

Course on (algebraic aspects of) Convolutional Codes

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CIMPA RESEARCH SCHOOL

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My most heartfelt thanks to the organizers

CIMPA RESEARCH SCHOOL
ALGEBRAIC METHODS IN CODING THEORY

- 1 Error-correcting codes: From block codes to convolutional codes
 - Basics: Polynomial encoders
- 2 Distance properties of convolutional codes
 - Maximum Distance Profile (MDP) and Maximum Distance Separable (MDS)
 - Construction of MDP and MDS: Superregular matrices
- 3 Decoding of Convolutional codes
 - Viterbi algorithm
 - Decoding of convolutional codes over the erasure channel
- 4 Network coding with convolutional codes
- 5 Avenues for further research
 - Motivated by applications: Video streaming and storage systems
 - More theoretical: Multidimensional convolutional codes and convolutional codes over \mathbb{Z}_p^r

Day 5: 2D Convolutional Codes

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- To encode data recorded geometrically in m dimensions (mD , with $m > 1$), e.g., pictures or videos.
- It is possible to work in a framework that takes advantage of the correlation of the data in several directions.
- Such framework would lead to m dimensional (mD) convolutional codes, generalizing the notion of 1D convolutional code.
- This generalization is nontrivial since 1D convolutional codes are represented over the polynomial ring in one variable whereas mD convolutional codes are represented over the polynomial ring in m independent variables.

Basics for 2D Convolutional Codes

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- The first attempts to develop the general theory and the basic algebraic properties of 2D/ m D convolutional codes were proposed in some papers. Questions about optimal distances and constructions of codes with large distance remained wide open for many years
- Next we introduce the basic notions of 2D convolutional codes that, despite its fundamental importance, have been very little investigated.

Definition

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Definition

A full row rank matrix $G(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times n}$ whose rows constitute a basis for \mathcal{C} is called a generator matrix or encoder of \mathcal{C} .

Definition

A matrix $U(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{n \times n}$ is *unimodular* if it has a polynomial inverse or equivalently if $\det U(z_1, z_2) \in \mathbb{F} \setminus \{0\}$.

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A matrix $G(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times n}$, with $n \geq k$, is called *left factor prime* (IFP) if for every factorization $G(z_1, z_2) = \bar{V}(z_1, z_2)G(z_1, z_2)$, with $V(z_1, z_2) \in \mathbb{F}^{k \times k}$, $V(z_1, z_2)$ is unimodular.

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And it is called *left zero prime* (IZP) if $G(z_1, z_2)$ admits a polynomial left inverse.

Theorem

Let $G(z_1, z_2) \in \mathbb{F}^{n \times k}$, with $n \geq k$. Then the following are equivalent:

- i) $G(z_1, z_2)$ is left factor prime;
- ii) there exists polynomial matrices $X_i(z_1, z_2)$ such that
 $G(z_1, z_2)X_i(z_1, z_2) = d_i(z_i)I_k$, where $d_i(z_i) \in \mathbb{F}[z_i] \setminus \{0\}$, for $i = 1, 2$;

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Theorem

- i) $G(z_1, z_2)$ is right zero prime;
- ii) $G(z_1, z_2)$ admits a polynomial left inverse;
- iii) there exists a polynomial matrix $V(z_1, z_2) \in \mathbb{F}^{n \times (n-k)}$ such that $[G(z_1, z_2) \ V(z_1, z_2)]$ is unimodular.

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It follows that zero primeness implies factor primeness, but the converse is not true.

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Lemma

Two generator matrices $G_1(z_1, z_2), G_2(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times n}$ define the same 2D convolutional code \mathcal{C} if and only if there exists $U(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times k}$, unimodular, such that

$$G_2(z_1, z_2) = U(z_1, z_2)G_1(z_1, z_2)$$

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Definition

The **degree** δ of a 2D convolutional code \mathcal{C} is the maximum of the degrees of the determinants of the $k \times k$ submatrices of any generator matrix of \mathcal{C} .

Minimal realization of 1D convolutional codes are very useful for many reasons: **less amount of memory**, help to construction of good codes, etc. These state-space representations are obtained via realizations of the encoders of the code. We know how to characterize them.

