

# 2D convolutional codes

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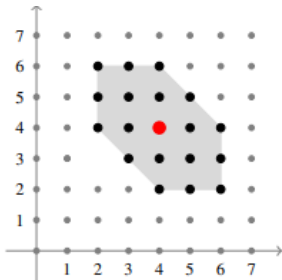
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# Overview

- 2D convolutional codes (Fornasini, Valcher, 94; Weiner, Rosenthal, 94)
- MDS 2D convolutional codes
- Decoding of 2D convolutional codes

## 2D (two-dimensional) data

- representation of pictures or videos;
- storage of digital data information.



$$\{u(4, 2), u(5, 2), u(6, 2), u(3, 3), u(4, 3), u(5, 3), u(6, 3), u(2, 4), \\ u(3, 4), u(4, 4), u(5, 4), u(6, 4), u(2, 5), u(3, 5), u(4, 5), u(5, 5), u(2, 6), \\ u(3, 6), u(4, 6)\}, \quad u(i, j) \in \mathbb{F}^k$$

$\{u(i, j) : u(i, j) \in \mathbb{F}^k, (i, j) \in \mathbb{N}^2\}$  with finite support

↓

$$u(z_1, z_2) = \sum_{(i, j) \in \mathbb{N}^2} u(i, j) z_1^i z_2^j \in \mathbb{F}^k[z_1, z_2] \simeq \mathbb{F}[z_1, z_2]^k$$

Former example:

$$u(z_1, z_2) = u(4, 2)z_1^4 z_2^2 + u(5, 2)z_1^5 z_2^2 + \cdots + u(4, 6)z_1^4 z_2^6$$

## 2D convolutional codes

### Definition

A two-dimensional (2D)  $(n, k)$  convolutional code  $\mathcal{C}$  is a free  $\mathbb{F}[z_1, z_2]$ -submodule of  $\mathbb{F}[z_1, z_2]^n$  of rank  $k$ .

A matrix  $G(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times n}$  whose rows form a basis for  $\mathcal{C}$  is called an encoder.

$$\begin{aligned}\mathcal{C} &= \text{Im}_{\mathbb{F}[z_1, z_2]} G(z_1, z_2) \\ &= \left\{ v(z_1, z_2) = u(z_1, z_2)G(z_1, z_2) : u(z_1, z_2) \in \mathbb{F}[z_1, z_2]^k \right\}\end{aligned}$$

## Definition

$U(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times k}$  is **unimodular** if it has a polynomial inverse.

## Lemma

$U(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times k}$  is unimodular iff  $\det U(z_1, z_2) \in \mathbb{F} \setminus \{0\}$ .

## Definition

$G(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times n}$ , with  $n \geq k$ , is,

1. **left factor prime ( $\ell$ FP)** if for every factorization

$$G(z_1, z_2) = T(z_1, z_2) \bar{G}(z_1, z_2),$$

with  $\bar{G}(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times n}$  and  $T(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times k}$ , then  $T(z_1, z_2)$  is unimodular.

2. **left zero prime ( $\ell$ ZP)** if the ideal generated by the  $k \times k$  minors of  $G(z_1, z_2)$  is  $\mathbb{F}[z_1, z_2]$ .

## Lemma

Let  $G(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times n}$  be an encoder of  $\mathcal{C}$ . Then  $\bar{G}(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times n}$  is an encoder of  $\mathcal{C}$  if and only if

$$\bar{G}(z_1, z_2) = U(z_1, z_2)G(z_1, z_2),$$

for some unimodular matrix  $U(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times k}$ .

Thus:

- if  $\mathcal{C}$  admits a left factor prime encoder, all its encoders are left factor prime encoders;
- if  $\mathcal{C}$  admits a left zero prime encoder, all its encoders are left zero prime encoders.

## Lemma

Let  $G(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times n}$ . Then

1.  $G(z_1, z_2)$  is  $\ell$ FP if and only if for all  $u(z_1, z_2) \in \mathbb{F}(z_1, z_2)^k$ ,  
 $u(z_1, z_2)G(z_1, z_2) \in \mathbb{F}[z_1, z_2]^n \Rightarrow u(z_1, z_2) \in \mathbb{F}[z_1, z_2]^k$ .
2.  $G(z_1, z_2)$  is  $\ell$ ZP if and only if  $G(z_1, z_2)$  admits a polynomial right inverse.

