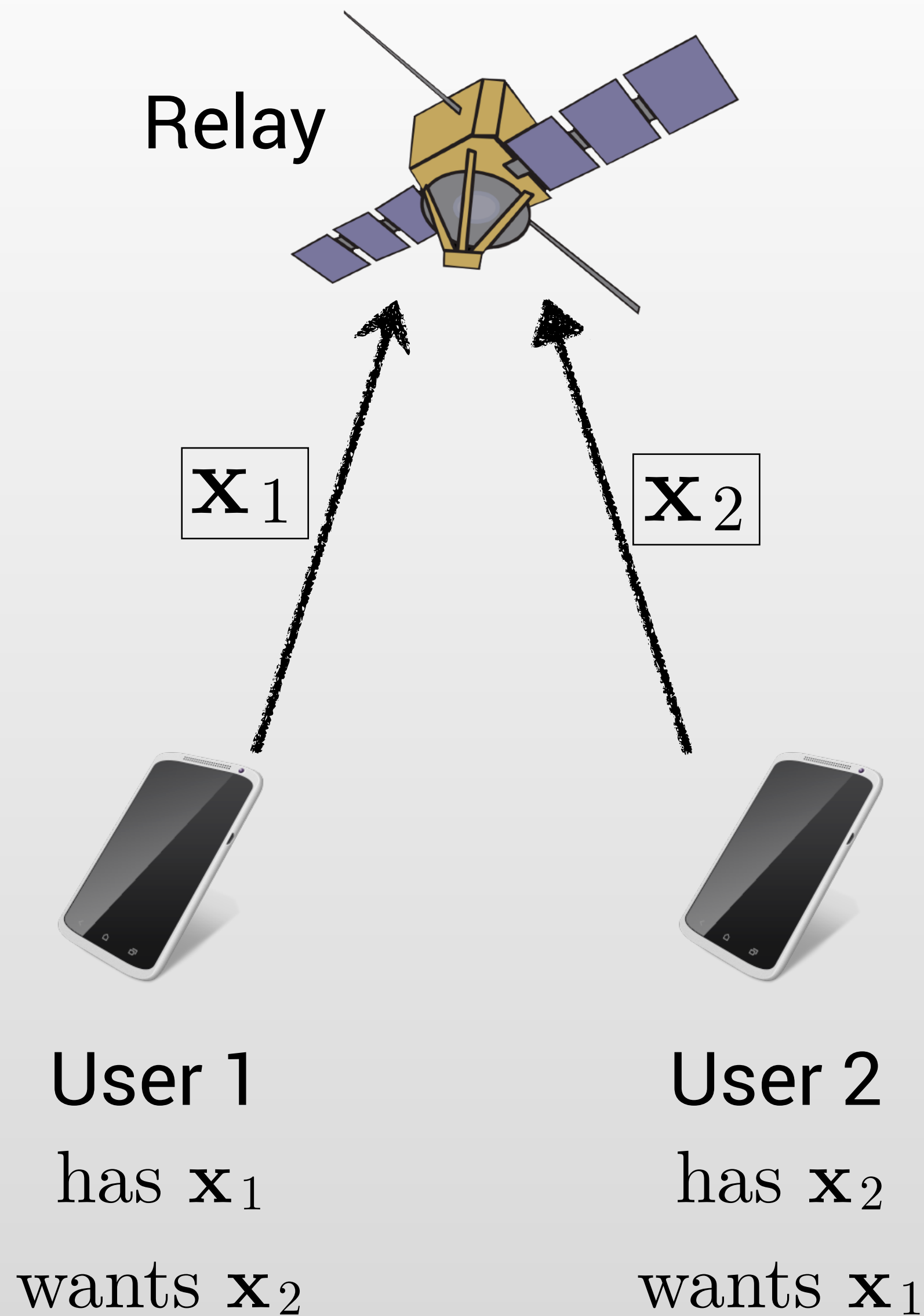


# Bidirectional Relay Channel



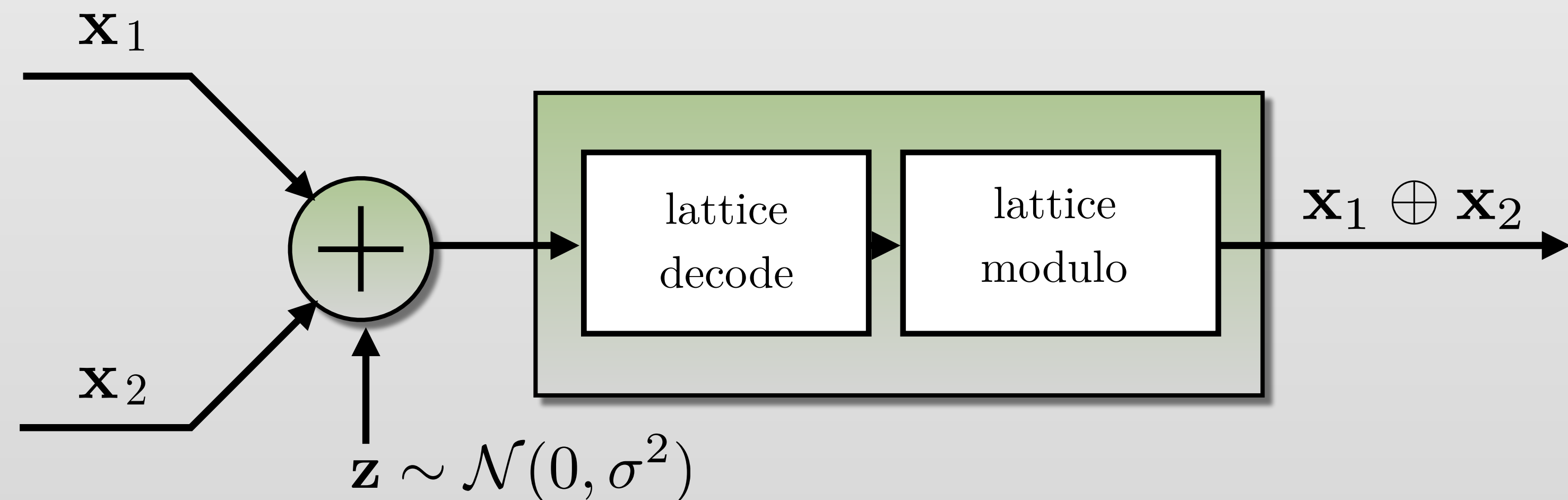
- Orthogonal: uses 4 time slots
- Network coding: uses 3 time slots
- Physical layer network coding (PLNC): 2 time slots