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Open problem: Minimal realizations

When considering the 2D case, there exist several state-space models (e.g. Marchesini-Fornasini or Roesser model). While in the 1D case there exists a characterization of minimality for realization via state-space models, the same does not happen in the 2D case.

Definition

The **weight** of $v(z_1, z_2) = \sum_{(i,j) \in \mathbb{N}^2} v_{(i,j)} z_1^i z_2^j \in \mathbb{F}[z_1, z_2]^n$ is defined as

$$\tilde{\mathbf{w}}v(z_1, z_2) = \sum_{(i,j) \in \mathbb{N}^2} \tilde{\mathbf{w}}v_{(i,j)}$$

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Definition

The **(free) distance** of a 2D convolutional code \mathcal{C} is defined as

$$d_{\text{free}}\mathcal{C} = \min\{\tilde{\mathbf{w}}v(z_1, z_2) \mid v(z_1, z_2) \in \mathcal{C} \setminus \{0\}\}$$

Theorem

If \mathcal{C} is a 2D convolutional code of rate k/n and degree δ , then

$$d_{\text{free}}\mathcal{C} \leq \frac{(\lfloor \delta/k \rfloor + 1)(\lfloor \delta/k \rfloor + 2)}{2}n - (k - \delta + k\lfloor \delta/k \rfloor) + 1$$

Definition

A 2D convolutional code of rate k/n and degree δ with distance achieving this bound is called **2D MDS convolutional code**.

A construction of a 2D convolutional code of rate $1/n$

Theorem

Let $n, \delta \in \mathbb{N}$. Assume that $n \geq \ell = \frac{(\delta+1)(\delta+2)}{2}$ and consider a superregular matrix

$$G = [G_0 \quad G_1 \quad \cdots \quad G_{\ell-1}] \in \mathbb{F}^{n \times \ell}.$$

Define

$$\begin{aligned} G(z_1, z_2) = & G_0 + G_1 z_1 + G_2 z_2 + G_3 z_1^2 + G_4 z_1 z_2 + G_5 z_2^2 + \cdots \\ & + G_{\frac{\delta(\delta+1)}{2}} z_1^\delta + G_{\frac{\delta(\delta+1)}{2}+1} z_1^{\delta-1} z_2 + \cdots + G_{\ell-1} z_2^\delta. \end{aligned}$$

Let $G(z_1, z_2)$ is the generator matrix of an 2D MDS convolutional code of rate $1/n$ and degree δ .

Example

In order to construct a 2D convolutional code of rate $1/12$ and $\delta = 3$ we build a superregular Cauchy matrix of size 12×10 . We need a field with at least 22 elements and then we consider the field $\mathbb{F} = GF(23)$. Take for instance,

$$\vec{x} = [0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11], \quad \vec{y} = [13, 14, 15, 16, 17, 18, 19, 20, 21, 22]$$

then we obtain the Cauchy matrix

$$A = \begin{bmatrix} 7 & 18 & 3 & 10 & 4 & 14 & 6 & 8 & 12 & 1 \\ 21 & 7 & 18 & 3 & 10 & 4 & 14 & 6 & 8 & 12 \\ 2 & 21 & 7 & 18 & 3 & 10 & 4 & 14 & 6 & 8 \\ 16 & 2 & 21 & 7 & 18 & 3 & 10 & 4 & 14 & 6 \\ 5 & 16 & 2 & 21 & 7 & 18 & 3 & 10 & 4 & 14 \\ 20 & 5 & 16 & 2 & 21 & 7 & 18 & 3 & 10 & 4 \\ 13 & 20 & 5 & 16 & 2 & 21 & 7 & 18 & 3 & 10 \\ 19 & 13 & 20 & 5 & 16 & 2 & 21 & 7 & 18 & 3 \\ 9 & 19 & 13 & 20 & 5 & 16 & 2 & 21 & 7 & 18 \\ 17 & 9 & 19 & 13 & 20 & 5 & 16 & 2 & 21 & 7 \\ 15 & 17 & 9 & 19 & 13 & 20 & 5 & 16 & 2 & 21 \\ 11 & 15 & 17 & 9 & 19 & 13 & 20 & 5 & 16 & 2 \end{bmatrix}.$$

