# Algebraic reduction for low-complexity lattice decoding

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LATTICE CODING AND CRYPTO MEETING IMPERIAL COLLEGE LONDON - SEPTEMBER 24, 2018

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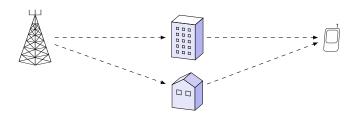
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  - from number fields through the canonical embedding
  - from division algebras through the left regular representation
- Question: can you exploit this extra structure to improve lattice reduction?
  - in coding theory: for decoding
  - in lattice-based cryptography: for attacks

#### **Outline**

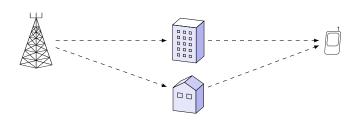
- Coding for wireless communications
  - Single antenna systems
  - MIMO systems
- 2 Decoding
- Algebraic reduction
  - Single antenna systems
  - MIMO systems

### Coding for wireless channels

- algebraic number theory is an effective tool to design codes that are full-rate, full-diversity and information-lossless
- in order to increase data rates, both the number of antennas and the size of the signal set can be increased
- this entails a high decoding complexity which is a challenge for practical implementation

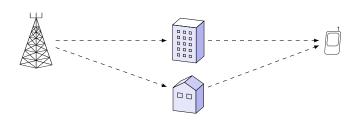


• fading channel: the signal is scattered by many obstacles and propagates through multiple paths



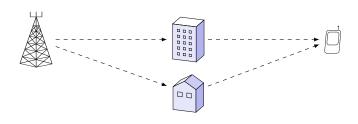
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- when the number of paths is large, fading and noise can be modelled as Gaussian random variables  $h \sim \mathcal{N}_{\mathbb{C}}(0,1), w \sim \mathcal{N}_{\mathbb{C}}(0,\sigma^2)$ :

$$y = h \quad x + w$$
 received signal channel codeword noise



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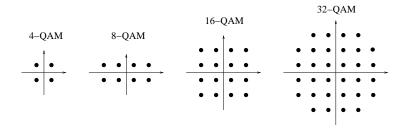


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• open loop: channel is known at the receiver, but not at the transmitter

# Digital modulation



- quadrature-amplitude modulation: a binary information vector is used to modulate an analog waveform
- ullet the set of waveforms  $s\in\mathbb{C}$  is a finite subset (constellation) in a lattice
- example: with 16-QAM modulation, each symbol carries 4 data bits

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## Single antenna systems: code design criteria

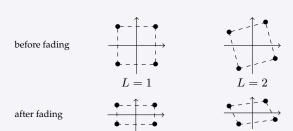
Received signal over n time slots:

$$\begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} = \begin{pmatrix} h_1 & & & \\ & h_2 & & \\ & & \ddots & \\ & & h_n \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} + \begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{pmatrix}$$

$$\mathbf{y} = \mathbf{H} \qquad \mathbf{x} + \mathbf{w}$$

#### Diversity order and product distance

To minimize the error probability, one should maximize the diversity order L, i.e. the minimum number of distinct components between any two constellation points, and the product distance  $d_p(\mathbf{x}, \mathbf{x}') = \prod_{\substack{i=1,\dots,n \\ x_i \neq x_i'}} |x_i - x_i'|$ 



#### Lattice codes from number fields

- K field extension of degree n of  $\mathbb{Q}(i)$ ,  $\sigma_1, \ldots, \sigma_n$  embeddings  $K \to \mathbb{C}$  that fix  $\mathbb{Q}(i)$
- $\mathcal{O}_K$  ring of integers of K,  $\{\theta_1, \dots, \theta_n\}$  basis of  $\mathcal{O}_K$  over  $\mathbb{Z}[i]$ .
- (relative) canonical embedding  $\phi: \mathcal{O}_K \to \mathbb{C}^n$

$$x \mapsto \mathbf{x} = (\sigma_1(x), \sigma_2(x), \dots, \sigma_n(x))^t$$

•  $x = s_1 \theta_1 + \ldots + s_n \theta_n \in \mathcal{O}_K$ ,  $\mathbf{s} = (s_1, \ldots, s_n) \in \mathbb{Z}[i]^n$  $\Rightarrow \mathbf{x} = \psi(x) = s_1 \psi(\theta_1) + \ldots + s_n \psi(\theta_n) = \Phi \mathbf{s}$  lattice point  $\Lambda = \psi(\mathcal{O}_K)$  ideal lattice