# Bidirectional Relay Channel



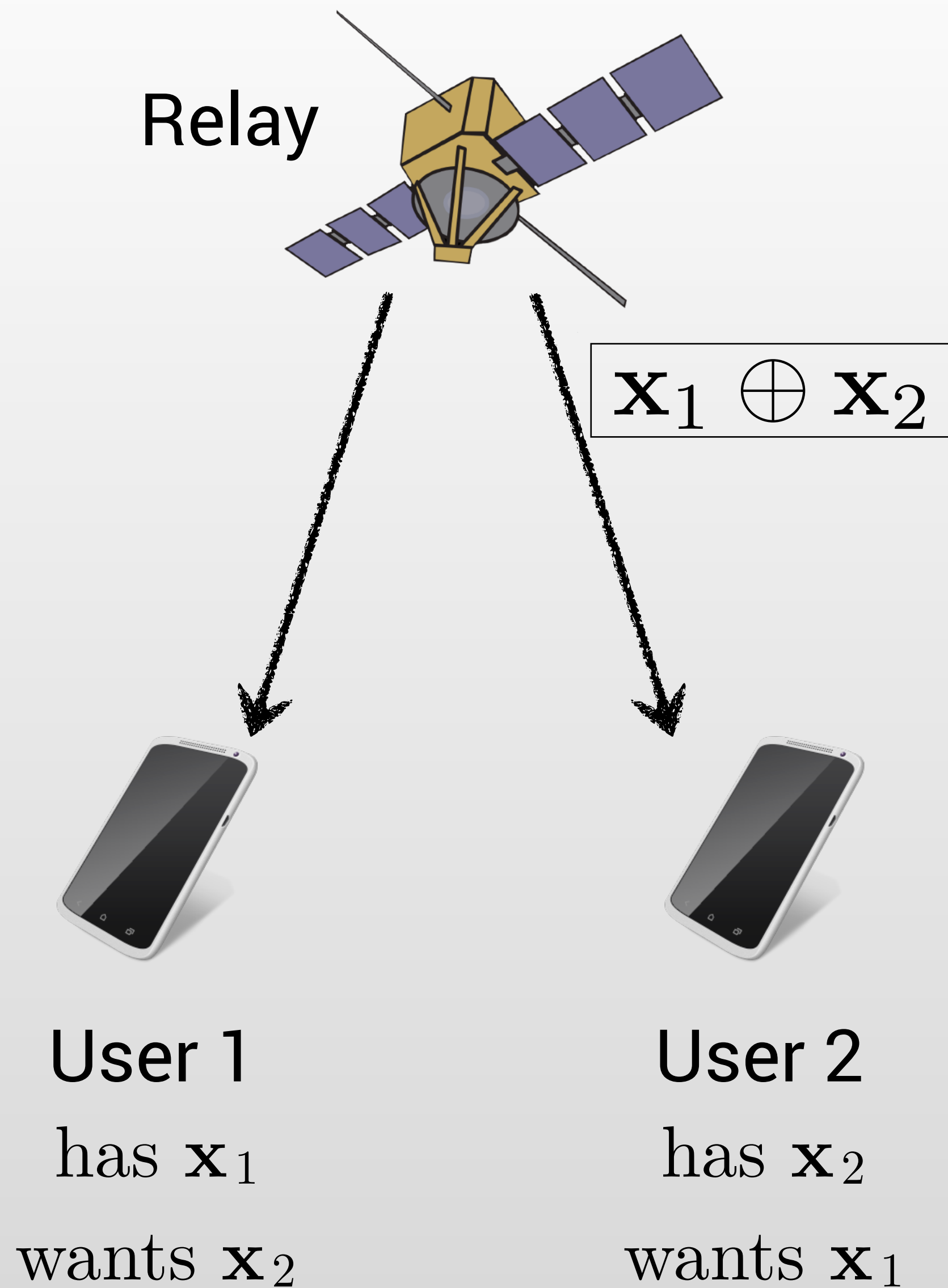
- Orthogonal: uses 4 time slots
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- Physical layer network coding (PLNC): 2 time slots