Example

Now using the theorem we have the 2D CC of rate $1/12$ and $\delta = 3$ generated by the matrix

$$\begin{bmatrix} 7 + 18z_1 + 3z_2 + 10z_1^2 + 4z_1z_2 + 14z_2^2 + 6z_1^3 + 8z_1^2z_2 + 12z_1z_2^2 + z_2^3 \\ 21 + 7z_1 + 18z_2 + 3z_1^2 + 10z_1z_2 + 4z_2^2 + 14z_1^3 + 6z_1^2z_2 + 8z_1z_2^2 + 12z_2^3 \\ 2 + 21z_1 + 7z_2 + 18z_1^2 + 3z_1z_2 + 10z_2^2 + 4z_1^3 + 14z_1^2z_2 + 6z_1z_2^2 + 8z_2^3 \\ 16 + 2z_1 + 21z_2 + 7z_1^2 + 18z_1z_2 + 3z_2^2 + 10z_1^3 + 4z_1^2z_2 + 14z_1z_2^2 + 6z_2^3 \\ 5 + 16z_1 + 2z_2 + 21z_1^2 + 7z_1z_2 + 18z_2^2 + 3z_1^3 + 10z_1^2z_2 + 4z_1z_2^2 + 14z_2^3 \\ 20 + 5z_1 + 16z_2 + 2z_1^2 + 21z_1z_2 + 7z_2^2 + 18z_1^3 + 3z_1^2z_2 + 10z_1z_2^2 + 4z_2^3 \\ 13 + 20z_1 + 5z_2 + 16z_1^2 + 2z_1z_2 + 21z_2^2 + 7z_1^3 + 18z_1^2z_2 + 3z_1z_2^2 + 10z_2^3 \\ 19 + 13z_1 + 20z_2 + 5z_1^2 + 16z_1z_2 + 2z_2^2 + 21z_1^3 + 7z_1^2z_2 + 18z_1z_2^2 + 3z_2^3 \\ 9 + 19z_1 + 13z_2 + 20z_1^2 + 5z_1z_2 + 16z_2^2 + 2z_1^3 + 21z_1^2z_2 + 7z_1z_2^2 + 18z_2^3 \\ 17 + 9z_1 + 19z_2 + 13z_1^2 + 20z_1z_2 + 5z_2^2 + 16z_1^3 + 2z_1^2z_2 + 21z_1z_2^2 + 7z_2^3 \\ 15 + 17z_1 + 9z_2 + 19z_1^2 + 13z_1z_2 + 20z_2^2 + 5z_1^3 + 16z_1^2z_2 + 2z_1z_2^2 + 21z_2^3 \\ 11 + 15z_1 + 17z_2 + 9z_1^2 + 19z_1z_2 + 13z_2^2 + 20z_1^3 + 5z_1^2z_2 + 16z_1z_2^2 + 2z_2^3 \end{bmatrix}$$

is a 2D MDS convolutional code.

Convolutional Codes over \mathbb{Z}_p^r

- Motivation: Convolutional codes over the ring \mathbb{Z}_M are the most suitable for phase modulation signals [1].



J.L. Massey and T. Mittelholzer. (1990)

"Systematicity of convolutional codes over rings".



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MDS Convolutional Codes Over a Finite Ring
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C. Feng, R W. Nobrega, F R. Kschischang and D Silva (2014)

Communication over Finite-Chain-Ring Matrix Channels
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- Motivation: Convolutional codes over the ring \mathbb{Z}_M are the most suitable for phase modulation signals [1].
- We start with the ring \mathbb{Z}_{p^r} . By the Chinese Remainder Theorem, results on codes over \mathbb{Z}_{p^r} can be extended to codes over \mathbb{Z}_M .



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Now we have to take care of the Zero divisors

Example

Let $\mathcal{C} = \text{span}\{g_0, g_1\} \subset \mathbb{Z}_{33}^3[D]$ be a convolutional code, with $g_0 = [1 \quad 1+D \quad 0]$ and $g_1 = [3 \quad 0 \quad 3+3D]$.