#### Full diversity property

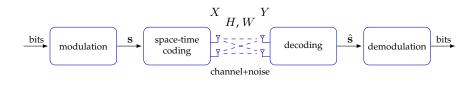
$$\forall \mathbf{x} \in \Lambda \setminus \{0\}, \quad \prod_{i=1}^{n} |x_i|^2 = \prod_{i=1}^{n} |\sigma_i(x)|^2 = N_{K/\mathbb{Q}}(x) \ge 1$$

• constructions of  $\mathbb{Z}[i]^n$  from ideal lattices [Bayer-Fluckiger *et al.* 2006]  $\Rightarrow \Phi$  unitary

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# Multiple antenna systems: space-time coding



$$Y_{n imes t} = H_{n imes m} \quad X_{m imes t} + W_{n imes t}$$
 received signal channel codeword noise

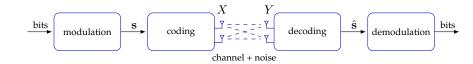
- $\bullet$  *m* transmit antennas, *n* receive antennas, *t* frame length
- introduce a dependency between the spatial (antenna) and temporal domain: codewords are represented by matrices or space-time blocks
- *H*, *W* random with i.i.d. complex Gaussian entries
- the matrix element  $x_{ij} \in \mathbb{C}$  represents the signal sent by antenna i at time j

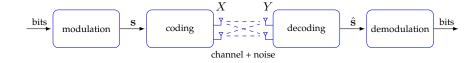
## MIMO techniques in communication standards

- HSPA+ (3G UMTS standard):  $2 \times 2$  MIMO for mobile phones, since 2010
- LTE (4G):  $2 \times 2$  and  $4 \times 4$  MIMO (2600 MHz and 800 MHz frequency bands), since 2014
- WiFi: routers and laptops have 2 or 3 antennas



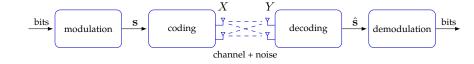
• 5G: hundreds of antennas at the base station (massive MIMO)





#### multiplexing gain:

- send independent data on each antenna
- improve the rate

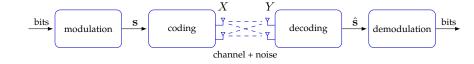


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- diversity gain:
  - send multiple copies of the same data through independent paths
  - improve the reliability
- can you do both things at the same time?
  - ⇒ diversity multiplexing gain trade-off (DMT) [Zheng and Tse 2003]

# Design criteria for space-time codes

#### **Union bound estimate of the error probability** [Tarokh *et al* 1998]

For a linear code, the difference of two codewords is still a codeword:

$$P_e \le \sum_{X \in \mathcal{C} \setminus \{0\}} \frac{1}{(\det(I + \operatorname{SNR} XX^{\dagger}))^n}$$

 $\Rightarrow$  At high signal-to noise ratio (SNR),  $P_e \leq \sum_{X \in \mathcal{C} \setminus \{0\}} \frac{1}{\text{SNR}^{nm} (\det(XX^{\dagger}))^n}$ 

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  - ullet determinant criterion: maximize  $\inf_{X \in \mathcal{C} \setminus \{0\}} \det(XX^{\dagger})$
  - ⇒ the multiplicative structure of the code plays a role
  - codes with non-vanishing determinant for any signal set achieve the DMT [Elia et al. 2006]

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#### Cyclic algebra

$$\mathcal{A} = (K/F, \sigma, \gamma) = K \oplus eK \oplus \cdots \oplus e^{n-1}K$$

where  $e \in \mathcal{A}$  satisfies the following properties:

- $xe = e\sigma(x) \ \forall x \in K$ ,
- $e^n = \gamma \in F^*$

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- $xe = e\sigma(x) \ \forall x \in K$ ,
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- $\bullet$  A is a division algebra if every nonzero element is invertible

#### Left regular representation $\psi: \mathcal{A} \to M_n(K) \subset M_n(\mathbb{C})$

$$a = x_0 + ex_1 + \dots + e^{n-1}x_{n-1} \in \mathcal{A}$$

$$\psi(a) = \begin{pmatrix} x_0 & \gamma\sigma(x_{n-1}) & \gamma\sigma^2(x_{n-2}) & \cdots & \gamma\sigma^{n-1}(x_1) \\ x_1 & \sigma(x_0) & \gamma\sigma^2(x_{n-1}) & & \gamma\sigma^{n-1}(x_2) \\ x_2 & \sigma(x_1) & \sigma^2(x_0) & & \gamma\sigma^{n-1}(x_3) \\ \vdots & & & \ddots & \vdots \\ x_{n-1} & \sigma(x_{n-2}) & \sigma^2(x_{n-3}) & \cdots & \sigma^{n-1}(x_0) \end{pmatrix}$$

$$\sigma^2(x_{n-3})$$

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- $\Lambda = \psi(\mathcal{O})$  is a matrix lattice in  $M_n(\mathbb{C})$

## Non-vanishing determinant property

• the determinant of the regular representation of an element is its reduced norm:

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#### Construction of NVD codes [Oggier et al. 2006], [Elia et al. 2006]

- if  $a \in \Lambda$ ,  $N_{\mathcal{A}/F}(a) \in \mathcal{O}_F$
- $F = \mathbb{Q}$  or  $\mathbb{Q}(\sqrt{-d}) \Rightarrow$  the ring of integers  $\mathcal{O}_F$  is discrete
- $\bullet \ \mathcal{C} \subset \psi(\mathcal{O}) \ \Rightarrow \ \inf_{X \in \mathcal{C} \setminus \{0\}} |\det X| \ge 1$

#### Examples

#### Alamouti code [Alamouti 1998]

- 2 transmit and 1 receive antenna, used in WiFi and 4G standards
- ullet  ${\cal A}$  is the algebra of Hamilton quaternions

$$X = \frac{1}{\sqrt{2}} \begin{pmatrix} s_1 & -\overline{s}_2 \\ s_2 & \overline{s}_1 \end{pmatrix}, \quad s_1, s_2 \in \mathbb{Z}[i]$$

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#### Golden Code [Belfiore et al 2005]

- $2 \times 2$  MIMO, optional profile in WiMAX standard
- $A = (\mathbb{Q}(i, \theta)/\mathbb{Q}(i), \sigma, i)$ ,  $\theta$  golden number,  $\alpha = 1 + i\sigma(\theta)$

$$X = \frac{1}{\sqrt{5}} \begin{pmatrix} \alpha(s_1 + s_2\theta) & \alpha(s_3 + s_4\theta) \\ \sigma(\alpha)i(s_3 + s_4\sigma(\theta)) & \sigma(\alpha)(s_1 + s_2\sigma(\theta)) \end{pmatrix}, \ s_1, s_2, s_3, s_4 \in \mathbb{Z}[i]$$

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#### Lattice point representation

Example: the Golden Code

$$\begin{split} X &= \begin{pmatrix} x_1 & x_3 \\ x_2 & x_4 \end{pmatrix} = \frac{1}{\sqrt{5}} \left( \begin{array}{ccc} \alpha(s_1 + s_2\theta) & \alpha(s_3 + s_4\theta) \\ \bar{\alpha}i(s_3 + s_4\bar{\theta}) & \bar{\alpha}(s_1 + s_2\bar{\theta}) \end{array} \right) \\ \mathbf{x} &= v(X) = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \frac{1}{\sqrt{5}} \left( \begin{array}{ccc} \alpha & \alpha\theta & 0 & 0 \\ 0 & 0 & \bar{\alpha}i & \bar{\alpha}\bar{\theta}i \\ 0 & 0 & \alpha & \alpha\theta \\ \bar{\alpha} & \bar{\alpha}\bar{\theta} & 0 & 0 \end{array} \right) \begin{pmatrix} s_1 \\ s_2 \\ s_3 \\ s_4 \end{pmatrix} = \Phi \mathbf{s} \end{split}$$

