# An Algebraic Approach to Physical-Layer Network Coding 

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## Finite-Field Matrix Channels

## Random Linear Network Coding

- Transmitter injects packets: vectors from $\mathbb{F}_{q}^{m}$, the rows of a matrix $X$
- Intermediate nodes forward random $\mathbb{F}_{q}$-linear combinations of packets
- Errors may also be injected, which randomly mix with the legitimate packets
- (Each) receiver gathers as many packets as possible, forming the rows of matrix $Y$


## At any particular receiver:

$$
Y=A X+Z
$$

where: $X$ is $n \times m ; Y, Z$ are $N \times m$; and $A$ is $N \times n$.

## A Basic Model



In previous work ${ }^{1}$ we considered a basic stochastic linear matrix channel model:

$$
Y=A X+Z
$$

where

- $X$ and $Y$ are $n \times m$ matrices over $\mathbb{F}_{q}$;
- $A$ is $n \times n$, nonsingular, drawn uniformly at random;
- $Z$ is $n \times m$ with rank $t$, drawn uniformly at random;
- $X, A$, and $Z$ are independent.
${ }^{1}$ D. Silva, K., R. Kötter, "Communication over Finite-Field Matrix Channels," IEEE Trans. Inf. Theory, vol. 56, pp. 1296-1305, Mar. 2010.


## MAMC: Capacity

## Theorem (upper bound)

For $n \leq m / 2$,

$$
C_{\text {MAMC }} \leq(m-n)(n-t)+\log _{q} 4(n+1)(t+1)
$$

## Theorem (lower bound)

Assume $n \leq m$. For any $\epsilon \geq 0$, we have

$$
C_{\mathrm{MAMC}} \geq(m-n)(n-t-\epsilon t)-\log _{q} 4-\frac{2 t n m}{q^{1+\epsilon t}}
$$

These upper and lower bounds match when $q \rightarrow \infty$ or $m \rightarrow \infty$ (with $n / m$ and $t / n$ fixed).

## MAMC: Capacity

## Corollary

For large $m$ or large $q$,

$$
C_{\mathrm{MAMC}} \approx(m-n)(n-t) .
$$



## A Simple Coding Scheme

## Strategy: Channel Sounding + Error Trapping

Use channel sounding "inside" and error trapping "outside" (but not the opposite!)


## MAMC: A Coding Scheme

First, rewrite the channel model as

$$
Y=A X+Z=A\left(X+A^{-1} Z\right)=A(X+W), \quad \text { where } W=A^{-1} Z
$$

and suppose a "genie" gives the receiver $X+W$.
Let data matrix $D$ be $(n-t) \times(m-n)$.
We have:

$$
X=\left[\begin{array}{lll}
0 & 0 & 0 \\
0 & 1 & D
\end{array}\right] \quad W=\left[\begin{array}{lll}
W_{1} & W_{2} & W_{3} \\
W_{4} & W_{5} & W_{6}
\end{array}\right]
$$

Assume that rank $W_{1}=t=\operatorname{rank} W(=\operatorname{rank} Z)$. In this case, for some matrix $B$, we have

$$
W=\left[\begin{array}{ccc}
W_{1} & W_{2} & W_{3} \\
B W_{1} & B W_{2} & B W_{3}
\end{array}\right]
$$

Now convert $X+W$ to reduced row echelon (RRE) form:

$$
\begin{aligned}
& X+W=\left[\begin{array}{ccc}
W_{1} & W_{2} & W_{3} \\
B W_{1} & I+B W_{2} & D+B W_{3}
\end{array}\right] \\
& \xrightarrow{\text { row op. }}\left[\begin{array}{ccc}
I & W_{1}^{-1} W_{2} & W_{1}^{-1} W_{3} \\
B W_{1} & I+B W_{2} & D+B W_{3}
\end{array}\right] \\
& \xrightarrow{\text { row op. }}\left[\begin{array}{ccc}
I & W_{1}^{-1} W_{2} & W_{1}^{-1} W_{3} \\
0 & I & D
\end{array}\right] \\
& \xrightarrow{\text { row op. }}\left[\begin{array}{ccc}
I & 0 & \tilde{W}_{3} \\
0 & I & D
\end{array}\right]=\operatorname{RRE}(X+W) .
\end{aligned}
$$

But we have $Y$, not $X+W$ !

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$$

But we have $Y$, not $X+W$ !

## Observation

$Y=A(X+W), A$ is full rank, so $Y$ and $X+W$ have the same row space, which implies that

$$
\operatorname{RRE}(Y)=\operatorname{RRE}(X+W)
$$

Thus, $D$ is exposed by reducing $Y$ to RRE form!

## MAMC: A Coding Scheme

- Decoding amounts to performing full Gaussian elimination on the received matrix $Y$.

Complexity: $\mathcal{O}\left(n^{2} m\right)$ operations in $\mathbb{F}_{q}$ to recover $(n-t)(m-n)$ symbols. Defining $R=(n-t)(m-t) / m n$, we have a complexity of $\mathcal{O}(n / R)$ operations per decoded symbol.

- The scheme fails if $W_{1}$ is not invertible.

The probability of failure falls exponentially (for fixed $m$ ) in the number of bits per field-element, or exponentially (for fixed $q$ ) in $m$ (assuming fixed aspect ratio of $m / n$ and fixed $t / n$ ).

## Theorem

This coding scheme can achieve the capacity of the MAMC when either $q \rightarrow \infty$ or $m \rightarrow \infty$.

## This Talk:

$$
\begin{gathered}
\text { Generalize } \\
\text { from } \\
\text { finite-field matrix channels } \\
\text { to } \\
\text { finite-ring matrix channels. } \\
\text { Mhy? }
\end{gathered}
$$

> Generalize from
> finite-field matrix channels
> to
> finite-ring matrix channels.

## Why?

A: it could be useful for nested-lattice-based physical-layer network coding (LNC), a form of compute-and-forward relaying à la
B. Nazer and M. Gastpar, "Compute-and-forward: Harnessing interference through structured codes," IEEE Trans. Inf. Theory, vol. 57, pp. 6463-6486, Oct. 2011.

## Compute-and-Forward: Nested Lattices

## Nested Lattices

Fine lattice $\Lambda$, coarse lattice $\Lambda^{\prime} \subseteq \Lambda$, and lattice quotient $\Lambda / \Lambda^{\prime}$


$$
\mathbf{G}_{\Lambda}=\left[\begin{array}{cc}
\sqrt{3} & 1 \\
0 & 2
\end{array}\right]
$$

$$
\Lambda=\left\{\mathbf{r} \mathbf{G}_{\Lambda}: \mathbf{r} \in \mathbb{Z}^{2}\right\}
$$

$$
\Lambda^{\prime}=3 \Lambda
$$

## Compute-and-Forward: Complex Lattices

## Complex R-Lattices

Let $R$ be a discrete subring of $\mathbb{C}$ forming a principal ideal domain.
Let $N \leq n$. An $R$-lattice of dimension $N$ in $\mathbb{C}^{n}$ is defined as the set of all $R$-linear combinations of $N$ linearly independent vectors, i.e.,

$$
\Lambda=\left\{\mathbf{r} \mathbf{G}_{\Lambda}: \mathbf{r} \in R^{N}\right\}
$$

where $\mathbf{G}_{\Lambda} \in \mathbb{C}^{N \times n}$ is called a generator matrix for $\Lambda$.
$R=\mathbb{Z}[\omega] \Rightarrow$ Eisenstein lattices; $R=\mathbb{Z}[i] \Rightarrow$ Gaussian lattices

| $\bullet$ | $\bullet$ | $\bullet$ |
| :---: | :---: | :---: |
| $\bullet$ | $\bullet$ | $\bullet$ |
| $\bullet$ | $\bullet$ | $\bullet$ |

$$
\mathbb{Z}[\omega] \triangleq\left\{a+b \omega: a, b \in \mathbb{Z}, \omega=e^{i 2 \pi / 3}\right\}
$$

$$
\mathbb{Z}[i] \triangleq\{a+b i: a, b \in \mathbb{Z}\}
$$

$$
\Lambda=\mathbb{Z}[i]
$$

$$
\Lambda^{\prime}=3 \mathbb{Z}[i]
$$

## Compute-and-Forward: Structure of $\Lambda / \Lambda^{\prime}$

## Theorem

$$
\Lambda / \Lambda^{\prime} \cong R /\left\langle\pi_{1}\right\rangle \times \cdots \times R /\left\langle\pi_{k}\right\rangle
$$

for some nonzero, non-unit $\pi_{1}, \ldots, \pi_{k} \in R$ such that $\pi_{1}|\cdots| \pi_{k}$. Moreover, there exists a surjective $R$-module homomorphism $\varphi: \Lambda \rightarrow R /\left\langle\pi_{1}\right\rangle \times \cdots \times R /\left\langle\pi_{k}\right\rangle$ whose kernel is $\Lambda^{\prime}$.

| $2+i$ | $i$ <br> $\bullet$ | $1+i$ <br> $\bullet$ |
| :---: | :---: | :---: |
| 2 | 0 | $\bullet$ |
| $\bullet$ | $\bullet$ |  |
| $2+2 i$ | $2 i$ |  |
| $\bullet$ | $\bullet$ | $1+2 i$ <br>  |

$$
\begin{gathered}
\Lambda / \Lambda^{\prime} \cong \mathbb{Z}[i] /\langle 3\rangle \\
\varphi(a+b i)=(a+b i) \bmod 3 \\
\varphi^{-1}(c+d i)=(c+d i)+\Lambda^{\prime}
\end{gathered}
$$

## Compute-and-Forward: Architecture


$R /\left\langle\pi_{1}\right\rangle \times \cdots \times R /\left\langle\pi_{k}\right\rangle$ is the message space $\Omega$