- * **Encoder** $\longrightarrow \tilde{G}(D) = \begin{bmatrix} 1 & 1+D & 0 \\ 3 & 0 & 3+3D \end{bmatrix}$
- * g_0, g_1 are not linearly independent!

We only have a minimum number of generators but not necessarily linearly independent.

In order to solve this problem we will restrict to linear combinations with coefficients in $\mathcal{A}_p[D]$ where

$$\mathcal{A}_p = \{0, 1, 2, \dots, p-1\} \subset \mathbb{Z}_{p^r}.$$

↓

Obviously any element $a \in \mathbb{Z}_{p^r}$ can be written uniquely as (the p -adic expansion)

$$a = \alpha_0 + \alpha_1 p + \dots + \alpha_{r-1} p^{r-1}, \quad \alpha_i \in \mathcal{A}_p.$$

Example

Back to example
encoder

$$\tilde{G}(D) = \begin{bmatrix} 1 & 1+D & 0 \\ 3 & 0 & 3+3D \end{bmatrix}$$

new type of encoder

$$G(D) = \begin{bmatrix} g_0 \\ 3g_0 \\ 9g_0 \\ g_1 \\ 3g_1 \end{bmatrix} = \begin{bmatrix} 1 & 1+D & 0 \\ 3 & 3+D & 0 \\ 9 & 9+9D & 0 \\ 3 & 0 & 3+3D \\ 9 & 0 & 9+9D \end{bmatrix}$$

with $u(D) \in \mathcal{A}_p[D]^5$.

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with $u(D) \in \mathcal{A}_p[D]^5$. Only the message $u(D) = [1 \ 0 \ 0 \ 0 \ 0] \in \mathcal{A}_p[D]^5$ produces the codeword $[1 \ 1+D \ 0]$.

Let $\{v_1(D), \dots, v_k(D)\} \subset \mathbb{Z}_{p^r}^n[D]$.

$$\sum_{j=1}^k a_j(D)v_j(D), \quad a_j(D) \in \mathcal{A}_p[D],$$

is said to be a **p-linear combination** of $v_1(D), \dots, v_k(D)$.

The set of all p -linear combination of $v_1(D), \dots, v_k(D)$ is called the **p -span** of $\{v_1(D), \dots, v_k(D)\}$:

$$p\text{-span}(v_1(D), \dots, v_k(D)).$$

Obviously, $p\text{-span}(v_1(D), \dots, v_k(D))$ is not always a $\mathbb{Z}_{p^r}[D]$ -module!

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Obviously, $p\text{-span}(v_1(D), \dots, v_k(D))$ is not always a $\mathbb{Z}_{p^r}[D]$ -module!

Example: In $\mathbb{Z}_{33}^3[D]$

$$[3 \ 3 + 3D \ 0] \notin p\text{-span}([1 \ 1 + D \ 0])$$

Thus not a submodule of $\mathbb{Z}_{33}^3[D]$.

An ordered sequence of vectors $(v_1(D), \dots, v_k(D))$ in $\mathbb{Z}_p^n[D]$ is said to be a **p-generator sequence** if:

- 1 $p v_i(D)$ is a p -linear combination of $v_{i+1}(D), \dots, v_k(D)$,
 $i = 1, \dots, k - 1$;
- 2 $p v_k(D) = 0$.

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- 2 $p v_k(D) = 0$.

Example: in $\mathbb{Z}_{3^3}^3[D]$

$$([1 \ 1 + D \ 0], [3 \ 3 + 3D \ 0], [9 \ 9 + 9D \ 0])$$

is a p -generator sequence

If $V = (v_1(D), \dots, v_k(D))$ is a p -generator sequence then

$$p\text{-span } V = \text{span } V.$$

→ $p\text{-span } V$ is a submodule of $\mathbb{Z}_{p^r}^n[D]$, and we say that V is a p -generator sequence of $M = \text{span } V$.

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If $M = \text{span}(v_1(D), \dots, v_k(D))$ is a submodule of $\mathbb{Z}_{p^r}^n[D]$ then

$$(v_1(D), pv_1(D), \dots, p^{r-1}v_1(D), v_2(D), pv_2(D), \dots, \\ \dots, p^{r-1}v_2(D), \dots, v_l(D), pv_l(D), \dots, p^{r-1}v_l(D)).$$

is a p -generator sequence of M .