#### Vectorized system

$$\mathbf{y} = H_l \Phi \mathbf{s} + \mathbf{w}$$

- $H_l$  linear map corresponding to multiplication by H
- Φ (unitary) generator matrix
- s information vector

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#### Maximum likelihood (ML) decoding

Solve the closest vector problem (CVP) in the lattice generated by  $H_l$ :

$$\hat{\mathbf{x}} = \underset{\mathbf{x}' \in v(\mathcal{C})}{\operatorname{argmin}} \|\mathbf{y} - H_l \mathbf{x}'\|^2$$

# How hard are lattice problems in wireless communications?

- for general lattices, SVP and CVP are NP-hard [Ajtai 1998, Goldreich 1999]
- in lattice-based cryptography, average-case hardness is needed rather than worst-case hardness
- Ajtai discovered a connection between worst-case and average-case complexity of lattice problems

#### Different notions of random lattices

- in mathematics: use the invariant measure on the space of lattices  $\mathrm{SL}_n(\mathbb{R})/\mathrm{SL}_n(\mathbb{Z})$  derived from the Haar measure on  $\mathrm{SL}_n(\mathbb{R})$
- in cryptography: generator matrix is uniform mod q
- in communications: generator matrix has Gaussian entries

### **Decoding MIMO lattices**

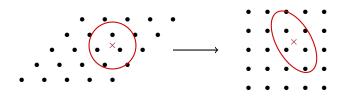
#### ML decoders

- Sphere Decoder, Schnorr-Euchner algorithm...
- optimal performance but exponential complexity

#### Suboptimal decoders

- zero forcing (ZF), successive interference cancellation (SIC)...
- polynomial complexity, but poor performance
- can be improved by preprocessing techniques

### Sphere-decoding algorithm (Finkhe-Pohst)



- enumerate all the lattice points inside a sphere centered in the received signal
- when a lattice point is found, the radius of the sphere can be updated
- apply a change of basis which maps the lattice into  $\mathbb{Z}^N$ : the sphere becomes an ellipsoid

### Complexity of sphere decoding

J. Jalden, B. Ottersten, "On the Complexity of Sphere Decoding in Digital Communications", *IEEE Transactions on Signal Processing* vol 53 n.4, 2005

- [Jaldén and Ottersten 2005]: the average complexity of the sphere decoding algorithm at *fixed SNR* is exponential and scales like  $L^{\gamma N}$ , where  $\gamma \in (0,1]$  depends on the SNR
- various techniques to reduce the complexity of sphere decoding: pruning of the decision tree, pre-processing, design of special fast-decodable codes...
- is it possible to achieve good performance with polynomial complexity?

### Channel preprocessing

#### Example: ZF decoding

$$\mathbf{y} = H\mathbf{x} + \mathbf{w}$$
  
 $\hat{\mathbf{x}}_{ZF} = \lfloor H^{-1}\mathbf{y} \rfloor = \lfloor \mathbf{x} + H^{-1}\mathbf{w} \rfloor$ 

- if *H* is orthogonal, ZF decoding is optimal
- if H is ill-conditioned, the noise  $H^{-1}\mathbf{w}$  is amplified

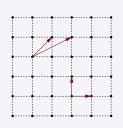
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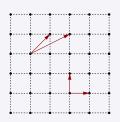
• **Solution:** channel preprocessing by lattice reduction improves the performance of suboptimal decoders