## Relay Using PLNC



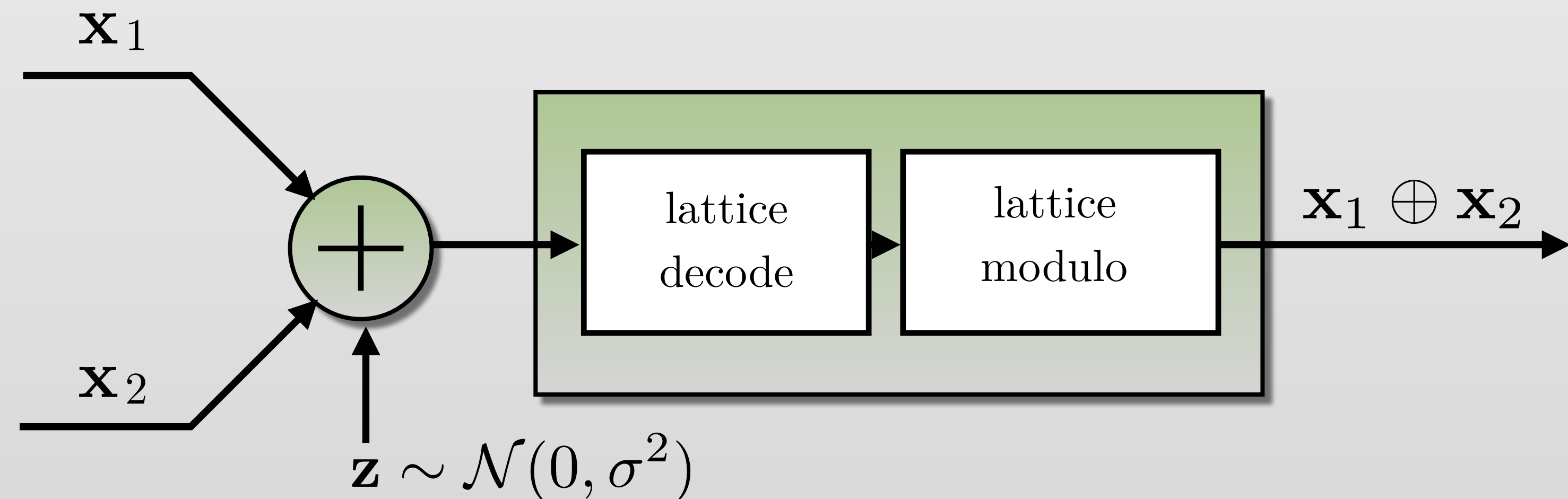


# Bidirectional Relay Channel

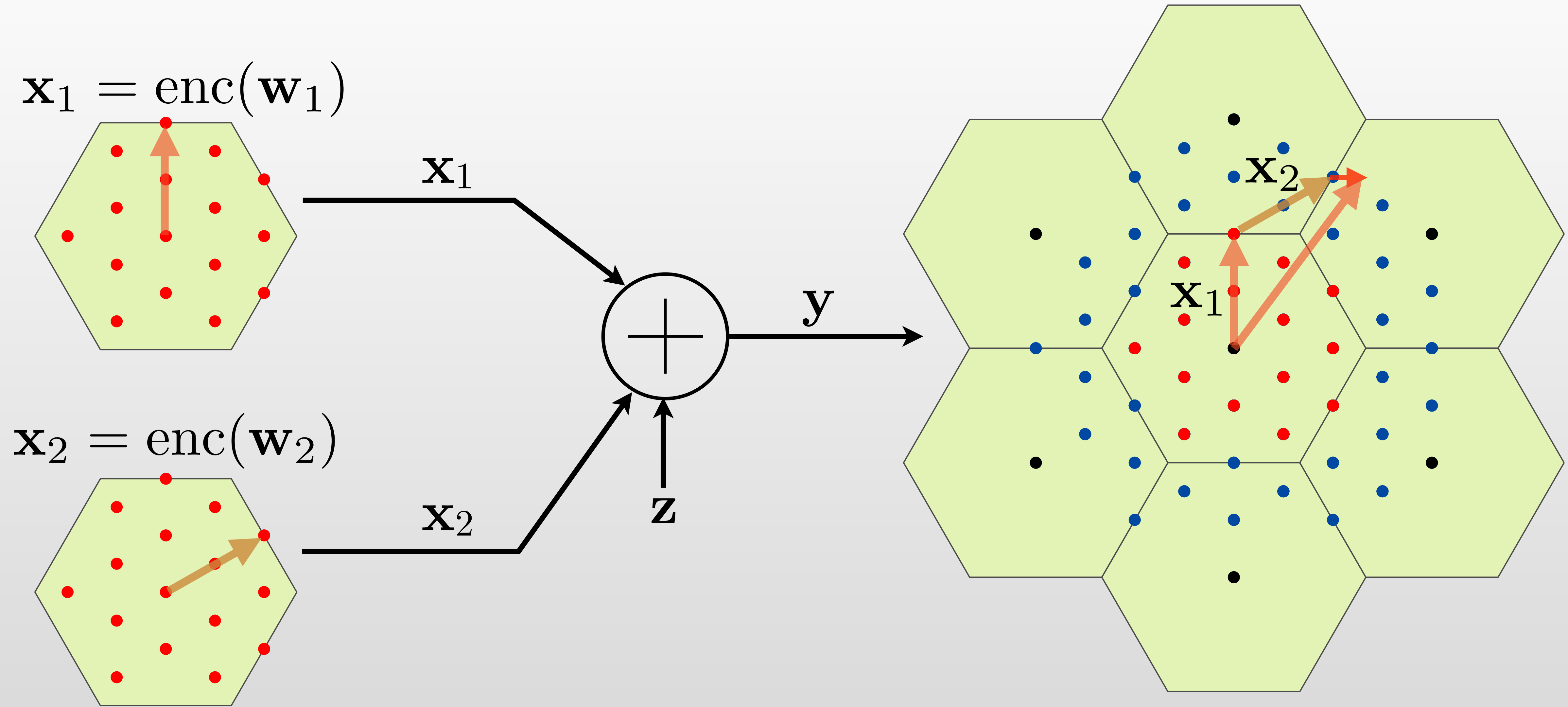


- Orthogonal: uses 4 time slots
- Network coding: uses 3 time slots
- Physical layer network coding (PLNC): 2 time slots