- A 2D convolutional code that admits a left factor prime encoder is said to be **noncatastrophic**.
- A 2D convolutional code that admits a left zero prime encoder is said to be **basic**.

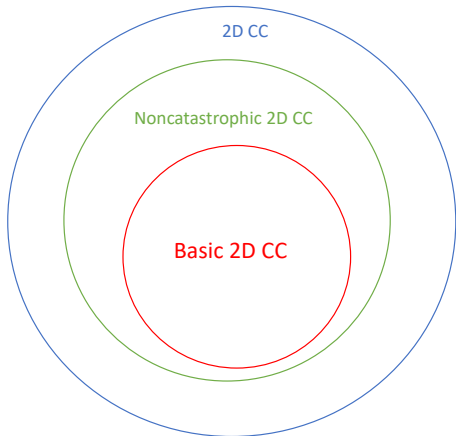


## Example

The 2D convolutional code with encoder

$$G(z_1, z_2) = \begin{bmatrix} z_1 & z_2 \end{bmatrix}$$

is noncatastrophic but it is not basic.



## Lemma

A 2D  $(n, k)$  convolutional code is noncatastrophic if and only if there exists a full row rank matrix  $H(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{(n-k) \times n}$  such that

$$\mathcal{C} = \{v(z_1, z_2) \in \mathbb{F}[z_1, z_2]^n : H(z_1, z_2)v(z_1, z_2)^T = 0\}.$$

## MDS 2D convolutional codes

The (total) degree of a polynomial  $p(z_1, z_2) = \sum_{(i,j) \in \mathbb{N}^2} p_{ij} z_1^i z_2^j$  in two indeterminates is defined as

$$\max\{i + j : p_{ij} \neq 0\}$$

The  $i$ -th row degree of a matrix  $G(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times n}$  is the maximum of the degrees of the polynomials in this row, and we call the **external degree** of  $G(z_1, z_2)$  as the sum of its row degrees.

The **internal degree** of a matrix  $G(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times n}$  is the maximum of the degrees of full size minors of  $G(z_1, z_2)$ .

## Definition

The **degree**,  $\delta$ , of  $\mathcal{C}$  is the minimum of the external degrees of the encoders of  $\mathcal{C}$  and  $\mathcal{C}$  is said to be an  $(n, k, \delta)$  convolutional code.

## Definition

The **complexity**,  $\delta_c$ , of  $\mathcal{C}$  is the internal degree of an encoder.

Obviously,  $\delta_c \leq \delta$ .

## Example

The 2D convolutional code with encoder

$$G(z_1, z_2) = \begin{bmatrix} 0 & z_2 & 1 \\ 1 & z_1 & 1 \end{bmatrix}$$

has complexity  $\delta_c = 1$  and degree  $\delta = 2$ .

## Distance

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

$$d(\mathcal{C}) = \{wt(v(z_1, z_2)) : v(z_1, z_2) \in \mathcal{C}, v(z_1, z_2) \neq 0\}$$

where  $wt(v(z_1, z_2))$  is the weight of a polynomial vector

$$v(z_1, z_2) = \sum_{(i,j) \in \mathbb{N}^2} v_{ij} z_1^i z_2^j \in \mathbb{F}^n[z_1, z_2]$$

given by

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

### Theorem (Climent, Napp, Perea, P., 2015)

Let  $\mathcal{C}$  be a 2D  $(n, k, \delta)$  convolutional code. Then

$$d(\mathcal{C}) \leq n \frac{(\lfloor \frac{\delta}{k} \rfloor + 1)(\lfloor \frac{\delta}{k} \rfloor + 2)}{2} - k(\lfloor \frac{\delta}{k} \rfloor + 1) + \delta + 1$$

*This upper bound is called the 2D generalized Singleton bound.*

A 2D  $(n, k, \delta)$  convolutional code is said to be a **Maximum Distance Separable (MDS)** 2D convolutional code if its distance equals the generalized Singleton bound.

## Constructions of 2D MDS convolutional codes

### Lemma

If  $\mathcal{C}$  is a MDS  $(n, k, \delta)$  2D convolutional code then  $\mathcal{C}$  has an encoder  $G(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times n}$  with  $t = k - \delta + k \lfloor \frac{\delta}{k} \rfloor$  rows of degree  $\lfloor \frac{\delta}{k} \rfloor$  and  $k - t$  rows of degree  $\lfloor \frac{\delta}{k} \rfloor + 1$ .