• find a better lattice basis

$$H_{\rm red} = HT$$
,

 ${\cal T}$  unimodular



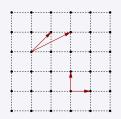
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#### LLL-ZF decoder

- = LLL + Babai rounding
  - ullet compute the pseudoinverse  $H_{\mathrm{red}}^{\dagger}$
  - $\hat{\mathbf{x}}_{\text{LLL-ZF}} = T\left(\left\lfloor H_{\text{red}}^{\dagger}\mathbf{y}\right
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- = LLL + Babai rounding
  - compute the pseudoinverse  $H_{\text{red}}^{\dagger}$
  - $\hat{\mathbf{x}}_{\text{LLL-ZF}} = T\left(\left[H_{\text{red}}^{\dagger}\mathbf{y}\right]\right)$

#### LLL-SIC decoder

- = LLL + Babai nearest plane
  - ullet QR decomposition of  $H_{\mathrm{red}}$
  - $\widetilde{\mathbf{y}} = Q^H \mathbf{y} = R \mathbf{x} + Q^H \mathbf{w}$
  - recursively compute  $\tilde{x}_N = \left\lfloor \frac{\tilde{y}_N}{r_{NN}} \right\rfloor$ ,  $\tilde{x}_i = \left\lfloor \frac{\tilde{y}_i \sum_{j=i+1}^N r_{ij} \tilde{x}_j}{r_{ij}} \right\rfloor$ ,  $i = N-1, \dots, 1$
  - $\hat{\mathbf{x}}_{\text{LLL}-\text{SIC}} = T\tilde{\mathbf{x}}$

#### Complexity

- average number of iterations in the LLL algorithm for Rayleigh fading matrices  $\sim O\left(N^2 \log N\right)$  [Jalden *et al.* 2008]
- the worst-case number of iterations is unbounded
- each iteration requires  $O(N^2)$  operations, which can be reduced to O(N) for LLL-SIC [Ling, Howgrave-Graham 2007]
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- ullet the average complexity of LLL-SIC is bounded by  $O(N^3 \log N)$
- improved decoding techniques based on LLL:
  - decoding by embedding [Luzzi, Rekaya, Belfiore 2010], [Luzzi, Stehlé, Ling 2013]
  - decoding by sampling [Liu, Ling, Stehlé 2011], [Wang, Liu, Ling 2013]

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- Coding for wireless communications
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  - MIMO systems
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Single antenna case:

$$\mathbf{y} = H\mathbf{x} + \mathbf{w},$$

$$\mathbf{x} = \psi(x) \in \Lambda = \psi(\mathcal{O}_K)$$
 ideal lattice

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#### Principle

Approximate  $H_1 = \operatorname{diag}(h'_1, \dots, h'_n)$  with  $U_l = \operatorname{diag}(\sigma_1(u), \sigma_2(u), \dots, \sigma_n(u))$ , where u is a unit of  $\mathcal{O}_K$ 

# The group of units $\mathcal{O}_K^*$

#### Dirichlet's unit theorem

K algebraic number field with  $r_1$  real  $\mathbb{Q}$ -embeddings and  $2r_2$  complex  $\mathbb{Q}$ -embeddings,  $r=r_1+r_2-1$ .

 $\exists u_1, \dots, u_r$  fundamental units such that every  $u \in \mathcal{O}_K^*$  can be written as

$$u = \zeta u_1^{e_1} \cdots u_r^{e_r},$$

where  $\zeta \in \mathcal{R}$ , the cyclic group of roots of unity in  $\mathcal{O}_K$ .