## Encoding

Transmitter $\ell$ sends $\mathbf{x}_{\ell} \in \Lambda$, a coset representative of $\varphi^{-1}\left(\mathbf{w}_{\ell}\right)$

## Decoding

Receiver first recovers $\sum_{\ell} a_{\ell} \mathbf{x}_{\ell}$ from $\alpha \mathbf{y}$; Receiver then maps $\sum_{\ell} a_{\ell} \mathbf{x}_{\ell}$ onto $\sum_{\ell} a_{\ell} \mathbf{w}_{\ell}$ via $\varphi$

Remark: $\alpha \mathbf{y}-\sum_{\ell} a_{\ell} \mathbf{x}_{\ell}=\sum_{\ell}\left(\alpha h_{\ell}-a_{\ell}\right) \mathbf{x}_{\ell}+\alpha \mathbf{z}$ "effective noise"

## Construction Examples

Example 1: [Ordentlich, Zhan, Erez, Gastpar, Nazer, ISIT'11]

- $\Lambda$ is obtained using Construction A applied to binary ( $n=64800, k=54000$ ) LDPC code $C$, with mod-4 shaping:

$$
\Lambda=C+2 \mathbb{Z}^{n}, \quad \Lambda^{\prime}=4 \mathbb{Z}^{n}
$$

- Induced message space: $\mathbb{Z}_{4}^{54000} \times \mathbb{Z}_{2}^{10800}$

Example 2: Turbo Lattices [Sakzad, Sadeghi, Panario, Allerton'10]

- $\Lambda$ is obtained using Construction D applied to nested turbo codes $C_{2}:\left(n=10131, k_{2}=3377\right)$ and $C_{1}:\left(n=10131, k_{1}=5065\right)$;

$$
\Lambda=C_{2}+2 C_{1}+4 \mathbb{Z}^{n}, \quad \Lambda^{\prime}=4 \mathbb{Z}^{n} .
$$

- Induced message space: $\mathbb{Z}_{4}^{3377} \times \mathbb{Z}_{2}^{1688}$

In general, for most practical constructions, we have

$$
\Omega=R /\left\langle\pi^{t_{0}}\right\rangle \times \cdots \times R /\left\langle\pi^{t_{m-1}}\right\rangle, t_{0} \geq \cdots \geq t_{m-1} .
$$

## Much Ongoing Work:

B. Nazer and M. Gastpar, "Compute-and-forward: Harnessing interference through structured codes," IEEE Trans. Inf. Theory, vol. 57, no. 10, pp. 6463-6486, Oct. 2011.
M. P. Wilson, K. Narayanan, H. D. Pfister, and A. Sprintson, "Joint physical layer coding and network coding for bidirectional relaying," IEEE Trans. Inf. Theory, vol. 56, no. 11, pp. 5641-5654, Nov. 2010.
N. E. Tunali, K. R. Narayanan, J. J. Boutros, and Y.-C. Huang, "Lattices over Eisenstein integers for compute-and-forward," in Proc. 2012 Allerton Conf. Commun., Control, and Comput., Monticello, IL, Oct. 2012, pp. 33-40.
S. Qifu and J. Yuan, "Lattice network codes based on Eisenstein integers," in Proc. 2012 IEEE Int. Conf. on Wireless and Mobile Comput., Barcelona, Spain, Oct. 2012, pp. 225-231.
A. Osmane and J.-C. Belfiore, "The compute-and-forward protocol: implementation and practical aspects," 2011.
S. Gupta and M. A. Vázquez-Castro, "Physical-layer network coding based on integer-forcing precoded compute-and-forward," 2013.

## Chain Rings,

Modules,

## Matrices



## Commutative Rings with Identity $1 \neq 0$

- Ideals in a ring can be partially ordered by subset inclusion.
- The resulting poset is called the lattice of ideals of the ring.


Chain ring: ideals are linearly ordered. Ex: $\mathbb{Z}_{8}$.
Principal ideal ring: every ideal gen. by 1 element. Ex: $\mathbb{Z}_{8}, \mathbb{Z}_{2} \times \mathbb{Z}_{4}$. Local ring: unique maximal proper ideal. Ex: $\mathbb{Z}_{8}, \mathbb{Z}_{2}[X, Y] /\langle X, Y\rangle^{2}$.

## Finite Rings: Important Facts

## Proposition

If $R$ is a ring and $N$ is a maximal ideal of $R$, then $R / N$ is a field.
This is called a residue field.

## Proposition

A finite ring is a chain ring if and only if it is both local and principal.

## Proposition

Every finite principal ideal ring is a product of finite chain rings.

## Finite Chain Rings: The Ideals


$\{0\}=\left\langle\pi^{s}\right\rangle$

Let $R$ be a finite chain ring, where

- $\langle\pi\rangle$ is the unique maximal ideal,
- $q$ is the order of the residue field,
- $s$ is the number of proper ideals.


## Proposition

The lattice of ideals of $R$ is

$$
R \supset\langle\pi\rangle \supset\left\langle\pi^{2}\right\rangle \supset \cdots \supset\left\langle\pi^{s-1}\right\rangle \supset\left\langle\pi^{s}\right\rangle=\{0\} .
$$

We have $\left|\left\langle\pi^{i}\right\rangle\right|=q^{s-i}$; in particular $|R|=q^{s}$.

Notation: $(q, s)$ chain ring.

## Finite Chain Rings: Examples

The following are two non-isomorphic $(q=2, s=2)$ chain rings.


In other words, specifying $q$ and $s$ does not uniquely specify the chain ring.

## Finite Chain Rings: The $\pi$-adic Decomposition

Let $R$ be a ( $q, s$ ) chain ring.

## Proposition

Fix the following:

- $\pi \in R$, a generator for the maximal ideal $\langle\pi\rangle$.
- $\mathcal{R}(R, \pi)$, a complete set of residues with respect to $\pi$.

Then every element $r \in R$ can be written uniquely as

$$
r=r_{0}+r_{1} \pi+r_{2} \pi^{2}+\cdots+r_{s-1} \pi^{s-1}
$$

where $r_{i} \in \mathcal{R}(R, \pi)$.
This is known as the $\pi$-adic decomposition.

## Element Degree

## Definition

The degree, $\operatorname{deg}(r)$, of a nonzero element $r \in R^{*}$, where

$$
r=r_{0}+r_{1} \pi+\cdots+r_{s-1} \pi^{s-1}
$$

is defined as the least index $j$ for which $r_{j} \neq 0$.

- by convention, $\operatorname{deg}(0)=s$
- units have degree zero
- elements of the same degree are associates
- a divides $b$ if and only if $\operatorname{deg}(a) \leq \operatorname{deg}(b)$
- $\operatorname{deg}(a+b) \geq \min \{\operatorname{deg}(a), \operatorname{deg}(b)\}$


## Shapes

An s-shape $\mu=\left(\mu_{1}, \mu_{2}, \ldots, \mu_{s}\right)$ is a sequence of non-decreasing non-negative integers, i.e., $0 \leq \mu_{1} \leq \mu_{2} \leq \cdots \leq \mu_{s}$. We denote by $|\mu|$ the sum of its components, i.e., $|\mu|=\sum_{i=1}^{s} \mu_{i}$.

## Example: $\mu=(4,6,8)$

```
* * * *
****** }|(4,6,8)|=1
* * * * * * * *
```

For convenience, we will sometimes identify the integer $t$ with the $s$-shape $(t, \ldots, t)$.
An $s$-shape $\kappa=\left(\kappa_{1}, \ldots, \kappa_{s}\right)$ is said to be a subshape of $\mu=\left(\mu_{1}, \ldots, \mu_{s}\right)$, written $\kappa \preceq \mu$, if $\kappa_{i} \leq \mu_{i}$ for all $i=1, \ldots, s$.

$$
\begin{array}{llll}
* & * & * & * \\
* & * & * & *
\end{array} *_{*} \quad(4,4,5) \preceq(4,6,8)
$$

## From Shape to Module

When $R$ is a finite chain ring, an $R$-module is always isomorphic to a direct product of various ideals of $R$; this structure can be described by a shape.

## Definition

Let $R$ be a $(q, s)$ chain ring with maximal ideal $\langle\pi\rangle$. For any $s$-shape $\mu$, we define the $R$-module $R^{\mu}$ as

$$
R^{\mu} \triangleq \underbrace{\langle 1\rangle \times \cdots \times\langle 1\rangle}_{\mu_{1}} \times \underbrace{\langle\pi\rangle \times \cdots \times\langle\pi\rangle}_{\mu_{2}-\mu_{1}} \times \cdots \times \underbrace{\left\langle\pi^{s-1}\right\rangle \times \cdots \times\left\langle\pi^{s-1}\right\rangle}_{\mu_{s}-\mu_{s-1}} .
$$

$R^{\left(\mu_{1}, \ldots, \mu_{s}\right)}$ is a collection of $\mu_{s}$-tuples over $R$, whose $\pi$-adic coordinate array must satisfy degree constraints specified by $\left(\mu_{1}, \ldots, \mu_{s}\right)$.

Note that $\left|R^{\mu}\right|=q^{|\mu|}$.


$$
s=3, \mu=(4,6,8)
$$

## From Module to Shape

Conversely, we have the following theorem (see, e.g., $[\mathrm{HLOO}]^{2}$ ).

## Theorem

For any finite $R$-module $M$ over a $(q, s)$ chain ring $R$, there is a unique s-shape $\mu$ such that $M \cong R^{\mu}$.

- We call the unique shape $\mu$ associated with a module $M$ the shape of $M$, and write $\mu=$ shape $M$.
- If $M^{\prime}$ is a submodule of $M$, then shape $M^{\prime} \preceq$ shape $M$, i.e., the shape of a submodule is a subshape of the module.