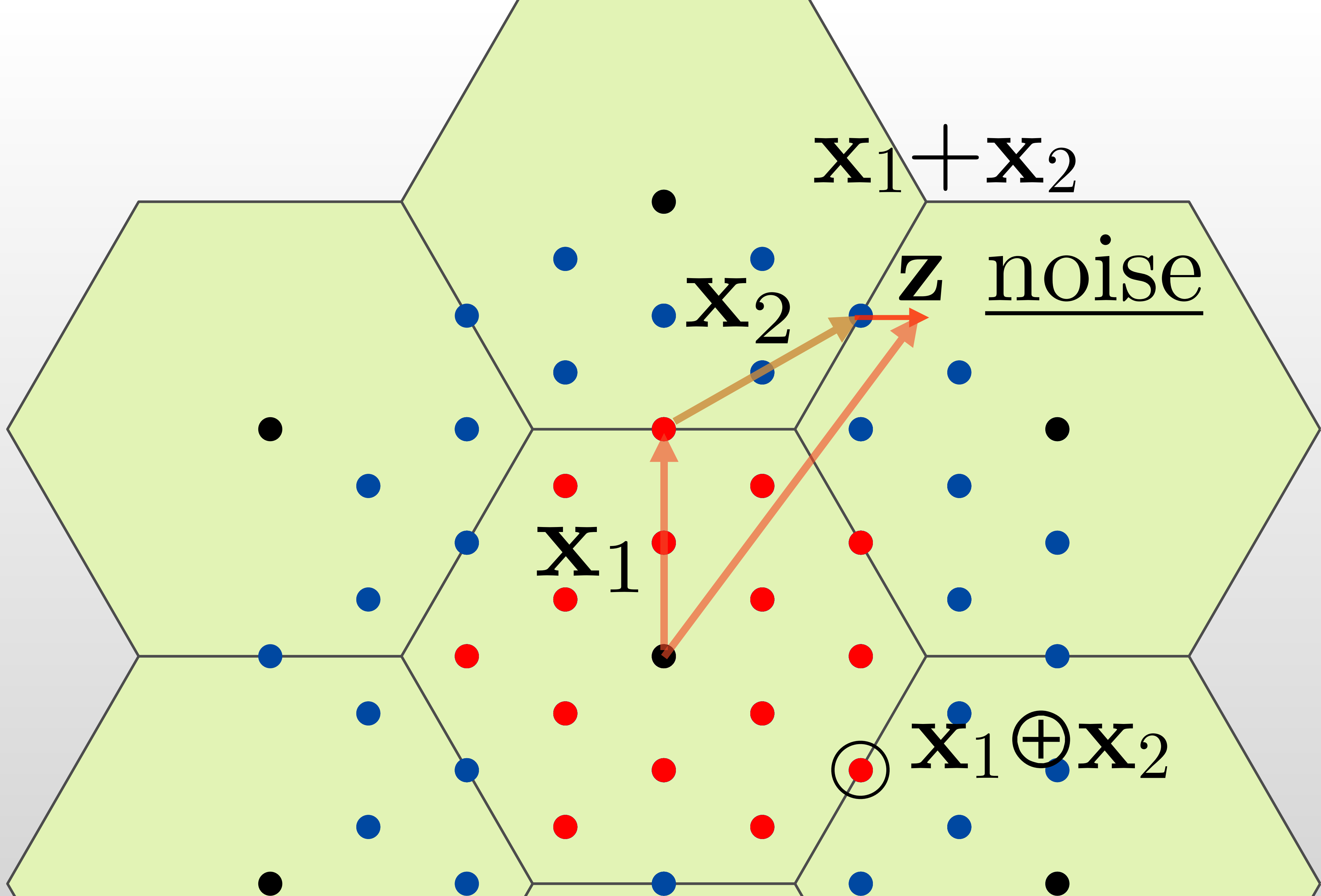
## Relay Using PLNC



# Relay Using PLNC

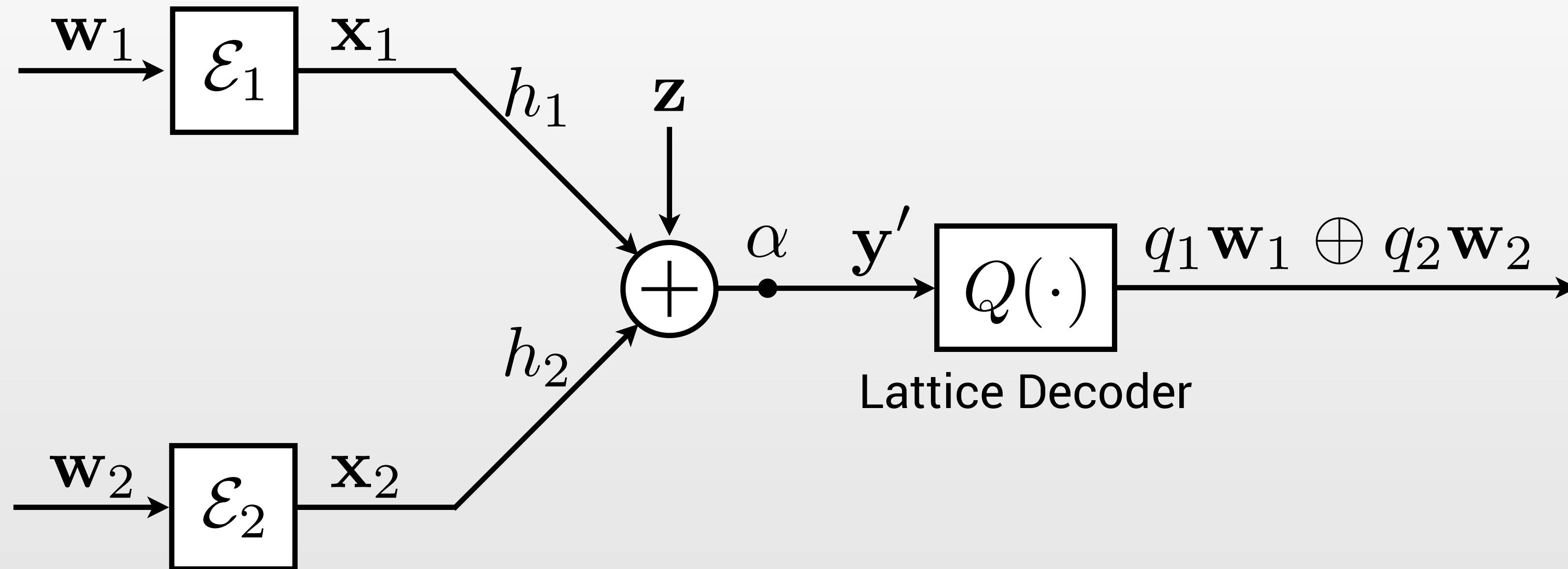






# What if channel coefficients are not integers?

## Compute-and-Forward



In practice, fading coefficients  $h$  are arbitrary values, not integers.

PLNC can still work. This is “compute and forward”