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#### The logarithmic lattice

Focus on the totally complex case:  $r_1 = 0$ ,  $r_2 = n$ . Consider  $f : \mathcal{O}_K^* \to \mathbb{R}^n$ 

$$u \mapsto f(u) = (\log |\sigma_1(u)|, \dots, \log |\sigma_n(u)|)$$

Then  $f(\mathcal{O}_K^*)$  is an (n-1)-dimensional lattice in  $\mathbb{R}^n$ :

$$\prod_{i=1}^{n} |\sigma_i(x)|^2 = N_{K/\mathbb{Q}}(x) = 1 \quad \Rightarrow \quad \sum_{i=1}^{n} \log |\sigma_i(x)| = 0$$

 the volume of the logarithmic lattice depends on the regulator of the number field

Approximate  $H_1 = \text{diag}(h'_1, \dots, h'_n)$  with  $U_l = \text{diag}(\sigma_1(u), \sigma_2(u), \dots, \sigma_n(u))$ , where u is a unit of  $\mathcal{O}_K$ :

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#### Units and unimodular transformations

u unit of  $\mathcal{O}_K \quad \Leftrightarrow \quad U_l \Phi = \Phi T_u$  with  $T_u$  unimodular (with entries in  $\mathbb{Z}[i]$ ).

Proof: 
$$ux \in \mathcal{O}_K \Rightarrow ux = \sum_i s_i' \theta_i$$
 
$$U_l \psi(x) = \psi(ux) = \Phi \mathbf{s}' = U_l \Phi \mathbf{s}$$
 
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• Received signal:

$$\mathbf{y}' = EU_l\Phi\mathbf{s} + \mathbf{w}' = E\Phi T_u\mathbf{s} + \mathbf{w}' = E\Phi\mathbf{s}' + \mathbf{w}', \quad \mathbf{s}' \in \mathbb{Z}[i]^n$$

• apply a suboptimal decoder (i.e. ZF):

$$\hat{\mathbf{s}}' = \left[ \Phi^{-1} E^{-1} \mathbf{y}' \right] = \left[ \mathbf{s}' + \underbrace{\Phi^{-1}}_{\text{unitary}} E^{-1} \mathbf{w}' \right]$$

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- the *i*-th component of the equivalent noise is  $(E^{-1}\mathbf{w}')_i = \frac{\sigma_i(u)}{h_i'}w_i'$
- to minimize noise variance,  $\left|\frac{\sigma_i(u)}{h_i'}\right|$  should be small  $\forall i = 1, \dots, n$   $\Rightarrow |\log |\sigma_i(u)| \log |h_i'|$  should be small

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 advantage: the logarithmic lattice is fixed once and for all and doesn't depend on the channel

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- it outperforms LLL + ZF in high dimension

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 used in [Campello, Ling, Belfiore 2017] to show that mod-p lattices achieve constant gap to compound capacity for n-antenna systems with reduced complexity

# Algebraic reduction for fast fading channels

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### Recent results

 used in [Campello, Ling, Belfiore 2017] to show that mod-p lattices achieve constant gap to compound capacity for n-antenna systems with reduced complexity

- ullet the performance depends on the covering radius  $r_{
  m cov}$  of the logarithmic lattice
- no known general bounds for  $r_{\rm cov}$
- ullet bounds for  $r_{\rm cov}$  in cyclotomic fields of prime power index [Cramer, Ducas, Peikert, Reigev 2016]

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• Multiple antenna case:

$$Y = HX + W$$

- $\mathcal{A} = (K/\mathbb{Q}(i), \sigma, \gamma)$  division algebra,  $[K : \mathbb{Q}(i)] = n$
- $X \in \psi(\mathcal{O}\alpha)$ ,  $\mathcal{O}$  maximal order of  $\mathcal{A}$

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• **Idea:** approximate  $H_1$  with a unit  $U \in \mathcal{O}^1$ 

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• in vectorized form:

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- $A_l$  linear map corresponding to left multiplication by A
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### How to choose U?