Note that a row with degree  $\lfloor \frac{\delta}{k} \rfloor$  has  $\ell_1 = \frac{(\lfloor \frac{\delta}{k} \rfloor + 1)(\lfloor \frac{\delta}{k} \rfloor + 2)}{2}$  coefficients in  $\mathbb{F}^n$  and a row with degree  $\lfloor \frac{\delta}{k} \rfloor + 1$  has  $\ell_2 = \frac{(\lfloor \frac{\delta}{k} \rfloor + 2)(\lfloor \frac{\delta}{k} \rfloor + 3)}{2}$  coefficients in  $\mathbb{F}^n$ .

Let us construct an encoder  $G(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times n}$  with the first  $k - t$  rows of degree  $\lfloor \frac{\delta}{k} \rfloor + 1$  and the last  $t$  rows of degree  $\lfloor \frac{\delta}{k} \rfloor$  (Climent, Napp, Perea, P., 2016).

Consider the matrices,

$$\mathcal{G}_r = \begin{cases} \begin{bmatrix} g_0^{(r)} \\ g_1^{(r)} \\ \vdots \\ g_{\ell_2-1}^{(r)} \end{bmatrix} \in \mathbb{F}^{\ell_2 \times n} \text{ for } r = 1, 2, \dots, k-t. \\ \begin{bmatrix} g_0^{(r)} \\ g_1^{(r)} \\ \vdots \\ g_{\ell_1-1}^{(r)} \end{bmatrix} \in \mathbb{F}^{\ell_1 \times n} \text{ for } r = k-t+1, k-t+2, \dots, k. \end{cases}$$



$$G(z_1, z_2) = \begin{bmatrix} G_1(z_1, z_2) \\ G_2(z_1, z_2) \\ \vdots \\ G_k(z_1, z_2) \end{bmatrix} \in \mathbb{F}[z_1, z_2]^{n \times k}$$

with

$$G_r(z_1, z_2) = \begin{cases} \sum_{0 \leq i+j \leq \lfloor \frac{\delta}{k} \rfloor + 1} g_{\mu(i,j)}^{(r)} z_1^i z_2^j & \text{for } r = 1, 2, \dots, k-t \\ \sum_{0 \leq i+j \leq \lfloor \frac{\delta}{k} \rfloor} g_{\mu(i,j)}^{(r)} z_1^i z_2^j & \text{for } r = k-t+1, \dots, k \end{cases}$$

where  $\mu(i, j) = j + \frac{(i+j)(i+j+1)}{2}$ .

Consider also

$$\bar{g}_r = \begin{cases} \left[ g_0^{(r)} & g_1^{(r)} & \cdots & g_{\ell_2-1}^{(r)} \right] & \text{for } r = 1, 2, \dots, k-t. \\ \left[ g_0^{(r)} & g_1^{(r)} & \cdots & g_{\ell_1-1}^{(r)} \right] & \text{for } r = k-t+1, k-t+2, \dots, k. \end{cases}$$

If

$$\mathcal{G} = \begin{bmatrix} \mathcal{G}_1 \\ \mathcal{G}_2 \\ \vdots \\ \mathcal{G}_k \end{bmatrix}$$

is full superregular and

$$\bar{\mathcal{G}} = \begin{bmatrix} \bar{\mathcal{G}}_1 \\ \vdots \\ \bar{\mathcal{G}}_{k-t} \\ \bar{\mathcal{G}}_{k-t+1} & 0 \\ \vdots \\ \bar{\mathcal{G}}_k & 0 \end{bmatrix}$$

is superregular, then  $G(z_1, z_2)$  is an encoder of an MDS  $(n, k, \delta)$  2D convolutional code if  $n \geq \ell_2 k$ .

- Constructions of MDS  $(n, k, \delta)$  2D convolutional codes if  $n > k + \delta$  (Almeida, Napp, P., 2018).

## Decoding of 2D convolutional codes

There is no decoding algorithm over the  $q$ -ary symmetric channel where errors occur.

Decoding algorithms over the erasure channel:

- using parity-check matrices (Climent, Napp, P., Simões, 2016).