Example

$$M = \text{span}\{[1 + D \ 2 + 2D], [9 \ 0], [0 \ 9]\} \subset \mathbb{Z}_{33}^3[D]$$

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$$([1 + D \ 2 + 2D], [3 + 3D \ 6 + 6D], [9 \ 0], [0 \ 9])$$

is a p -generator sequence of M :

$$3[1 + D \ 2 + 2D] = [3 + 3D \ 6 + 6D]$$

$$3[3 + 3D \ 6 + 6D] = (1 + D)[9 \ 0] + (2 + 2D)[0 \ 9]$$

$$3[9 \ 0] = 3[0 \ 9] = [0 \ 0]$$

The vectors $v_1(D), \dots, v_k(D)$ are said to be **p-linearly independent** if the only p -linear combination of $v_1(D), \dots, v_k(D)$ that is equal to 0 is the trivial one.

The vectors $v_1(D), \dots, v_k(D)$ are said to be **p-linearly independent** if the only p -linear combination of $v_1(D), \dots, v_k(D)$ that is equal to 0 is the trivial one.

An ordered sequence of vectors $V = (v_1(D), \dots, v_k(D))$ which is a p -linearly independent p -generator sequence is said to be a **p-basis** and we say that V is a p -basis of $M = p\text{-span } V$.

Lemma

Two p -bases of a submodule of $\mathbb{Z}_{p^r}^n[D]$ have the same number of elements.

The number of elements of a p -basis of a submodule M of $\mathbb{Z}_{p^r}^n[D]$ is called **p -dimension** of M , denoted as $p\text{-dim}(M)$.

Example: $M = \text{span}([1 \ 1 + D \ 0], [3 \ 0 \ 3 + 3D]) \subset \mathbb{Z}_{3^3}^3[D]$

$([1 \ 1 + D \ 0], [3 \ 3 + 3D \ 0], [9 \ 9 + 9D \ 0], [3 \ 0 \ 3 + 3D], [9 \ 0 \ 9 + 9D])$

is a p -basis of M and consequently $p\text{-dim}(M) = 5$.

A particular p -basis

Let $v(D)$ be a nonzero vector in $\mathbb{Z}_{p^r}^n[D]$:

$$v(D) = v_0 + v_1 D + \cdots + v_\nu D^\nu,$$

with $v_i \in \mathbb{Z}_{p^r}^n$, $i = 0, \dots, \nu$, and $v_\nu \neq 0$.

- $v(D)$ has **degree** ν , $\deg v(D) = \nu$;
- v_ν is called the **leading coefficient vector** of $v(D)$, denoted by v^{lc} .

Let M be a submodule of $\mathbb{Z}_{p^r}^n[D]$ written as the p -span of a p -generator sequence $V = (v_1(D), \dots, v_k(D))$.

V is called a **reduced p -basis** for M if the leading coefficient vectors $v_1^{lc}, \dots, v_k^{lc}$ are p -linearly independent.

Example

$$M = \text{span}([1 \ 1 + D \ 0], [3 \ 0 \ 3 + 3D]) \subset \mathbb{Z}_{3^3}^3[D]$$

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$M = \text{span}([1 \ 1 + D \ 0], [3 \ 0 \ 3 + 3D]) \subset \mathbb{Z}_{3^3}[D]$

$([1 \ 1 + D \ 0], [3 \ 3 + 3D \ 0], [9 \ 9 + 9D \ 0], [3 \ 0 \ 3 + 3D], [9 \ 0 \ 9 + 9D])$

is a reduced p -basis of M ? Yes, since the leading coefficient vectors

$([0 \ 1 \ 0], [0 \ 3 \ 0], [0 \ 9 \ 0], [0 \ 0 \ 3], [0 \ 0 \ 9])$

are p -linearly independent.

Every submodule of $\mathbb{Z}_{p^r}^n[D]$ has a reduced p -basis.

A reduced p -basis for a submodule M of $\mathbb{Z}_{p^r}^n[D]$ gives rise to several invariants of M .

Let $V = (v_1(D), \dots, v_k(D))$ be a reduced p -basis of M .