 $\Rightarrow$  Choose U that minimizes  $\left\|E^{-1}\right\|_F = \left\|UH_1^{-1}\right\|_F$ 

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### Action of $SL_2(\mathbb{C})$ on hyperbolic 3-space

$$\mathbb{H}^3 = \{(z, r) \mid z \in \mathbb{C}, \ r \in \mathbb{R}^+\}$$

with the hyperbolic distance  $\rho$  such that  $\cosh \rho(P,P')=1+rac{d(P,P')}{2rr'}$ 

$$\begin{split} A &= \begin{pmatrix} a & b \\ c & d \end{pmatrix} \\ J &= (0,0,1) &\mapsto \quad A(J) &= \left(\frac{\operatorname{Re}(b\bar{d} + a\bar{c})}{|c|^2 + |d|^2}, \frac{\operatorname{Im}(b\bar{d} + a\bar{c})}{|c|^2 + |d|^2}, \frac{1}{|c|^2 + |d|^2}\right) \end{split}$$

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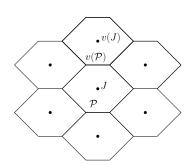
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$$||H_1U^{-1}||_F$$
 is small  $\Leftrightarrow U^{-1}(J)$  is close to  $H_1^{-1}(J)$ 

# Fundamental domain and generators of the group

### Poincaré's polyhedron theorem

- the fundamental domain  $\mathcal P$  for the action of  $\Gamma$  on  $\mathbb H^3$  is a compact hyperbolic polyhedron
- the copies  $v(\mathcal{P}), v \in \Gamma$  are isometric and form a tiling of  $\mathbb{H}^3$
- there is a correspondence between a set of generators of the group and the set of side-pairings which map a face of  $\mathcal{P}$  into another face



### Tamagawa volume formula

$$\operatorname{Vol}(\mathcal{P}) = \frac{\zeta_F(2)}{4\pi^2} |D_F|^{\frac{3}{2}} \prod_{p|\delta_{\mathcal{O}}} (N_p - 1)$$

### Example: action of $\mathbb{Z}^2$ on $\mathbb{R}^2$

 the area enclosed by bisectors is a fundamental domain for the action

. . .

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. . .

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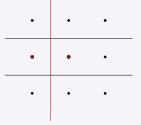
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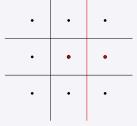
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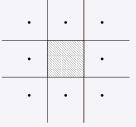
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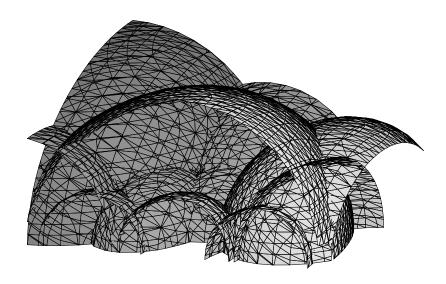
### Action of $\Gamma$ on $\mathbb{H}^3$

the bisectors are Euclidean spheres



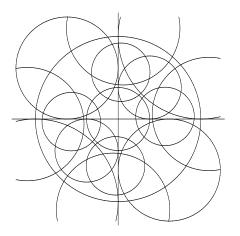
- the fundamental domain is a hyperbolic polyhedron
- the images of the fundamental domain form a tiling of  $\mathbb{H}^3$

# Intersecting bisectors: the Golden Code



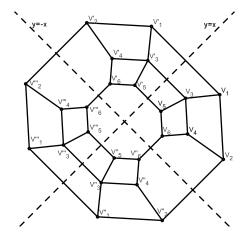
# Intersecting bisectors: the Golden Code

Projection on the plane  $\{r=0\}$ 



# The fundamental polyhedron

Projection on the plane  $\{r=0\}$ 



# Finding the generators

 The generators of the group correspond to the side-pairings of the fundamental polyhedron





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 The generators of the group correspond to the side-pairings of the fundamental polyhedron





### Golden Code: 8 generators for the unit group

$$U_{1} = \begin{pmatrix} i\theta & 0 \\ 0 & i\bar{\theta} \end{pmatrix} \qquad U_{5} = \begin{pmatrix} 1+i & 1+i\bar{\theta} \\ i(1+i\theta) & 1+i \end{pmatrix}$$

$$U_{2} = \begin{pmatrix} i & 1+i \\ i-1 & i \end{pmatrix} \qquad U_{6} = \begin{pmatrix} 1+i & 1+i\bar{\theta} \\ i(1+i\bar{\theta}) & 1+i \end{pmatrix}$$