For example, the module spanned by 1111 and 0022 over $\mathbb{Z}_{8}$ has shape $(1,2,2)$. This module contains $2^{5} 4$-tuples, and is isomorphic to $\langle 1\rangle \times\langle 2\rangle$.
${ }^{2}$ T. Honold and I. Landjev, "Linear Codes over Finite Chain Rings," The Electronic J. of Combinatorics, vol. 7, 2000.

## Counting Submodules

It is also known [HLOO] that the number of submodules of $R^{\mu}$ whose shape is $\kappa$ is given by

$$
\llbracket \begin{align*}
& \mu  \tag{1}\\
& \kappa
\end{align*} \rrbracket_{q}=\prod_{i=1}^{s} q^{\left(\mu_{i}-\kappa_{i}\right) \kappa_{i-1}}\left[\begin{array}{l}
\mu_{i}-\kappa_{i-1} \\
\kappa_{i}-\kappa_{i-1}
\end{array}\right]_{q}
$$

where

$$
\left[\begin{array}{c}
m \\
k
\end{array}\right]_{q} \triangleq \prod_{i=0}^{k-1} \frac{q^{m}-q^{i}}{q^{k}-q^{i}}
$$

is the Gaussian coefficient.
In particular, when the chain length $s=1, R$ becomes the finite field $\mathbb{F}_{q}$ of $q$ elements, and $\left[\begin{array}{c}\mu \\ \kappa\end{array}\right]_{q}$ becomes $\left[\begin{array}{c}\mu_{1} \\ \kappa_{1}\end{array}\right]_{q}$, which is the number of $\kappa_{1}$-dimensional subspaces of $\mathbb{F}_{q}^{\mu_{1}}$.

## Matrices over Finite Chain Rings

Notation for matrices:

- $R^{n \times m}$ : the set of all $n \times m$ matrices with entries from ring $R$.
- $U \in R^{n \times n}$ is invertible if $U V=V U=I_{n}$ for some $V \in R^{n \times n}$, where $I_{n}$ denotes the $n \times n$ identity matrix. The set of invertible matrices in $R^{n \times n}$ forms the general linear group $G L_{n}(R)$ under multiplication.
- $A, B \in R^{n \times m}$ are left-equivalent if there exists a matrix $U \in \mathrm{GL}_{n}(R)$ such that $U A=B$.
- $A, B \in R^{n \times m}$ are equivalent if there exist matrices $U \in \mathrm{GL}_{n}(R)$ and $V \in \mathrm{GL}_{m}(R)$ such that $U A V=B$.
- $D \in R^{n \times m}$ is a diagonal matrix if $D[i, j]=0$ whenever $i \neq j$. A diagonal matrix need not be square.

$$
\left[\begin{array}{llll}
* & 0 & 0 & 0 \\
0 & * & 0 & 0 \\
0 & 0 & * & 0
\end{array}\right] \quad\left[\begin{array}{lll}
* & 0 & 0 \\
0 & * & 0 \\
0 & 0 & * \\
0 & 0 & 0
\end{array}\right]
$$

## Smith Normal Form

## Definition

A diagonal matrix $D=\operatorname{diag}\left(d_{1}, \ldots, d_{r}\right) \in R^{n \times m}$ is called a Smith normal form of $A \in R^{n \times m}$, if $D$ is equivalent to $A$ and $d_{1}\left|d_{2}\right| \cdots \mid d_{r}$ in $R$, where $r=\min \{n, m\}$.

Every matrix over a PIR (in particular, a finite chain ring) has a Smith normal form whose diagonal entries are unique up to equivalence of associates.
Over $R=\mathbb{Z}_{8}$

$$
A=\left[\begin{array}{llll}
4 & 6 & 2 & 1 \\
0 & 0 & 0 & 2 \\
2 & 4 & 6 & 1 \\
2 & 0 & 2 & 1
\end{array}\right]=\underbrace{\left[\begin{array}{llll}
1 & 2 & 0 & 0 \\
2 & 0 & 1 & 0 \\
1 & 1 & 0 & 0 \\
1 & 1 & 1 & 1
\end{array}\right]}_{U} \underbrace{\left[\begin{array}{llll}
1 & 0 & 0 & 0 \\
0 & 2 & 0 & 0 \\
0 & 0 & 4 & 0 \\
0 & 0 & 0 & 0
\end{array}\right]}_{S} \underbrace{\left[\begin{array}{llll}
0 & 2 & 2 & 1 \\
1 & 1 & 2 & 0 \\
0 & 1 & 1 & 0 \\
0 & 0 & 1 & 0
\end{array}\right]}_{V}
$$

with invertible $U$ and $V$. Since $1|2| 4 \mid 0$ in $\mathbb{Z}_{8}, S$ is the Smith normal form of $A$.

## Row and Column Span

- For $A \in R^{n \times m}$, denote by row $A$ and $\operatorname{col} A$ the row span and column span of $A$, respectively.
- From the Smith normal form, it is easy to see that row $A \cong \operatorname{col} A$.
- Two matrices $A, B \in R^{n \times m}$ are left-equivalent if and only if row $A=$ row $B$.
- Two matrices $A, B \in R^{n \times m}$ are equivalent if and only if row $A \cong$ row $B$.


## Shape of a Matrix

## Definition

The shape of a matrix $A$ is defined as the shape of the row span of A, i.e.,

$$
\text { shape } A=\text { shape }(\text { row } A)
$$

Clearly, shape $A=$ shape $(\operatorname{col} A)$. shape $A=\mu$ if and only if the Smith normal form of $A$ is given by

$$
\operatorname{diag}(\underbrace{1, \ldots, 1}_{\mu_{1}}, \underbrace{\pi, \ldots, \pi}_{\mu_{2}-\mu_{1}}, \ldots, \underbrace{\pi^{s-1}, \ldots, \pi^{s-1}}_{\mu_{s}-\mu_{s-1}}, \underbrace{0, \ldots, 0}_{r-\mu_{s}})
$$

where $r=\min \{n, m\}$.
A matrix $U \in R^{n \times n}$ is invertible if and only if shape $U=(n, \ldots, n)$.

## Example

If $A$ has Smith normal form $D=\operatorname{diag}(1,2,4,0)$ over $\mathbb{Z}_{8}$ then shape $A=(1,2,3)$.

## Properties of Matrix Shape

Let $A \in R^{n \times m}$ and $B \in R^{m \times k}$. Then

- shape $A=$ shape $A^{T}$, where $A^{T}$ is the transpose of $A$.
- For any $P \in \mathrm{GL}_{n}(R), Q \in \mathrm{GL}_{m}(R)$, shape $A=$ shape $P A Q$.
- shape $A B \preceq$ shape $A$, shape $A B \preceq$ shape $B$.
- For any submatrix $C$ of $A$, shape $C \preceq$ shape $A$.


## Row Canonical Form

Let $R$ be a $(q, s)$ chain ring with maximal ideal $\langle\pi\rangle$, fixing a complete set of residues $\mathcal{R}(R, \pi)$ (including 0 ), and for $1<\ell<s$, fixing

$$
\mathcal{R}\left(R, \pi^{\ell}\right)=\left\{\sum_{i=0}^{\ell-1} a_{i} \pi^{i}: a_{0}, \ldots, a_{\ell-1} \in \mathcal{R}(R, \pi)\right\}
$$

## Example: $R=\mathbb{Z}_{8}$

If $R=\mathbb{Z}_{8}$, with $\pi=2$, we might fix $\mathcal{R}(R, 2)=\{0,1\}$, so that $\mathcal{R}(R, 4)=\{0,1,2,3\}$.

## Row Canonical Form (cont'd)

In a matrix $A$ :

- The element $A[i, j]$ occurs above $A\left[i^{\prime}, j^{\prime}\right]$ if $i<i^{\prime}$. (Equivalently, $A\left[i^{\prime}, j^{\prime}\right]$ occurs below $A[i, j]$.)
- The element $A[i, j]$ occurs earlier than $A\left[i^{\prime}, j^{\prime}\right]$ if $j<j^{\prime}$. (Equivalently, $A\left[i^{\prime}, j^{\prime}\right]$ occurs later than $A[i, j]$.)
- The first element in row $i$ with property $P$ occurs earlier than any other element in row $i$ with property $P$.
- The pivot of a nonzero row of $A$ is the first entry among the entries having least degree in that row. For example, the pivot of [0 462 2] over $\mathbb{Z}_{8}$ is the element 6 .


## Row Canonical Form (cont'd)

## Definition

A matrix $A$ is in row canonical form if it satisfies the following conditions.
(1) Nonzero rows of $A$ are above any zero rows.
(2) If $A$ has two pivots of the same degree, the one that occurs earlier is above the one that occurs later. If $A$ has two pivots of different degree, the one with smaller degree is above the one with larger degree.
(3) Every pivot is of the form $\pi^{\ell}$ for some $\ell \in\{0, \ldots, s-1\}$.
(4) For every pivot (say $\pi^{\ell}$ ), all entries below and in the same column as the pivot are zero, and all entries above and in the same column as the pivot are elements of $\mathcal{R}\left(R, \pi^{\ell}\right)$.

For example, over $\mathbb{Z}_{8}$,

$$
A=\left[\begin{array}{llll}
0 & 2 & 0 & \overline{1} \\
\overline{2} & 2 & 0 & 0 \\
0 & 0 & \overline{2} & 0 \\
0 & \overline{4} & 0 & 0 \\
0 & 0 & 0 & 0
\end{array}\right]
$$

is in row
canonical form.