- The degrees of $v_1(D), \dots, v_k(D)$ are called the **p -indices** of M ;
- The **p -degree** of M is defined as the sum of the p -indices of M .

 V.V. Vazirani, H. Saran and B.J. Rajan (1996)

An efficient algorithm for constructing minimal trellises for codes over finite abelian groups.

IEEE Trans. Information Theory, Vol. 42, pp. 1832-1854, 1996.

 M. Kuijper, R. Pinto and J.W.Polderman (2007)

The predictable degree property and row reducedness for systems over a finite ring

Linear Alg. Appl., Vol. 425, pp. 776-796, 2007.

A **convolutional code** \mathcal{C} of length n is a $\mathbb{Z}_{p^r}[D]$ -submodule of $\mathbb{Z}_{p^r}^n[D]$. If \mathcal{C} has p -dimension k and p -degree δ , we say that \mathcal{C} is an (n, k, δ) -convolutional code.

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A **p -encoder** $G(D) \in \mathbb{Z}_{p^r}[D]^{k \times n}$ of \mathcal{C} is a polynomial matrix whose rows are a p -basis of \mathcal{C} and therefore

$$\mathcal{C} = \text{Im}_{\mathcal{A}_p[D]} G(D) = \left\{ u(D)G(D) \in \mathbb{Z}_{p^r}^n[D] : u(D) \in \mathcal{A}_p[D]^k \right\}.$$

A **reduced p -encoder** is a polynomial matrix whose rows are a reduced p -basis of \mathcal{C} .

Note that all convolutional codes have a reduced p -encoder since every submodule of $\mathbb{Z}_{p^r}^n[D]$ has a reduced p -basis.



M. Kuijper, R. Pinto (2009)

On minimality of convolutional ring encoders

IEEE Trans. Information Theory, Vol. 55, No. 11, pp. 4890-4897, November 2009.

Block Codes

If a convolutional code admits a constant generator matrix, it is called a **block code**.

We introduce the notion of p-standard form from the definition of standard form.

Definition [G. H. Norton and A. Salagean, (2001)]

Let \mathcal{C} be a block code over $\mathbb{Z}_{p^r}^n$. A generator matrix \tilde{G} for \mathcal{C} is said to be in **standard form** if

$$\tilde{G} = \begin{bmatrix} I_{k_0} & A_{1,0}^0 & A_{2,0}^0 & A_{3,0}^0 & \cdots & A_{r-1,0}^0 & A_{r,0}^0 \\ 0 & pI_{k_1} & pA_{2,1}^1 & pA_{3,1}^1 & \cdots & pA_{r-1,1}^1 & pA_{r,1}^1 \\ 0 & 0 & p^2 I_{k_2} & p^2 A_{3,2}^2 & \cdots & p^2 A_{r-1,2}^2 & p^2 A_{r,2}^2 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & \cdots & p^{r-1} I_{k_{r-1}} & p^{r-1} A_{r,r-1}^{r-1} \end{bmatrix},$$

where the columns are grouped into blocks of sizes $k_0, \dots, k_{r-1}, n - \sum_{i=0}^{r-1} k_i$.

[6] Graham H. Norton and Ana Salagean (2001)
On the Hamming distance of linear codes over a finite chain ring
IEEE Trans. Information Theory, Vol. 46-3, pp. 1060-1067, 2001.

Block Codes

Definition

Let \mathcal{C} be a block code over $\mathbb{Z}_p^n[D]$. A p -encoder G of \mathcal{C} is said to be in **p -standard form** if