$$U_{3} = \begin{pmatrix} \theta & 1+i \\ i-1 & \bar{\theta} \end{pmatrix} \qquad U_{7} = \begin{pmatrix} 1-i & \bar{\theta}+i \\ i(\bar{\theta}+i) & 1-i \end{pmatrix}$$

$$U_{4} = \begin{pmatrix} \theta & -1-i \\ -i+1 & \bar{\theta} \end{pmatrix} \qquad U_{8} = \begin{pmatrix} 1-i & \theta+i \\ i(\bar{\theta}+i) & 1-i \end{pmatrix}$$

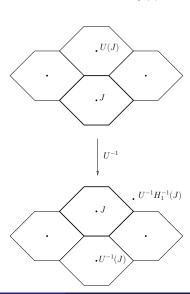
• actually this is not a minimal set: 6 units are enough

# The approximation algorithm

• the polyhedra adjacent to the fundamental polyhedron  $\mathcal{P}$  are of the form  $U(\mathcal{P})$ , with U a generator

### Unit search algorithm

- 1) find the generator U such that U(J) is closest to  $H_1^{-1}(J)$
- 2) every U is an isometry  $\Rightarrow$  apply  $U^{-1}$
- Repeat steps 1-2 until *J* is the closest point to H<sub>1</sub><sup>-1</sup>(*J*)



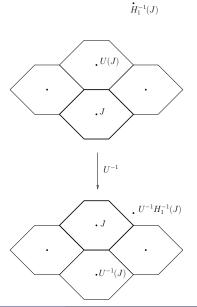
 $\dot{H}_1^{-1}(J)$ 

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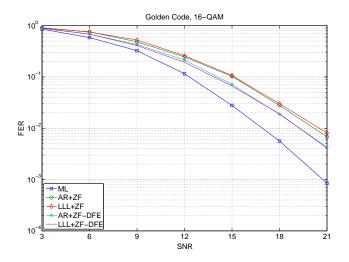
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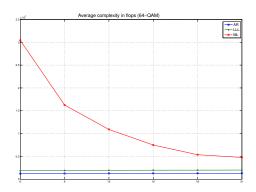
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- Repeat steps 1-2 until *J* is the closest point to H<sub>1</sub><sup>-1</sup>(*J*)
- this algorithm is suboptimal does not solve the word problem for groups!



# Performance of algebraic reduction



# Complexity of algebraic reduction



- the average number of iterations in the AR algorithm is only 1.923
- with high probability,  $H_1^{-1}(J)$  is already contained in  $\mathcal P$  or one of the neighboring polyhedra
- advantage: if fading is slow, AR requires only a slight adjustment of the previous approximation

# Generalization to other codes (quaternion algebras)

 general algorithm to find generators of the unit group [Swan 1971, Corrales et al. 2004, Page 2015]

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### Design codes that are optimal for algebraic reduction

- the quality of the approximation depends on the diameter of the fundamental polyhedron (not directly related to volume!)
- the speed of the approximation depends on the number of generators of the unit group

# Generalization to other codes (quaternion algebras)

 general algorithm to find generators of the unit group [Swan 1971, Corrales et al. 2004, Page 2015]

### Design codes that are optimal for algebraic reduction

- the quality of the approximation depends on the diameter of the fundamental polyhedron (not directly related to volume!)
- the speed of the approximation depends on the number of generators of the unit group
- the unit group can be very complex in general
- for the "Golden +" code algebra [Vehkalahti et al. 2009] it seems to have hundreds of generators
- quaternion algebras over  $\mathbb{Q}(\zeta_3)$  with 3 generators [Alves-Belfiore 2012] and over  $\mathbb{Q}(\sqrt{-7})$  with small Tamagawa volume [Alves-Belfiore 2015]

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### Related work

• the growth rate of units of bounded norm characterizes DMT and error performance of division algebra codes [Vehkalahti, Lu, Luzzi 2013], [Luzzi, Vehkalahti 2018]

Thank you for your attention!!