## Basic Facts

Let $A \in R^{n \times m}$ be a matrix in row canonical form, let $p_{k}$ be the pivot of the $k$ th row, let $c_{k}$ be the index of the column containing $p_{k}$. (If the $k$ th row is zero, let $p_{k}=0$ and $c_{k}=0$.) Let $d_{k}=\operatorname{deg}\left(p_{k}\right)$, and let $w=\left(w_{1}, \ldots, w_{m}\right)$ be an arbitrary element of row $A$.
(1) Any column of $A$ contains at most one pivot.
(2) If $A$ has more than one row, deleting a row of $A$ results in a matrix also in row canonical form.
(3) $i \geq k$ implies $\operatorname{deg}(A[i, j]) \geq d_{k}$.
(4) $\left(i \geq k\right.$ and $\left.j<c_{k}\right)$ or $\left(i>k\right.$ and $\left.j \leq c_{k}\right)$ implies $\operatorname{deg}(A[i, j])>d_{k}$.
(5) $p_{1}$ divides $w_{1}, w_{2}, \ldots, w_{m}$.
(6) $j<c_{1}$ implies $\operatorname{deg}\left(w_{j}\right)>d_{1}$.

## Reduction to Row Canonical Form

PivotSelection: given a submatrix, return the row and column index of the earliest occurring pivot of least possible degree; otherwise declare the submatrix to be zero.
Given a matrix $A$ :

- Step $k=1$ : apply PivotSelection to $A$; move the selected row to row 1 , normalize (make sure the first pivot is of the form $\pi^{\ell}$ ), and cancel all elements below the pivot (which must all be multiples of the first pivot). Call the resulting matrix $A_{1}$, and increment $k$.
- For $k \geq 2$, apply PivotSelection to the rows of $A_{k-1}$, excluding the first $k-1$ rows. If no pivot can be found, stop; otherwise, move the selected row to row $k$, normalize to $\pi^{\ell}$, cancel all elements below the pivot, and reduce all elements above the pivot to elements of $\mathcal{R}\left(R, \pi^{\ell}\right)$. Call the resulting matrix $A_{k}$, and increment $k$.


## Row Canonical Form (cont'd)

## Theorem

For any $A \in R^{n \times m}$, the algorithm described above computes a row canonical form of $A$.

## Theorem

For any $A \in R^{n \times m}$, the row canonical form of $A$ is unique.

## Example:

$$
\begin{aligned}
A & =\left[\begin{array}{llll}
4 & 6 & 2 & \overline{1} \\
0 & 0 & 0 & 2 \\
2 & 4 & 6 & 1 \\
2 & 0 & 2 & 1
\end{array}\right] \rightarrow A_{1}=\left[\begin{array}{llll}
4 & 6 & 2 & 1 \\
0 & 4 & 4 & 0 \\
\overline{6} & 6 & 4 & 0 \\
6 & 2 & 0 & 0
\end{array}\right] \rightarrow \\
A_{1}^{\prime} & =\left[\begin{array}{llll}
4 & 6 & 2 & 1 \\
\overline{2} & 2 & 4 & 0 \\
0 & 4 & 4 & 0 \\
6 & 2 & 0 & 0
\end{array}\right] \rightarrow A_{2}=\left[\begin{array}{llll}
0 & 2 & 2 & 1 \\
2 & 2 & 4 & 0 \\
0 & \overline{4} & 4 & 0 \\
0 & 4 & 4 & 0
\end{array}\right] \rightarrow \\
A_{3} & =\left[\begin{array}{llll}
0 & 2 & 2 & \overline{1} \\
\overline{2} & 2 & 4 & 0 \\
0 & \overline{4} & 4 & 0 \\
0 & 0 & 0 & 0
\end{array}\right] \quad \text { which is in row canonical form. }
\end{aligned}
$$

## Matrix Shape via Row Canonical Form

Let $B$ be the row canonical form of $A \in R^{n \times m}$ with $k$ nonzero rows. Let $p_{i}$ be the pivot in the $i$ th row of $B$, where $i \in\{1, \ldots, k\}$. Let $r=\min \{n, m\}$. Clearly, $k \leq r$. Then the Smith normal form of $A$ is given by

$$
\operatorname{diag}(p_{1}, \ldots, p_{k}, \underbrace{0, \ldots, 0}_{r-k}) \in R^{n \times m},
$$

from which the shape of $A$ is readily available.
Example:

$$
A=\left[\begin{array}{llll}
4 & 6 & 2 & 1 \\
0 & 0 & 0 & 2 \\
2 & 4 & 6 & 1 \\
2 & 0 & 2 & 1
\end{array}\right] \rightarrow B=\left[\begin{array}{llll}
0 & 2 & 2 & 1 \\
2 & 2 & 4 & 0 \\
0 & 4 & 4 & 0 \\
0 & 0 & 0 & 0
\end{array}\right]
$$

over $\mathbb{Z}_{8}$. Since $B$ is the row canonical form of $A$, we see that the Smith normal form is $\operatorname{diag}(1,2,4,0)$, and hence shape $A=(1,2,3)$.

## $\pi$-adic Decomposition

Let $R^{n \times \mu}$ denote the set of matrices in $R^{n \times m}$ whose rows are elements of $R^{\mu}$. Every matrix $X$ in $R^{n \times \mu}$ decomposes according to its $\pi$-adic decomposition as

$$
X=X_{0}+\pi X_{1}+\cdots+\pi^{s-1} X_{s-1}
$$

with each auxiliary matrix $X_{i}$
( $i=0, \ldots, s-1$ ) satisfying:
(1) $X_{i}\left[1: n, 1: \mu_{i+1}\right]$ is an arbitrary matrix over $\mathcal{R}(R, \pi)$, and
(2) all other entries in $X_{i}$ are zero.

Example: $n=6, \mu=(4,6,8)$.





## Row Canonical Forms in $\mathcal{T}_{k}\left(R^{n \times \mu}\right)$

Let $\mathcal{T}_{\kappa}\left(R^{n \times \mu}\right)$ denote the set of matrices in $R^{n \times \mu}$ whose shape is $\kappa$, where $\kappa \preceq n$ and $\kappa \preceq \mu$.
The row canonical forms in $\mathcal{T}_{\kappa}\left(R^{n \times \mu}\right)$ are in one-to-one correspondence with the submodules of $R^{\mu}$ having shape $\kappa$; thus there are $\left[\begin{array}{c}\mu \\ \kappa\end{array}\right]_{q}$ such row canonical forms.

## Example:

Let $R=\mathbb{Z}_{4}$, and let $n=2, \mu=(2,3), \kappa=(1,2)$. Then $\left[\begin{array}{c}\mu \\ \kappa\end{array}\right]_{q}=18$.
These 18 row canonical forms can be classified into 4 categories based on the positions of their pivots:


Clearly, the first category, whose pivots occur as early as possible, contains a significant portion of all possible row canonical forms.

## Principal RCFs - The "Thick Cell"

## Definition

A row canonical form in $\mathcal{T}_{\kappa}\left(R^{n \times \mu}\right)$ is called principal if its diagonal entries $d_{1}, d_{2}, \ldots, d_{r}(r=\min \{n, m\})$ have the following form:


All principal RCFs in $\mathcal{T}_{\kappa}\left(R^{n \times \mu}\right)$ can be constructed via a $\pi$-adic decomposition:


Illustration of the construction of principal row canonical forms for $\mathcal{T}_{\kappa}\left(R^{n \times \mu}\right)$ with $s=3, n=6, \mu=(4,6,8)$, and $\kappa=(2,3,4)$.

## Counting Principal RCFs in $\mathcal{T}_{\kappa}\left(R^{n \times \mu}\right)$

Note that the number of principal row canonical forms in $\mathcal{T}_{\kappa}\left(R^{n \times \mu}\right)$ is

$$
P_{q}(\mu, \kappa)=q^{\sum_{i=1}^{s} \kappa_{i}\left(\mu_{i}-\kappa_{i}\right)}
$$

The number of row canonical forms in $\mathcal{T}_{\kappa}\left(R^{n \times \mu}\right)$ in total is

$$
\left.\llbracket \begin{array}{l}
\mu \\
\kappa
\end{array}\right]_{q}=\prod_{i=1}^{s} q^{\left(\mu_{i}-\kappa_{i}\right) \kappa_{i-1}}\left[\begin{array}{l}
\mu_{i}-\kappa_{i-1} \\
\kappa_{i}-\kappa_{i-1}
\end{array}\right]_{q}
$$

Since $q^{k(m-k)} \leq\left[\begin{array}{c}m \\ k\end{array}\right]_{q}<4 q^{k(m-k)}$, we have

$$
1 \leq \frac{\llbracket \begin{array}{c}
\mu \\
\kappa
\end{array} \rrbracket_{q}}{P_{q}(\mu, \kappa)}<4^{s}
$$

i.e., the number of principal RCFs in $\mathcal{T}\left(R^{n \times \mu}\right)$ grows at the same rate as the number of RCFs in total.

## Counting All Matrices in $\mathcal{T}_{\kappa}\left(R^{n \times \mu}\right)$

We can partition the matrices in $\mathcal{T}_{\kappa}\left(R^{n \times \mu}\right)$ based on their row canonical forms: two matrices belong to the same class if and only if they have the same row canonical form.

- The number of classes is $\left[\begin{array}{c}\mu \\ \kappa\end{array} \rrbracket_{q}\right.$.
- The number of matrices in each class is

$$
\left|R^{n \times \kappa}\right| \prod_{i=0}^{\kappa_{s}-1}\left(1-q^{i-n}\right)=q^{n|\kappa|} \prod_{i=0}^{\kappa_{s}-1}\left(1-q^{i-n}\right)=
$$

- It follows that

$$
\left|\mathcal{T}_{\kappa}\left(R^{n \times \mu}\right)\right|=q^{n|\kappa|} \prod_{i=0}^{\kappa_{s}-1}\left(1-q^{i-n}\right) \llbracket\left[\begin{array}{l}
\mu \\
\kappa
\end{array} \rrbracket_{q}\right.
$$

## Matrix Channels over Finite Chain Rings

Let $R$ be a ( $q, s$ ) chain ring, and let $\mu$ be an $s$-shape. We think of $R^{\mu}$ as the "packet space" associated with a network. The length, $m$, of each packet is given by $\mu_{s}$.