$$\mathbf{y}' = \alpha h_1 \mathbf{x}_1 + \alpha h_2 \mathbf{x}_2 + \alpha \mathbf{z} \quad \text{fading coefficients } h \in \mathbb{R}$$

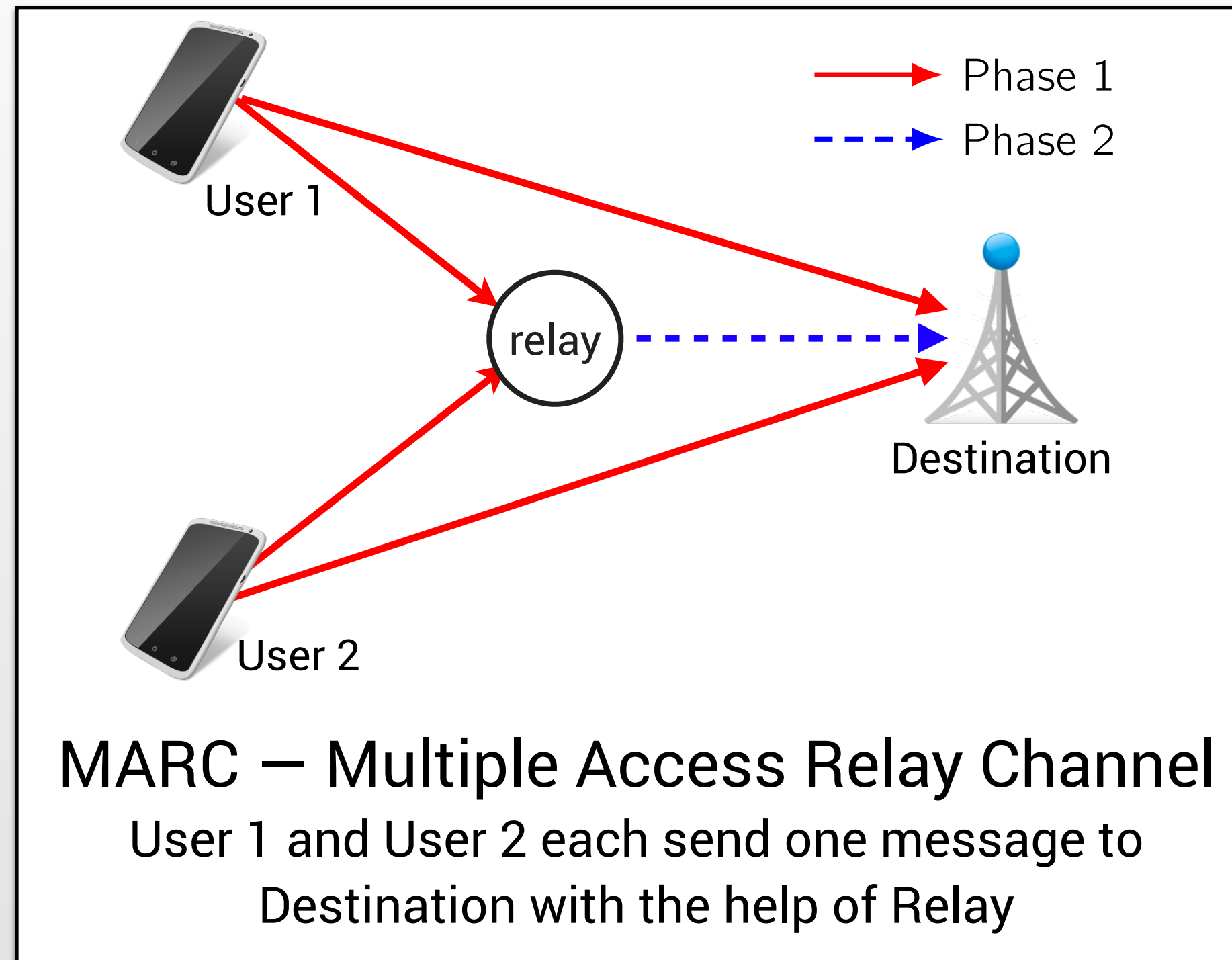
$$\mathbf{y}' = a_1 \mathbf{x}_1 + a_2 \mathbf{x}_2 + \mathbf{z}_{\text{eff}} \quad \text{integer approximation } a \in \mathbb{Z}$$