## Decoding algorithm over the erasure channel

(Lieb, P., 2023)

Let  $\mathcal{C}$  be a 2D convolutional code and  $G(z_1, z_2) \in \mathbb{F}[z_1, z_2]^{k \times n}$  be an encoder of  $\mathcal{C}$ .

Write

$$G(z_1, z_2) = G_0(z_1) + G_1(z_1)z_2 + G_2(z_1)z_2^2 + \dots$$

and

$$u(z_1, z_2) = u_0(z_1) + u_1(z_1)z_2 + u_2(z_1)z_2^2 + \dots \in \mathbb{F}^k[z_1, z_2]$$

Then

$$\begin{aligned}v(z_1, z_2) &= u(z_1, z_2)G(z_1, z_2) \\&= u_0(z_1)G_0(z_1) + (u_1(z_1)G_0(z_1) + u_0(z_1)G_1(z_1))z_2 + \\&\quad + (u_2(z_1)G_0(z_1) + u_1(z_1)G_1(z_1) + u_0(z_1)G_2(z_1))z_2^2 + \dots \\&= u_0(z_1)G_0(z_1) + \begin{bmatrix} u_1(z_1) & u_0(z_1) \end{bmatrix} \begin{bmatrix} G_0(z_1) \\ G_1(z_1) \end{bmatrix} z_2 + \\&\quad + \begin{bmatrix} u_2(z_1) & u_1(z_1) & u_0(z_1) \end{bmatrix} \begin{bmatrix} G_0(z_1) \\ G_1(z_1) \\ G_2(z_1) \end{bmatrix} z_2^2 \\&\quad + \dots\end{aligned}$$

Let  $\ell$  be as large as possible such that

$$\mathcal{C}_0 = \text{Im}_{\mathbb{F}[z_1]} G_0(z_1), \quad \mathcal{C}_1 = \text{Im}_{\mathbb{F}[z_1]} \begin{bmatrix} G_0(z_1) \\ G_1(z_1) \end{bmatrix}, \dots$$

$$\mathcal{C}_\ell = \text{Im}_{\mathbb{F}[z_1]} \begin{bmatrix} G_0(z_1) \\ G_1(z_1) \\ \vdots \\ G_\ell(z) \end{bmatrix}$$

are full row rank (note that  $k(\ell + 1) < n$ ).



## Decoding (over the erasure channel)

Input data: received word

$$\tilde{v}(z_1, z_2) = \tilde{v}_0(z_1) + \tilde{v}_1(z_1)z_2 + \tilde{v}_2(z_1)z_2^2 + \dots$$

- 1) let  $i = \ell$ ;
- 2)  $\tilde{v}_i(z_1)$  is a codeword of  $\mathcal{C}_i$  with erasures;
- 3) recover the most number of erasures of  $\tilde{v}_i(z)$  as possible;

3.1) if the the erasures of  $\tilde{v}_i(z)$  are completely recovered and  $v(z_1)$  is the corrected codeword, then compute  $u_0(z_1), u_1(z_1), \dots, u_i(z_1)$  such that

$$v(z_1) = [ u_i(z_1) \quad u_{i-1}(z_1) \quad \dots \quad u_0(z_1) ] \begin{bmatrix} G_0(z_1) \\ G_1(z_1) \\ \vdots \\ G_i(z_1) \end{bmatrix};$$

for  $j \geq i$ , let  $\bar{v}_{i+j}(z_1) =$

$$\tilde{v}_{i+j}(z_1) - [ u_i(z_1) \quad u_{i-1}(z_1) \quad \dots \quad u_0(z_1) ] \begin{bmatrix} G_j(z_1) \\ \vdots \\ G_{i+j}(z_1) \end{bmatrix} \text{ and}$$

restart the decoding algorithm with input data

$$v(z_1, z_2) = \bar{v}_i(z_1) + \tilde{v}_{i+1}(z_1)z_2 + \tilde{v}_{i+2}(z_1)z_2^2 + \dots$$

3.2) if the erasures of  $\tilde{v}_i(z)$  are not completely recovered let  $i := i - 1$  and go to 2)

## Future work

- New constructions of MDS 2D convolutional codes.
- Decoding algorithms for error correction.
- Decoding algorithms for erasure correction - define an equivalent notion of column distances.

*Thank you for your attention!*

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