- The transmitter sends $n$ packets, each constrained to be an element of $R^{\mu}$. These form the rows of the transmitted matrix $X \in R^{n \times \mu}$.
- The receiver gathers $N$ packets, each also an element of $R^{\mu}$. These form the rows of the received matrix $Y \in R^{N \times \mu}$.
- Noise is modelled by the injection of $t$ packets into the network, each also an element of $R^{\mu}$. These form the rows of the noise matrix $Z \in R^{t \times \mu}$.
- In general, we have


$$
Y=A X+B Z
$$

B
for some transfer matrices $A \in R^{N \times n}$ and $B \in R^{N \times t}$.

## Capacity of Matrix Channels over Finite Chain Rings

Our model is $Y=A X+B Z$.

- A well-defined discrete memoryless channel with input alphabet $R^{n \times \mu}$, output alphabet $R^{N \times \mu}$ and channel transition probability $p_{Y \mid X}$ is obtained once a joint distribution for $p_{Z, A, B \mid X}$ is specified.
- The capacity of this channel is given, as usual, by

$$
C=\max _{p_{X}} I(X ; Y)
$$

where $p_{X}$ is the input distribution. (We will take logarithms to base $q$, so the capacity is given in $q$-qary symbols per channel use.)

## Asymptotic Capacity, $\bar{C}$

How does capacity scale with packet length?
Given a channel with a given $n, N, \mu$, and $t$, we define the $k$ th extension as the channel in which the transmitter sends kn packets of shape $k \mu$, the receiver gather $k N$ packets of this shape, the noise matrix has $k t$ rows, and the channel law is suitably generalized, giving capacity $C_{k}$.

## Definition

We define the asymptotic capacity as

$$
\bar{C}=\lim _{k \rightarrow \infty} \frac{1}{(k n)|k \mu|} C_{k}=\frac{1}{n|\mu|} \lim _{k \rightarrow \infty} \frac{C_{k}}{k^{2}} .
$$

Note that $\bar{C}$ is normalized such that $\bar{C}=1$ if the channel is noiseless (i.e., $A=I$ and $Z=0$ ).

## The Independent Transfer Model

Let $\tau$ be an $s$-shape such that $\tau \preceq t, \mu$.
We study the case where:

- the transfer matrix $A$ is uniform over $\mathrm{GL}_{n}(R)$ (in particular,

$$
N=n),
$$

- $B$ is uniform over $\mathcal{T}_{t}\left(R^{n \times t}\right)$,
- $Z$ is uniform over $\mathcal{T}_{\tau}\left(R^{t \times \mu}\right)$,
- $X, A, B$ and $Z$ are statistically independent.

In this case we can re-write the channel model as

$$
Y=A\left(X+A^{-1} B Z\right)=A(X+W),
$$

where $A \in \mathrm{GL}_{n}(R)$ and $W \triangleq A^{-1} B Z \in \mathcal{T}_{\tau}\left(R^{n \times \mu}\right)$ are chosen
 uniformly at random and independently from any other variables.

## MMC: Model

## First warmup problem

The multiplicative matrix channel (MMC):


$$
Y=A X
$$

where

- $X, Y \in R^{n \times \mu}$;
- $A \sim \operatorname{Unif}\left[G L_{n}(R)\right] ;$
- $A$ and $X$ are independent.


## MMC: Exact Capacity

It is easy to find the capacity of a channel defined by a group action.

- Let $\mathcal{G}$ be a finite group that acts on a finite set $\mathcal{S}$.
- Consider a channel with input $X \in \mathcal{S}$, output $Y \in \mathcal{S}$ and channel law $Y=A X$, where $A \sim \operatorname{Unif}[\mathcal{G}]$ and $A$ and $X$ are independent.
- The capacity of this channel is

$$
C=\log |\mathcal{S} / \mathcal{G}|
$$

where $|\mathcal{S} / \mathcal{G}|$ is the number of orbits of the action.

- One capacity-achieving input distribution is to sample uniformly over a complete system of orbit-representatives.


## MMC: Exact Capacity

In the case of the MMC,

- $\mathrm{GL}_{n}(R)$ acts on $R^{n \times \mu}$ by left-multiplication.
- The orbits are the sets of matrices that share the same row module.
- The number of such orbits is the number of submodules of $R^{\mu}$ with shape $\preceq n, \mu$.


## Theorem

The capacity of the MMC, in q-ary symbols per channel use, is given by

$$
C_{M M C}=\log _{q} \sum_{\lambda \preceq n, \mu} \llbracket\left[\begin{array}{l}
\mu \\
\lambda
\end{array}\right]_{q} .
$$

A capacity-achieving code $\mathcal{C} \subseteq R^{n \times \mu}$ consists of all possible row canonical forms in $R^{n \times \mu}$.
(This scheme encodes information in the choice of submodules, generalizing the "transmission via subspaces" approach of [KK08].)

## MMC: Asymptotic Capacity

The capacity $C_{\mathrm{MMC}}$ is bounded by

$$
\begin{equation*}
\sum_{i=1}^{s} \kappa_{i}\left(\mu_{i}-\kappa_{i}\right) \leq C_{\mathrm{MMC}} \leq \sum_{i=1}^{s} \kappa_{i}\left(\mu_{i}-\kappa_{i}\right)+\log _{q} 4^{s}\binom{n+s}{s} \tag{2}
\end{equation*}
$$

where $\kappa_{i}=\min \left\{n,\left\lfloor\mu_{i} / 2\right\rfloor\right\}$.

## Theorem

$$
\bar{C}_{M M C}=\frac{\sum_{i=1}^{s} \kappa_{i}\left(\mu_{i}-\kappa_{i}\right)}{n|\mu|}
$$

where $\kappa_{i}=\min \left\{n,\left\lfloor\mu_{i} / 2\right\rfloor\right\}$.
The choice of subshape $\kappa$ essentially maximizes the number of principal row canonical forms having fixed subshape.
Thus, asymptotically, capacity can be achieved by always transmitting principal row canonical forms with a fixed subshape!

## MMC: Encoding and Decoding

Let $\kappa=\left(\kappa_{1}, \ldots, \kappa_{s}\right)$ with $\kappa_{i}=\min \left\{n,\left\lfloor\mu_{i} / 2\right\rfloor\right\}$.

- Encoding: choose the input matrix $X$ from the set of principal RCFS for $\mathcal{T}_{\kappa}\left(R^{n \times \mu}\right)$ using the $\pi$-adic decomposition given earlier. The encoding rate is

$$
R_{\mathrm{MMC}}=\sum_{i=1}^{s} \kappa_{i}\left(\mu_{i}-\kappa_{i}\right)
$$

- Decoding: upon receiving $Y=A X$, the decoder simply computes the row canonical form of $Y$. The decoding is always correct by the uniqueness of the row canonical form.
This coding scheme achieves the asymptotic capacity $\bar{C}_{\text {MMC }}$.


## AMC: Model

## Second warmup problem

The additive matrix channel (AMC):


$$
Y=X+W
$$

where

- $X, Y \in R^{n \times \mu}$;
- $W \sim \operatorname{Unif}\left[\mathcal{T}_{\tau}\left(R^{n \times \mu}\right)\right]$;
- $W$ and $X$ are independent.


## AMC: Exact Capacity

The AMC is an example of a discrete symmetric channel.
Theorem
The capacity of the AMC, in q-ary symbols per channel use, is given by

$$
C_{A M C}=\log _{q}\left|R^{n \times \mu}\right|-\log _{q}\left|\mathcal{T}_{\tau}\left(R^{n \times \mu}\right)\right|,
$$

achieved by the uniform input distribution.

## AMC: Asymptotic Capacity

The capacity $C_{\text {AMC }}$ is bounded by

$$
\begin{gathered}
\sum_{i=1}^{s}\left(n-\tau_{i}\right)\left(\mu_{i}-\tau_{i}\right)-\log _{q} 4^{s} \prod_{i=0}^{\tau_{s}-1}\left(1-q^{i-n}\right) \\
<C_{\mathrm{AMC}}< \\
\sum_{i=1}^{s}\left(n-\tau_{i}\right)\left(\mu_{i}-\tau_{i}\right)-\log _{q} \prod_{i=0}^{\tau_{s}-1}\left(1-q^{i-n}\right)
\end{gathered}
$$

## Theorem

The asymptotic capacity $\bar{C}_{A M C}$ is given by

$$
\bar{C}_{A M C}=\frac{\sum_{i=1}^{s}\left(n-\tau_{i}\right)\left(\mu_{i}-\tau_{i}\right)}{n|\mu|} .
$$

## AMC: Error-trapping Encoding

We focus on the special case when $\tau=t=(t, \ldots, t)$. Set $v \geq t$ and transmit a matrix $X$ of the form

$$
X=\left[\begin{array}{cc}
0 & 0 \\
0 & U_{(n-v) \times(m-v)}
\end{array}\right]
$$



Clearly

$$
R_{\mathrm{AMC}}=\sum_{i=1}^{s}(n-v)\left(\mu_{i}-v\right)
$$

## AMC: Error-trapping Decoding

Write

$$
W=\left[\begin{array}{ll}
W_{1} & W_{2} \\
W_{3} & W_{4}
\end{array}\right]
$$

Suppose shape $W_{1}=t$. Then, since shape $W=t$ also, the pivots of $W$ are entirely contained in $W_{1}$. Since the row canonical form of $W$ has $t$ nonzero rows, this means that the upper rows of $W$ can cancel the lower rows, i.e., for some matrix $V$ we have

$$
\left[\begin{array}{ll}
l & 0 \\
V & I
\end{array}\right]\left[\begin{array}{ll}
W_{1} & W_{2} \\
W_{3} & W_{4}
\end{array}\right]=\left[\begin{array}{cc}
W_{1} & W_{2} \\
0 & 0
\end{array}\right]
$$

Indeed, $V$ can be chosen so that $V W_{1}=-W_{3}$, which automatically forces $V W_{2}=-W_{4}$ (since if $V W_{2}+W_{4} \neq 0, W$ would have pivots outside of $W_{1}$ ).
Applying this transformation to $Y=X+W$ yields

$$
\left[\begin{array}{ll}
l & 0 \\
V & 1
\end{array}\right]\left[\begin{array}{lc}
W_{1} & W_{2} \\
W_{3} & U+W_{4}
\end{array}\right]=\left[\begin{array}{cc}
W_{1} & W_{2} \\
0 & U
\end{array}\right],
$$

exposing the user's data matrix $U$.