$$Q(\mathbf{y}') = q_1 \mathbf{w}_1 \oplus q_2 \mathbf{w}_2 \quad \text{conversion to finite field } q, \mathbf{w} \in \mathbb{F}^n$$

Finding  $a_1, a_2$  is an optimization problem



# Compute-Forward for Multiple Access Relay Channel



## Naive application of CF to MARC

Relay and Destination independently choose coefficient vectors  
destination gets two independent vectors

$$\begin{bmatrix} q_{11} & q_{12} \\ q_{21} & q_{22} \end{bmatrix}^{-1} \cdot \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} w_1 \\ w_2 \end{bmatrix}$$

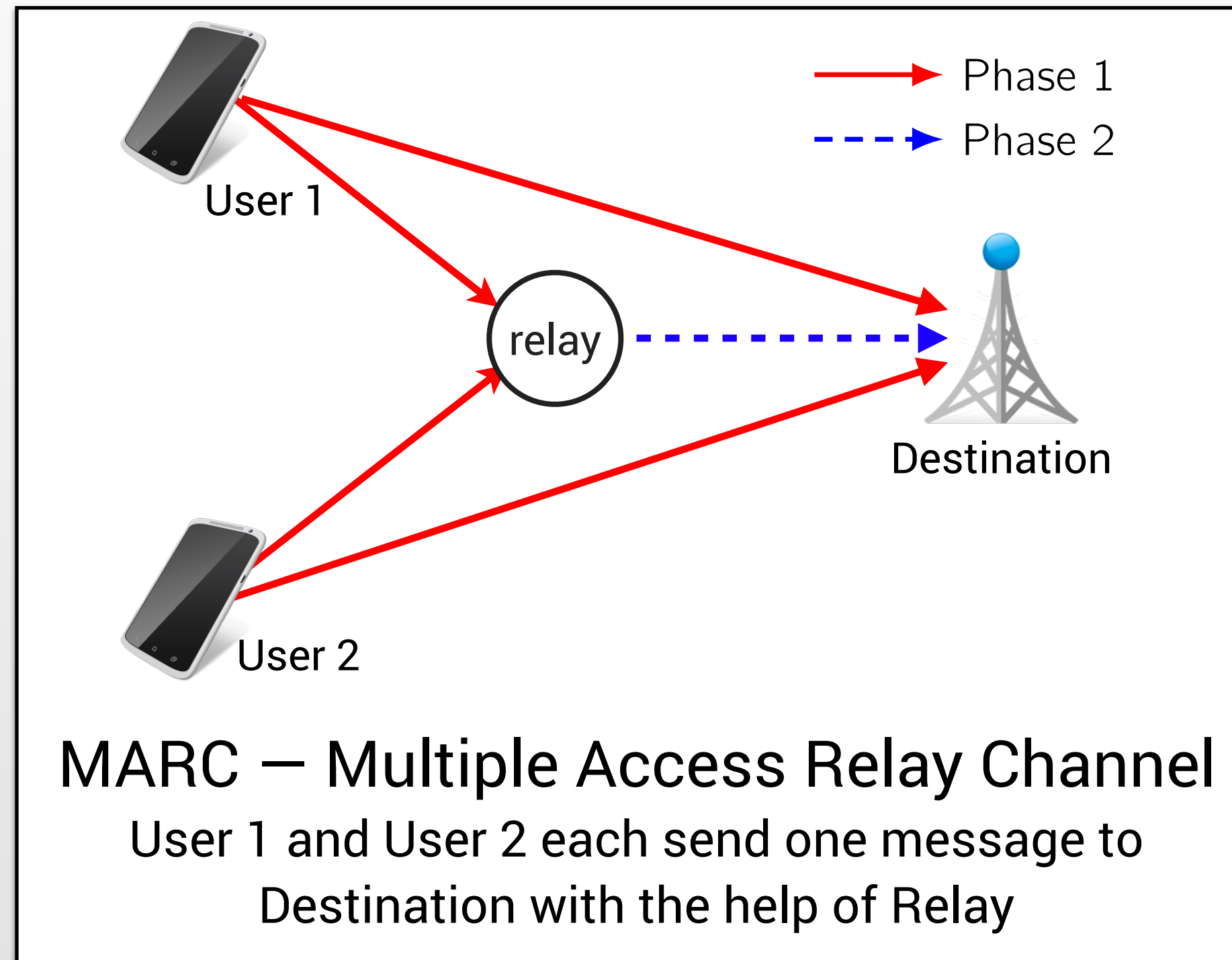
But  $\mathbf{Q}$  may not be invertible, with significant probability.