$$G = \begin{bmatrix} I_{k_0} & A_{1,0}^0 & A_{2,0}^0 & A_{3,0}^0 & \cdots & A_{r-1,0}^0 & A_{r,0}^0 \\ pI_{k_0} & 0 & pA_{2,1}^0 & pA_{3,1}^0 & \cdots & pA_{r-1,1}^0 & pA_{r,1}^0 \\ 0 & pI_{k_1} & pA_{2,1}^1 & pA_{3,1}^1 & \cdots & pA_{r-1,1}^1 & pA_{r,1}^1 \\ \vdots & \vdots & \vdots & \vdots & \cdots & \vdots & \vdots \\ p^2 I_{k_0} & 0 & 0 & p^2 A_{3,2}^0 & \cdots & p^2 A_{r-1,2}^0 & p^2 A_{r,2}^0 \\ 0 & p^2 I_{k_1} & 0 & p^2 A_{3,2}^1 & \cdots & p^2 A_{r-1,2}^1 & p^2 A_{r,2}^1 \\ 0 & 0 & p^2 I_{k_2} & p^2 A_{3,2}^2 & \cdots & p^2 A_{r-1,2}^2 & p^2 A_{r,2}^2 \\ \vdots & \vdots & \vdots & \vdots & \cdots & \vdots & \vdots \\ p^{r-1} I_{k_0} & 0 & 0 & 0 & \cdots & 0 & p^{r-1} A_{r,r-1}^0 \\ 0 & p^{r-1} I_{k_1} & 0 & 0 & \cdots & 0 & p^{r-1} A_{r,r-1}^1 \\ 0 & 0 & p^{r-1} I_{k_2} & 0 & \cdots & 0 & p^{r-1} A_{r,r-1}^2 \\ 0 & 0 & 0 & p^{r-1} I_{k_3} & \cdots & 0 & p^{r-1} A_{r,r-1}^3 \\ \vdots & \vdots & \vdots & \vdots & \cdots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & \cdots & p^{r-1} I_{k_{r-1}} & p^{r-1} A_{r,r-1}^{r-1} \end{bmatrix}$$

Definition

$G(D)$ is **noncatastrophic** if

$v(D) = u(D)G(D)$ with finite support $\rightarrow u(D)$ finite support

Definition

$G(D)$ is **noncatastrophic** if

$$v(D) = u(D)G(D) \text{ with finite support} \rightarrow u(D) \text{ finite support}$$

Open problem

any convolutional code over $\mathbb{Z}_{p^r}^n[D]$ admits a noncatastrophic p -encoder.

Conjecture

It was conjecture to be true.

The **free distance** of a convolutional code \mathcal{C} is defined as

$$d(\mathcal{C}) = \min\{wt(v(D)) : v(D) \in \mathcal{C}, v(D) \neq 0\},$$

where $wt(v(D))$ is the **weight** of a polynomial vector

$$v(D) = \sum_{i \geq 0} v_i D^i \in \mathbb{Z}_{p^r}^n[D]$$

given by

$$wt(v(D)) = \sum_{i \geq 0} wt(v_i),$$

with $wt(v_i)$ the number of non zero elements of v_i .

Main problem

How do we construct convolutional codes of a given length n , p -dimension k and p -degree δ with the largest possible distance?

Theorem

The free distance of an (n, k, δ) -convolutional code \mathcal{C} satisfies

$$d(\mathcal{C}) \leq n \left(\left\lfloor \frac{\delta}{k} \right\rfloor + 1 \right) - \left\lceil \frac{k}{r} \left(\left\lfloor \frac{\delta}{k} \right\rfloor + 1 \right) - \frac{\delta}{r} \right\rceil + 1.$$



M. El Oued and P. Solé (2013)

MDS Convolutional Codes Over a Finite Ring

IEEE trans. info. theory, Vol. 59, n. 11, november 2013.



D. Napp, R. Pinto and M. Toste

On MDS Convolutional Codes Over \mathbb{Z}_{p^r}

accepted in Designs, Codes and Cryptography.

An (n, k, δ) -convolutional code \mathcal{C} over \mathbb{Z}_{p^r} is said to be **Maximum Distance Separable (MDS)** if

$$d(\mathcal{C}) = n \left(\left\lfloor \frac{\delta}{k} \right\rfloor + 1 \right) - \left\lceil \frac{k}{r} \left(\left\lfloor \frac{\delta}{k} \right\rfloor + 1 \right) - \frac{\delta}{r} \right\rceil + 1.$$

Constructions of MDS convolutional codes

Given $n, k, \delta \in \mathbb{N}$, let us construct an MDS (n, k, δ) -convolutional code over \mathbb{Z}_{p^r} .

For simplicity, assume that $\mathbf{k} \mid \delta$.