## AMC: Error-trapping Decoding (cont'd)

In summary:

- The decoder observes $W_{1}, W_{2}$, and $W_{3}$ thanks to the error traps.
- If shape $W_{1}=t$, then the decoder applies the transformation on the previous slide to expose $U$.
- If shape $W_{1} \neq t$, a decoding failure (detected error) is declared. The probability of decoding failure $P_{f}=P$ shape $\left.W_{1} \neq t\right]$ is bounded as

$$
P_{f}<\frac{2 t}{q^{1+v-t}}
$$

If we set $v$ such that $v-t \rightarrow \infty$, and $\frac{v-t}{m} \rightarrow 0$, as $m \rightarrow \infty$, then we have $P_{f} \rightarrow 0$ and $\bar{R}_{\mathrm{AMC}}=\frac{R_{\mathrm{AMC}}}{n|\mu|} \rightarrow \bar{C}_{\mathrm{AMC}}$.

## Theorem

This coding scheme can achieve the asymptotic capacity of the AMC for the special case when $\tau=t$.

## AMMC: Model

## Now to the main event:

The additive-multiplicative matrix channel (AMMC):


$$
Y=A(X+W)
$$

where

- $X, Y \in R^{n \times \mu}$;
- $W \sim \operatorname{Unif}\left[\mathcal{T}_{\tau}\left(R^{n \times \mu}\right)\right]$;
- $A \sim \operatorname{Unif}\left[\mathrm{GL}_{n}(R)\right]$;
- $A, X$ and $W$ are independent.

Remark: This model is statistically identical to $Y=A X+Z$, where $Z \sim \operatorname{Unif}\left[\mathcal{T}_{\tau}\left(R^{n \times \mu}\right)\right]$

## AMMC: Upper Bound on Capacity

## Theorem

The capacity of the AMMC, in q-ary symbols per channel use, is upper-bounded by

$$
\begin{aligned}
& C_{A M M C} \leq \sum_{i=1}^{s}\left(\mu_{i}-\xi_{i}\right) \xi_{i}+\sum_{i=1}^{s}\left(n-\mu_{i}\right) \tau_{i}+2 s \log _{q} 4+\log _{q}\binom{n+s}{s} \\
& +\log _{q}\binom{\tau_{s}+s}{s}-\log _{q} \prod_{i=0}^{\tau_{s}-1}\left(1-q^{i-n}\right), \text { where } \xi_{i}=\min \left\{n,\left\lfloor\mu_{i} / 2\right\rfloor\right\}
\end{aligned}
$$

In particular, when $\mu \succeq 2 n$, the upper bound reduces to

$$
\begin{aligned}
& C_{\mathrm{AMMC}} \leq \sum_{i=1}^{s}\left(n-\tau_{i}\right)\left(\mu_{i}-n\right)+2 s \log _{q} 4 \\
&+\log _{q}\binom{n+s}{s}+\log _{q}\binom{\tau_{s}+s}{s}-\log _{q} \prod_{i=0}^{\tau_{s}-1}\left(1-q^{i-n}\right)
\end{aligned}
$$

## AMMC: Asymptotic Capacity

## Theorem

When $\mu \succeq 2 n$, the asymptotic capacity $\bar{C}_{A M M C}$ is upper-bounded by

$$
\bar{C}_{A M M C} \leq \frac{\sum_{i=1}^{s}\left(n-\tau_{i}\right)\left(\mu_{i}-n\right)}{n|\mu|}
$$

## AMMC: Coding Scheme

We again focus on the special case when $\tau=t$, and combine the two strategies for the MMC and the AMC.
To encode, construct $X$ as

$$
X=\left[\begin{array}{ll}
0 & 0 \\
0 & \bar{X}
\end{array}\right]
$$

where $\bar{X}$ is chosen from the set of principal row canonical forms for $\mathcal{T}_{\kappa}\left(R^{(n-v) \times(\mu-v)}\right)$ by the previous construction.


We have $R_{\text {AMMC }}=\sum_{i=1}^{s} \kappa_{i}\left(\mu_{i}-v-\kappa_{i}\right)$. In particular, when $\mu \succeq 2 n$, we have $\left\lfloor\left(\mu_{i}-v\right) / 2\right\rfloor \geq n-v$ for all $i$. Thus, $\kappa_{i}=n-v$ for all $i$, and the encoding rate is $R_{\text {AMMC }}=\sum_{i=1}^{s}(n-v)\left(\mu_{i}-n\right)$.

## AMMC: Coding Scheme (cont'd)

To decode, we must recover $\bar{X}$ from $Y=A(X+W)$.
If we had $X+W$, we could use the error-trapping decoder to recover

$$
\left[\begin{array}{cc}
W_{1} & W_{2} \\
0 & \bar{X}
\end{array}\right]
$$

But we have $Y$, not $X+W$. However, since $A$ is invertible, $\operatorname{RCF}(Y)=\operatorname{RCF}(X+W)$, and one easily sees that

$$
\operatorname{RCF}(X+W)=\left[\begin{array}{cc}
\bar{W}_{1} & \bar{W}_{2} \\
0 & \bar{X} \\
0 & 0
\end{array}\right]
$$

where the bottom $v-t$ rows are all zero.
In summary:

- The decoder first computes $\operatorname{RCF}(Y)$.
- It then checks the condition shape $W_{1}=t$.
- If the condition does not hold, a decoding failure is declared, otherwise the decoder outputs $\bar{X}$.


## Conclusions

- Nested-lattice-based physical layer network coding naturally transforms wireless multiple-access channels with random fading into random linear network coding channels.
- The algebraic structure of $\Lambda / \Lambda^{\prime}$ is that of a module over a ring.
- In many cases, the ring is a finite-chain ring, so end-to-end error control (for random errors) can be handled using a matrix-channel approach, with simple and asymptotically efficient coding schemes.


## Open Problems

- Relaxing the assumption on $A$
- What if $A$ is not invertible?
- Relaxing the assumption on $W$
- What if $W$ has shape other than $\tau=t$ ?
- Adversarial error models
- Always correcting errors when shape $(W) \leq \tau$ ?
- Rank-metric codes over finite chain rings
- Which properties can be preserved?


## Backup Slides

## Physical-Layer Network Coding



Motivation

## Current Wireless



## Current Wireless

## Router



## Current Wireless



## Current Wireless



## Current Wireless


$\pi$


## Current Wireless



## Current Wireless



## Current Wireless

## Router



Routing requires 4 time slots

## Network Coding



## Network Coding



## Network Coding



## Network Coding



## Network Coding

## Router



## Network Coding

## Router



Network coding requires 3 time slots

## Network Coding

## Router



Network coding requires 3 time slots. Can we do better?

## Physical-Layer Network Coding

## Router



## Physical-Layer Network Coding



## Physical-Layer Network Coding



## Physical-Layer Network Coding

## Router



Physical-layer network coding requires 2 time slots

## It Is More Than Going From 3 to 2

- A new way of dealing with interference process interference instead of avoiding it
- Can be extended to large networks each relay infers some linear combination

$$
\begin{aligned}
& \theta \\
& \bullet \\
& \bullet \Leftrightarrow
\end{aligned}
$$

## Extension to Large Wireless Networks



## Extension to Large Wireless Networks



## Extension to Large Wireless Networks



## Extension to Large Wireless Networks



## Extension to Large Wireless Networks



## Extension to Large Wireless Networks



Part 1: First PNC schemes

## Example 1 (BPSK, $h_{1}=h_{2}=1$ )

Zhang-Liew-Lam 2006, Popovski-Yomo 2006


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## Example 2 (QPSK, $h_{1} \approx h_{2}$ )

Zhang-Liew-Lam 2006, Popovski-Yomo 2006

| 01 | 11 | $\bigcirc_{(01,01)}$ | $(01,11) \bigcirc$ | ${ }^{(11,01)}$ | $(11,11) \bigcirc$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 0 | 0 |  |  |  |  |
|  |  |  |  |  |  |
|  |  |  |  |  |  |
| 00 | 10 |  |  |  |  |
| O | $\bigcirc$ | $\bigcirc(01,00)$ | $(01,10)$ | $\bigcirc(11,00)$ | $(11,10)$ |
| 01 | 11 | $(00,01)$ | $(00,11) \bigcirc$ | $(10,01)$ | $(10,11) \bigcirc$ |
| 0 | $\bigcirc$ |  |  |  |  |
|  |  |  |  |  |  |
|  |  |  |  |  |  |
| 00 | 10 |  |  |  |  |
| $\bigcirc$ | $\bigcirc$ |  |  |  |  |
|  |  | $\bigcirc(00,00)$ | $(00,10)$ | O (10,00) | $(10,10)$ |
|  |  | - | O | - | O |