## Full cooperation protocol

Destination sends  $\mathbf{q}$  vector to relay

Relay selects linearly independent  $q$ , to guarantee  $\mathbf{Q}$  is full rank

# Compute-Forward for Multiple Access Relay Channel



A list Ficke-Pohst algorithm finds  $L$  best rates:

$$R(\mathbf{a}^*) \geq R(\mathbf{a}_2) \geq \dots \geq R(\mathbf{a}_L)$$

and the corresponding coefficient vectors:

$$\mathbf{a}^*, \mathbf{a}_2, \dots, \mathbf{a}_L$$

The destination attempts to decode using the two best  $\mathbf{a}$ 's

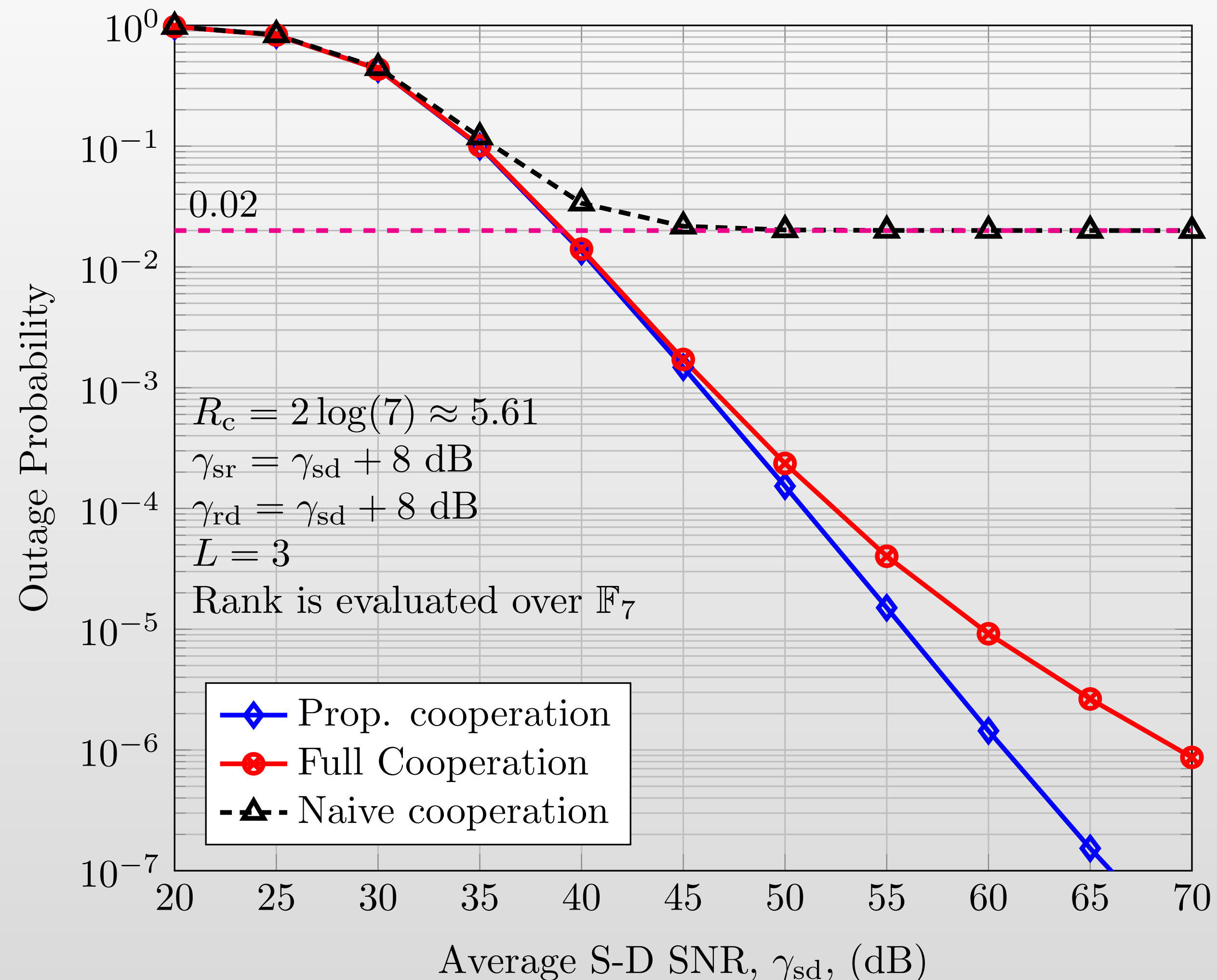
## Proposal: Form Multiple Linear Combinations at Destination