Determine $(k_0, k_1, \dots, k_{r-1})$ such that

$$\begin{aligned} k_0 + k_1 + \dots + k_{r-1} &= \min_{k=rk'_0+(r-1)k'_1+\dots+k'_{r-1}} (k'_0 + k'_1 + \dots + k'_{r-1}) \\ &= \left\lceil \frac{k}{r} \right\rceil. \end{aligned}$$

Consider an MDS $(\tilde{n}, \tilde{k}, \tilde{\delta})$ -convolutional code $\tilde{\mathcal{C}}$ over the field \mathbb{Z}_p [1] with

$$\tilde{n} = n,$$

$$\tilde{k} = k_0 + k_1 + \cdots + k_{r-1},$$

$$\tilde{\delta} = \frac{\delta}{k} \tilde{k}.$$



[1] Smarandache, R. and Gluesing-Luerssen, H. and Rosenthal, J. (2001)
Constructions for MDS-Convolutional Codes
IEEE Trans. Automat. Control, vol. 47-5, pp.2045-2049, 2001.

Let

$$\tilde{G}(D) = \begin{bmatrix} \tilde{G}_{k_0}(D) \\ \text{---} \\ \tilde{G}_{k_1}(D) \\ \text{---} \\ \vdots \\ \text{---} \\ \tilde{G}_{k_{r-1}}(D) \end{bmatrix} \in \mathbb{Z}_p[D]^{\tilde{k} \times n}$$

be an encoder of \tilde{C} in reduced form, where $\tilde{G}_{k_i}(D)$ is a $k_i \times n$ matrix, $i = 0, 1, \dots, r-1, .$

The distance of $\tilde{\mathcal{C}}$ equals (from [2])

$$d(\tilde{\mathcal{C}}) = (n - \tilde{k}) \left(\left\lfloor \frac{\tilde{\delta}}{\tilde{k}} \right\rfloor + 1 \right) + \tilde{\delta} + 1.$$

From $\tilde{k} = \left\lceil \frac{k}{r} \right\rceil$ and $\tilde{\delta} = \frac{\delta}{k} \tilde{k}$ we get that

$$\begin{aligned} d(\tilde{\mathcal{C}}) &= n \left(\frac{\delta}{k} + 1 \right) - \left\lceil \frac{k}{r} \right\rceil + 1 \\ &= n \left(\frac{\delta}{k} + 1 \right) - \left\lceil \frac{k}{r} \left(\frac{\delta}{k} + 1 \right) - \frac{\delta}{r} \right\rceil + 1 \end{aligned}$$

We lift $\tilde{G}(D)$ to construct a $k \times n$ matrix $G(D)$:

$$G(D) = \begin{bmatrix} \tilde{G}_{k_0}(D) \\ p\tilde{G}_{k_0}(D) \\ \vdots \\ p^{r-1}\tilde{G}_{k_0}(D) \\ \hline p\tilde{G}_{k_1}(D) \\ p^2\tilde{G}_{k_1}(D) \\ \vdots \\ p^{r-1}\tilde{G}_{k_1}(D) \\ \hline \vdots \\ \hline p^{r-1}\tilde{G}_{k_{r-1}}(D) \end{bmatrix} .$$

Theorem

The matrix $G(D)$ defined above is a reduced p -encoder of an (n, k, δ) -convolutional code \mathcal{C} with $k \mid \delta$. Moreover, \mathcal{C} is MDS, i.e.,

$$d(\mathcal{C}) = n \left(\frac{\delta}{k} + 1 \right) - \left[\frac{k}{r} \left(\frac{\delta}{k} + 1 \right) - \frac{\delta}{r} \right] + 1$$

Remarks

- These constructions of MDS convolutional codes over \mathbb{Z}_{p^r} are “based” on MDS convolutional codes over \mathbb{Z}_p .
- Lifting techniques: Solé et. al used the Hensel lifting of a cyclic code. We used direct lifting.
- The known constructions of a (n, k, δ) - convolutional code require very large fields.

Open problems

- What happens if we consider other metrics? Homogeneous weights?
- More general class of finite rings?
- Characterization and existence of the dual codes of a convolutional code over \mathbb{Z}_p^r

Thank you for your attention!

Thanks to the organizers!