## Example 2 (QPSK, $h_{1} \approx h_{2}$ )

Zhang-Liew-Lam 2006, Popovski-Yomo 2006


## Example 2 (QPSK, $h_{1} \approx h_{2}$ )

Zhang-Liew-Lam 2006, Popovski-Yomo 2006


## Example 3 (QPSK, $h_{1} \approx i h_{2}$ )

Zhang-Liew-Lam 2006, Popovski-Yomo 2006


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Zhang-Liew-Lam 2006, Popovski-Yomo 2006


## Limitation of Original PNC Schemes

Limitation: phase misalignment $\Rightarrow$ bad performance
Solution 1: require phase synchronization
Solution 2: mitigate phase misalignment by moving beyond XOR

- Popovski \& Yomo 2007
- Koike-Akino-Popovski-Tarokh 2008

Solution 3: mitigate phase misalignment by compute-and-forward

- Nazer \& Gastpar 2007


## Example 4 (QPSK, $h_{1} \approx i h_{2}$ )

Koike-Akino-Popovski-Tarokh: $(a b, c d) \rightarrow a b \oplus d c$

| $\begin{aligned} & 01 \\ & 0 \end{aligned}$ | $\begin{aligned} & 11 \\ & 0 \end{aligned}$ | $(11,01)$ | $(11,11) \bigcirc$ | $(10,01)$ | $(10,11) \bigcirc$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |
|  |  |  |  |  |  |
|  |  |  |  |  |  |
| $\begin{gathered} 00 \\ 0 \end{gathered}$ | $\begin{aligned} & 10 \\ & 0 \end{aligned}$ |  |  |  |  |
|  |  | O (11,00) | $(11,10)$ | $\bigcirc(10,00)$ | $(10,10)$ |
| $\begin{aligned} & 01 \\ & 0 \end{aligned}$ | $\begin{aligned} & 11 \\ & 0 \end{aligned}$ | $(01,01)$ | $(01,11) \bigcirc$ | $(00,01)$ | $(00,11) \bigcirc$ |
|  |  |  |  |  |  |
|  |  |  |  |  |  |
| $\begin{gathered} 00 \\ 0 \end{gathered}$ | 10 |  |  |  |  |
|  | $\bigcirc$ |  |  |  |  |
|  |  | $\bigcirc(01,00)$ | $(01,10)$ | $O^{(00,00)}$ | $(00,10)$ |

## Example 4 (QPSK, $h_{1} \approx i h_{2}$ )

Koike-Akino-Popovski-Tarokh: $(a b, c d) \rightarrow a b \oplus d c$


Part 2: Compute-and-Forward

## Compute-and-Forward Relaying Strategy

## Nazer \& Gastpar's Approach (2006)

- Voronoi constellations based on Erez-Zamir's construction
- Main result: achievable rates for one-hop networks
- CSI only at the receivers but not at the transmitters


## Similar Approaches

- Narayanan-Wilson-Sprintson (2007)
- Nam-Chung-Lee (2008)
- Wilson-Narayanan (2009)

Remark: all of these are based on Erez-Zamir's construction of Voronoi constellations

## Voronoi Constellations in One Slide

pick a fine lattice $\Lambda$

## Voronoi Constellations in One Slide

pick a coarse lattice $\Lambda^{\prime}$

## Voronoi Constellations in One Slide



Voronoi region for the fine lattice $\Lambda$

## Voronoi Constellations in One Slide



Voronoi region for the coarse lattice $\Lambda^{\prime}$

## Key Idea: the Case of Integer Channel Gains



Each transmitter applies the same Voronoi constellation $\Lambda / \Lambda^{\prime}$

## Key Idea: the Case of Integer Channel Gains


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Transmitter 1 maps $\mathbf{w}_{1}$ to a constellation point

## Key Idea: the Case of Integer Channel Gains



Transmitter 2 maps $\mathbf{w}_{2}$ to a constellation point

## Key Idea: the Case of Integer Channel Gains



$$
\boldsymbol{y}=h_{1} \boldsymbol{x}_{1}+h_{2} \boldsymbol{x}_{2}+\boldsymbol{z}
$$

$$
\left(h_{1}, h_{2}\right)=(2,1)
$$



The channel is given by $\mathbf{y}=2 \mathbf{x}_{1}+\mathbf{x}_{2}+\mathbf{z}$

## Key Idea: the Case of Integer Channel Gains



$$
\boldsymbol{y}=h_{1} \boldsymbol{x}_{1}+h_{2} \boldsymbol{x}_{2}+\boldsymbol{z}
$$

$$
\left(h_{1}, h_{2}\right)=(2,1)
$$



Hence, the received signal $\mathbf{y}$ is like this

## Key Idea: the Case of Integer Channel Gains



$$
\boldsymbol{y}=h_{1} \boldsymbol{x}_{1}+h_{2} \boldsymbol{x}_{2}+\boldsymbol{z}
$$

$$
\left(h_{1}, h_{2}\right)=(2,1)
$$



But how can we extract some information from $\mathbf{y}$ ?

## Key Idea: the Case of Integer Channel Gains



$$
\boldsymbol{y}=h_{1} \boldsymbol{x}_{1}+h_{2} \boldsymbol{x}_{2}+\boldsymbol{z}
$$

$$
\left(h_{1}, h_{2}\right)=(2,1)
$$



The receiver can decode $2 \mathbf{x}_{1}+\mathbf{x}_{2}$, if the noise $\mathbf{z}$ is small

## Key Idea: the Case of Integer Channel Gains



$$
\boldsymbol{y}=h_{1} \boldsymbol{x}_{1}+h_{2} \boldsymbol{x}_{2}+\boldsymbol{z}
$$

$$
\left(h_{1}, h_{2}\right)=(2,1)
$$



But we need a linear combination of messages...

## Key Idea: the Case of Integer Channel Gains

## Note that...

one can construct a one-to-one linear map between $\mathbb{F}_{3}^{2}$ and $\Lambda / \Lambda^{\prime}$


$$
\begin{gathered}
\boldsymbol{w}_{1}=(1,0) \\
\boldsymbol{w}_{2}=(0,1) \\
\hat{\boldsymbol{u}}=(2,1) \\
\hat{\boldsymbol{u}}=2 \boldsymbol{w}_{1}+\boldsymbol{w}_{2}
\end{gathered}
$$

Hence, an integer combination of lattice points
$=$ a linear combination of messages

## Key Idea: the Case of Real Channel Gains

What if channel gains are real numbers?
Applying a scaling operation $g(\mathbf{y})=\alpha \mathbf{y}$

$$
\begin{aligned}
\alpha \mathbf{y} & =\sum_{\ell} \alpha h_{\ell} \mathbf{x}_{\ell}+\alpha \mathbf{z} \\
& =\sum_{\ell} a_{\ell} \mathbf{x}_{\ell}+\underbrace{\sum_{\ell}\left(\alpha h_{\ell}-a_{\ell}\right) \mathbf{x}_{\ell}+\alpha \mathbf{z}}_{\text {effective noise }} \\
& =\sum_{\ell} a_{\ell} \mathbf{x}_{\ell}+\mathbf{n},
\end{aligned}
$$

where $\left\{a_{\ell}\right\}$ are integers, and $\alpha \in \mathbb{R}$ is the scalar.
Thus, real-valued channel gains $\Rightarrow$ integer channel gains

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where $\left\{a_{\ell}\right\}$ are integers, and $\alpha \in \mathbb{R}$ is the scalar.
Thus, real-valued channel gains $\Rightarrow$ integer channel gains

## But how shall we choose the scalar $\alpha$ ?

see Nazer-Gastpar (IEEE Trans. Info. Theory, 2011) for details

## Key Idea: the Case of Complex Channel Gains

What if channel gains are complex numbers?
Answer: lift real lattices to Gaussian lattices
Gaussian integers: $\mathbb{Z}[i]=\{a+b i: a, b \in \mathbb{Z}\}$
Gaussian lattices: any Gaussian integer combination of lattice points is a lattice point
Then, apply a scaling operation $g(\mathbf{y})=\alpha \mathbf{y}$

$$
\begin{aligned}
\alpha \mathbf{y} & =\sum_{\ell} \alpha h_{\ell} \mathbf{x}_{\ell}+\alpha \mathbf{z} \\
& =\sum_{\ell} a_{\ell} \mathbf{x}_{\ell}+\underbrace{\sum_{\ell}\left(\alpha h_{\ell}-a_{\ell}\right) \mathbf{x}_{\ell}+\alpha \mathbf{z}}_{\text {effective noise }} \\
& =\sum_{\ell} a_{\ell} \mathbf{x}_{\ell}+\mathbf{n},
\end{aligned}
$$

where $\left\{a_{\ell}\right\}$ are Gaussian integers, and $\alpha \in \mathbb{C}$ is the scalar

## Main Result: Computation Rate



## Computation Rate (Nazer-Gastpar)

$$
R_{\mathrm{comp}}=\log _{2}\left(\frac{P}{P \sum_{\ell}\left\|\alpha h_{\ell}-a_{\ell}\right\|^{2}+N_{0}|\alpha|^{2}}\right)
$$

Remark: Erez-Zamir's construction of Voronoi constellations $\Rightarrow$ asymptotically long block length and almost unbounded complexity

## Research Problems

What if practical Voronoi constellations are used?