1. Destination attempts to decode both  $u$  and  $u$  by forming linearly independent combinations. Relay does nothing.
2. If this fails, destination sends  $\mathbf{a}^*$  to relay. Relay chooses its best linearly independent combination. Using this, data is transmitted from relay to destination.

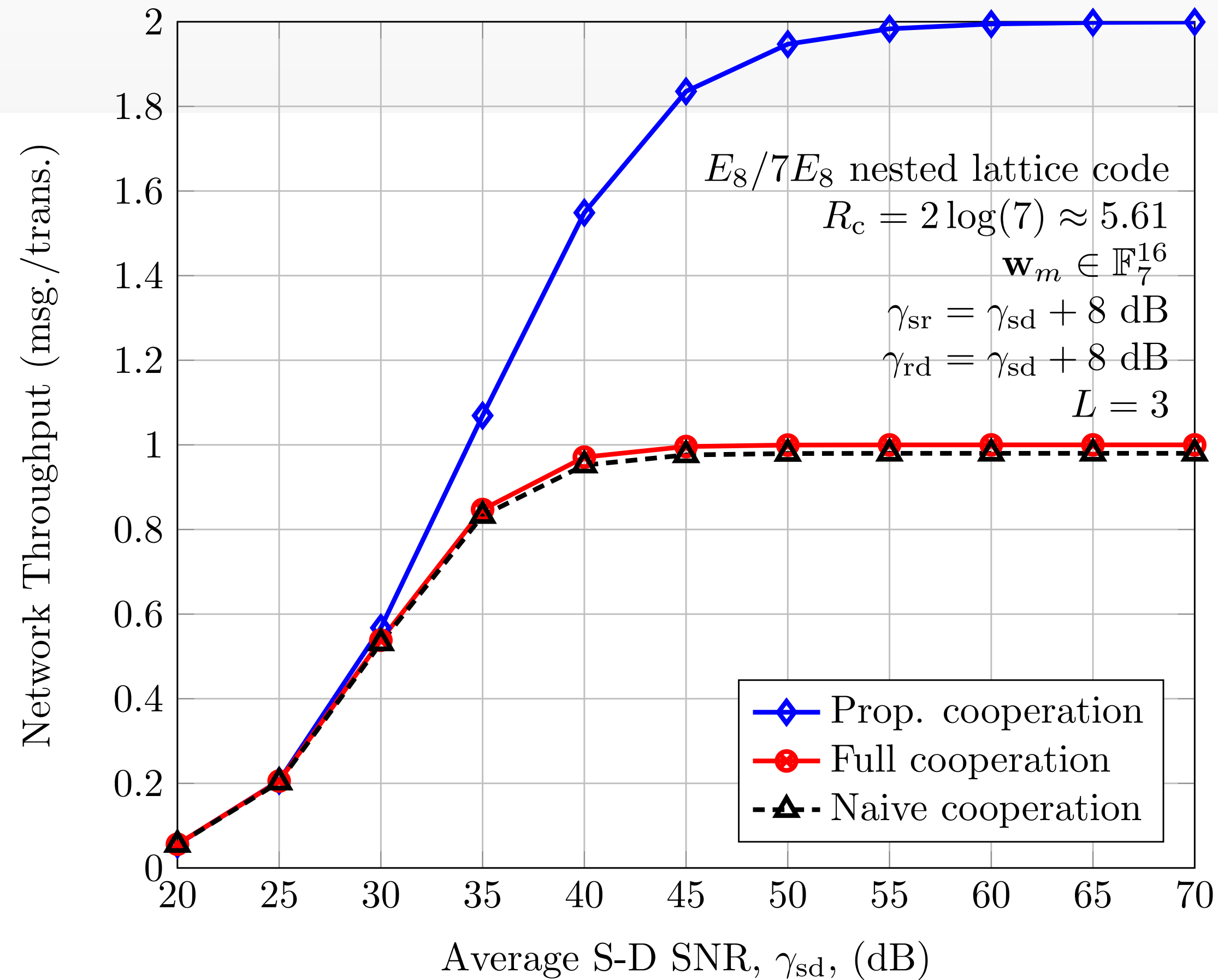


# Proposed Method has Lower Outage Probability

- Maximum diversity order of 2  
(competing systems have diversity order less than 2)



# 100% increase in throughput



100% improvement  
network throughput

- Network throughput increases 100% [HK18]



# Conclusion

**Central question: How might lattices effectively be used in wireless communication systems?**

Lattices with practical encoding and decoding are needed — Construction D' using QC-LDPC codes is a strong candidate

Lattices can provide shaping gain which is difficult otherwise — Convolutional code lattices provide  $> 1.0$  dB of shaping gain

Physical layer network coding provides significant throughput benefit — lattices enable PLNC