## Goal: Practical Design for Compute-and-Forward

- Short block length and low complexity
- Example: wireless fading channel with short coherent time


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## Related Work

- Ordentlich \& Erez (2010)
- Hern \& Narayanan (2011)
- Tunali \& Narayanan (2011)
- Ordentlich-Zhan-Erez-Gastpar-Nazer (2011)
- Feng-Silva-Kschischang (2011)
- Emerging work includes Osmane \& Belfiore (in submission)

Part 3: An Algebraic Approach

## Algebraic Approach: Key Elements

## $R$-Lattices

Let $R$ be a discrete subring of $\mathbb{C}$ forming a principal ideal domain. Let $N \leq n$. An $R$-lattice of dimension $N$ in $\mathbb{C}^{n}$ is defined as the set of all $R$-linear combinations of $N$ linearly independent vectors, i.e.,

$$
\Lambda=\left\{\mathbf{r} \mathbf{G}_{\Lambda}: \mathbf{r} \in R^{N}\right\}
$$

where $\mathbf{G}_{\Lambda} \in \mathbb{C}^{\Lambda \times n}$ is called a generator matrix for $\Lambda$.
$R=\mathbb{Z}[\omega] \Rightarrow$ Eisenstein lattices; $R=\mathbb{Z}[i] \Rightarrow$ Gaussian lattices

| $\bullet$ | $\bullet$ | $\bullet$ |
| :---: | :---: | :---: |
| $\bullet$ | $\bullet$ | $\bullet$ |
| $\bullet$ | $\bullet$ | $\bullet$ |

$$
\begin{gathered}
\mathbb{Z}[\omega] \triangleq\left\{a+b \omega: a, b \in \mathbb{Z}, \omega=e^{i 2 \pi / 3}\right\} \\
\mathbb{Z}[i] \triangleq\{a+b i: a, b \in \mathbb{Z}\} \\
\Lambda=\mathbb{Z}[i] \\
\Lambda^{\prime}=3 \mathbb{Z}[i]
\end{gathered}
$$

## Algebraic Approach: Key Concepts

## Key Concepts

Message space $W$ (with $|W|=\left|\Lambda / \Lambda^{\prime}\right|$ )
Labeling $\varphi: \Lambda \rightarrow W$ (consistent with $\Lambda / \Lambda^{\prime}$ )
Embedding map $\bar{\varphi}: W \rightarrow \Lambda$ such that $\varphi(\bar{\varphi}(\mathbf{w}))=\mathbf{w}$


$$
W=\mathbb{Z}_{3} \times \mathbb{Z}_{3}
$$

$\varphi\left(\mathbf{w} \mathbf{G}_{\wedge}\right)=\mathbf{w} \bmod 3$

$$
\bar{\varphi}(\mathbf{w})=\mathbf{w} \mathbf{G}_{\wedge}
$$

## Algebraic Approach: Key Property

## Key Property

If the message space $W$ is chosen carefully, then the labelling $\varphi$ can be made linear.
In general, $W$ can be chosen as $R /\left(\pi_{1}\right) \times \cdots \times R /\left(\pi_{k}\right)$, where $\pi_{1}, \ldots, \pi_{k}$ are invariant factors of $\Lambda / \Lambda^{\prime}$.


$$
\begin{gathered}
\varphi\left(\mathbf{w} \mathbf{G}_{\wedge}\right)=\mathbf{w} \bmod 3 \\
\varphi(\bar{\varphi}(2,1)+\bar{\varphi}(1,0)) \\
=(0,1)
\end{gathered}
$$

## Algebraic Approach: Key Property

## Key Property

If the message space $W$ is chosen carefully, then the labelling $\varphi$ can be made linear. In general, $W$ can be chosen as $R /\left(\pi_{1}\right) \times \cdots \times R /\left(\pi_{k}\right)$, where $\pi_{1}, \ldots, \pi_{k}$ are invariant factors of $\Lambda / \Lambda^{\prime}$.

| $2+i$ | $\bullet$ <br> $\bullet$ | $1+i$ |
| :---: | :---: | :---: |
| $\bullet$ | $\bullet$ | $\bullet$ |
| $\bullet$ | $\bullet$ |  |
| $2+2 i$ | $2 i$ <br> $\bullet$ | $1+2 i$ <br> $\bullet$ |

## Algebraic Approach: Encoding and Decoding



## Encoding and Decoding

Transmitter $\ell$ sends $\mathbf{x}_{\ell}=\bar{\varphi}\left(\mathbf{w}_{\ell}\right)-Q_{\Lambda^{\prime}}\left(\bar{\varphi}\left(\mathbf{w}_{\ell}\right)\right)$
Receiver computes $\hat{\mathbf{u}}=\varphi\left(\mathcal{D}_{\Lambda}(\alpha \mathbf{y})\right)$
Remark: complexity here $\approx$ complexity for point-to-point channels

## Algebraic Approach: Error Probability



## Error Probability

$$
\operatorname{Pr}[\text { error }]=\operatorname{Pr}\left[\mathcal{D}_{\Lambda}\left(\mathbf{n}_{\text {eff }}\right) \notin \Lambda^{\prime}\right]
$$

where $\mathbf{n}_{\text {eff }} \triangleq \sum_{\ell}\left(\alpha h_{\ell}-a_{\ell}\right) \mathbf{x}_{\ell}+\alpha \mathbf{z}$ is the effective noise.

## Proof Sketch:

$$
\alpha \mathbf{y}=\sum_{\ell} a_{\ell} \mathbf{x}_{\ell}+\sum_{\ell}\left(\alpha h_{\ell}-a_{\ell}\right) \mathbf{x}_{\ell}+\alpha \mathbf{z}=\sum_{\ell} a_{\ell} \mathbf{x}_{\ell}+\mathbf{n}_{\text {eff }}
$$

Thus, $\mathcal{D}_{\Lambda}(\alpha \mathbf{y})=\sum_{\ell} a_{\ell} \mathbf{x}_{\ell}+\mathcal{D}_{\Lambda}\left(\mathbf{n}_{\text {eff }}\right)$
Therefore, $\hat{\mathbf{u}}=\varphi\left(\mathcal{D}_{\Lambda}(\alpha \mathbf{y})\right)=\mathbf{u}+\varphi\left(\mathcal{D}_{\wedge}\left(\mathbf{n}_{\text {eff }}\right)\right)$

Application: Practical Designs for Short Block Length

## Union Bound Estimator

## Union Bound Estimator (UBE) of the Error Probability

Recall that the effective noise $\mathbf{n}_{\text {eff }}=\sum_{\ell}\left(\alpha h_{\ell}-a_{\ell}\right) \mathbf{x}_{\ell}+\alpha \mathbf{z}$. If $\Lambda / \Lambda^{\prime}$ admits hypercube shaping, then the UBE is

$$
\operatorname{Pr}[\text { error }] \lesssim K\left(\Lambda / \Lambda^{\prime}\right) \exp \left(-\frac{d^{2}\left(\Lambda / \Lambda^{\prime}\right)}{4 N_{0}\left(|\alpha|^{2}+\operatorname{SNR}\|\alpha \mathbf{h}-\mathbf{a}\|^{2}\right)}\right)
$$

$K\left(\Lambda / \Lambda^{\prime}\right)$ : \# of the shortest vectors in the set difference $\Lambda-\Lambda^{\prime}$ $d\left(\Lambda / \Lambda^{\prime}\right)$ : length of the shortest vectors in the set difference $\Lambda-\Lambda^{\prime}$

## Implications

- minimize $K\left(\Lambda / \Lambda^{\prime}\right)$ and maximize $d\left(\Lambda / \Lambda^{\prime}\right)$
- minimize $Q(\alpha, \mathbf{a}) \triangleq|\alpha|^{2}+\operatorname{SNR}\|\alpha \mathbf{h}-\mathbf{a}\|^{2}$

Remark: $\mathbf{n}_{\text {eff }}$ has i.i.d. component with variance $N_{0} Q(\alpha, \mathbf{a}) \Rightarrow$ minimum variance criterion

## Figures of Merit

## Signal-to-Effective-Noise Ratio (SENR)

$$
\mathrm{SENR} \triangleq P / N_{0} Q(\alpha, \mathbf{a})
$$

## Nominal Coding Gain

$$
\operatorname{Pr}[\text { error }] \lesssim K\left(\Lambda / \Lambda^{\prime}\right) \exp \left(-\frac{3}{2} \frac{d^{2}\left(\Lambda / \Lambda^{\prime}\right)}{V(\Lambda)^{1 / n}} \frac{\text { SENR }}{2^{R_{\text {mes }}}}\right)
$$

Thus, $\gamma_{c}\left(\Lambda / \Lambda^{\prime}\right) \triangleq d^{2}\left(\Lambda / \Lambda^{\prime}\right) / V(\Lambda)^{1 / n}$ is nominal coding gain

## Effective Coding Gain

Rule of thumb: effective coding gain $=$ nominal coding gain

$$
-0.2 \mathrm{~dB} \times \log _{2}\left(K\left(\Lambda / \Lambda^{\prime}\right) / 4\right)
$$

## Practical Designs via Complex Construction A

## Setup

- 9-QAM + linear codes over $\mathbb{Z}[i] /(3)$
- Idea: terminated (feedforward) convolutional codes
- Why? better performance-complexity tradeoff
- Low complexity $\Rightarrow$ constraint length $\nu=1$ or 2
- Block length $=200$

| $\nu$ | $\mathbf{g}(D)$ | $\gamma_{c}\left(\Lambda / \Lambda^{\prime}\right)$ |
| :---: | :---: | :---: |
| 1 | $[1+(1+i) D,(1+i)+D]$ | $2(3 \mathrm{~dB})$ |
| 2 | $\left[1+D+(1+i) D^{2},(1+i)+(1-i) D+D^{2}\right]$ | $3(4.77 \mathrm{~dB})$ |

Remark: $\nu=1 \Rightarrow 9$ states; $\nu=2 \Rightarrow 81$ states

## Practical Designs via Complex Construction A



