

# Column rank distances of rank metric convolutional codes

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# Motivation

- Multi-shot network coding  $\rightarrow$  rank metric convolutional codes
- Streaming applications  $\rightarrow$  column rank distance

# Overview

- Convolutional codes
- Rank metric convolutional codes
- Column rank distance
- Strongly-MRD rank metric convolutional codes

# Convolutional Codes

## Definition

A **convolutional code**  $\mathcal{C}$  of rate  $k/n$  is an  $\mathbb{F}_q[D]$ -submodule of  $\mathbb{F}_q^n[D]$  of rank  $k$ .

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

$$\mathcal{C} = \text{Im}_{\mathbb{F}_q[D]} G(D) = \left\{ u(D)G(D) : u(D) \in \mathbb{F}_q^k[D] \right\}$$

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Codewords are polynomial vectors

Any other encoder of  $\mathcal{C}$  can be obtained by  $\tilde{G}(D) = U(D)G(D)$ , where  $U(D) \in \mathbb{F}_q^{k \times k}[D]$  is **unimodular** (i.e., it admits an inverse in  $\mathbb{F}_q^{k \times k}[D]$ ).

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An encoder  $G(D) \in \mathbb{F}^{k \times n}[D]$  is **left prime** if

$$G(D) = L(D)\tilde{G}(D), \quad L(D) \in \mathbb{F}^{k \times k}[D], \quad \tilde{G}(D) \in \mathbb{F}^{k \times n}[D] \Rightarrow$$

$\Rightarrow L(D)$  unimodular.

If  $\mathcal{C}$  admits a **left prime** encoder then all its encoders are left prime  
 $\rightarrow \mathcal{C}$  is called **observable**.



Row reduced encoders are the ones that have minimal sum of the row degrees among all encoders of  $\mathcal{C} \rightarrow$  degree of  $\mathcal{C}$ .

If  $\mathcal{C}$  has rate  $k/n$  and degree  $\delta$  is said to be an  $(n, k, \delta)$ -convolutional code.

## Rank metric convolutional codes

A *rank metric convolutional code*  $\mathcal{C} \subset \mathbb{F}_q^{n \times m}[D]$  is the image of a monomorphism

$$\begin{aligned} \varphi : \mathbb{F}_q[D]^k &\xrightarrow{\gamma} \mathbb{F}_q[D]^{nm} \xrightarrow{\psi} \mathbb{F}_q[D]^{n \times m} \\ u(D) &\mapsto v(D) = u(D)G(D) \mapsto V(D) \end{aligned}$$

where  $G(D) \in \mathbb{F}_q^{k \times nm}[D]$  is full row rank and  $\psi$  is an isomorphism (e.g.:  $V(D) = \psi(v(D))$  s.t.  $V_{i,j}(D) = v_{mi+j}(D)$ ).

Codewords are polynomial matrices

- $G(D)$  is an **encoder** of  $\mathcal{C}$ ;
- $\mathcal{C}$  is called **observable** if all its encoders are left prime;
- the **degree**  $\delta$  of  $\mathcal{C}$  is the minimal sum of the row degrees of all encoders of  $\mathcal{C}$ .
- $\mathcal{C}$  is called an  $(n \times m, k, \delta)$ -rank metric convolutional code.

## Particular case ( $\delta = 0$ )

An  $(n \times m, k)$ -*linear rank metric code*  $\mathcal{C} \subset \mathbb{F}_q^{n \times m}$  is the image of a monomorphism

$$\begin{array}{ccccc} \varphi : \mathbb{F}_q^k & \xrightarrow{\gamma} & \mathbb{F}_q^{nm} & \xrightarrow{\psi} & \mathbb{F}_q^{n \times m} \\ u & \mapsto & v = uG & \mapsto & V \end{array}$$

where  $G \in \mathbb{F}_q^{k \times nm}$  is full row rank and  $\psi$  is an isomorphism (e.g.:  $V = \psi(v)$  s.t.  $V_{ij} = v_{mi+j}$ ).

Rank distance:  $d_R(A, B) = rk(A - B)$ ,  $A, B \in \mathbb{F}_q^{m \times n}$ .

Distance of  $\mathcal{C}$ :  $d_R(\mathcal{C}) = \min\{d_R(A, B) : A, B \in \mathcal{C}, A \neq B\}$ .

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If  $m \geq n$ :  $d_R(\mathcal{C}) \leq n - \lceil \frac{k}{m} \rceil + 1$  (Singleton bound).

If  $d_R(\mathcal{C}) = n - \lceil \frac{k}{m} \rceil + 1$ ,  $\mathcal{C}$  is MRD (Maximum Rank Distance).

## More known approach

A *linear rank metric code*  $\mathcal{C}$  is the image of a monomorphism

$$\begin{array}{ccccc} \varphi : \mathbb{F}_{q^m}^{\tilde{k}} & \xrightarrow{\gamma} & \mathbb{F}_{q^m}^n & \xrightarrow{\psi} & \mathbb{F}_q^{n \times m} \\ \tilde{u} & \mapsto & \tilde{v} = \tilde{u}\tilde{G} & \mapsto & V \end{array}$$

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where  $\tilde{G} \in \mathbb{F}_{q^m}^{\tilde{k} \times n}$  is full row rank and  $\psi$  is an isomorphism.

**Note:**  $\mathbb{F}_{q^m}^{\tilde{k}} \simeq \mathbb{F}_q^{\tilde{k}m}$ ;  $\mathbb{F}_{q^m}^n \simeq \mathbb{F}_q^{nm} \rightarrow k = m\tilde{k}$ , i.e.,  $m \mid k$



The previous scheme in coordinates, with  $G \in \mathbb{F}^{m\tilde{k} \times mn}$  is

$$\begin{array}{ccccc} \varphi : \mathbb{F}_q^{m\tilde{k}} & \xrightarrow{\gamma} & \mathbb{F}_q^{mn} & \xrightarrow{\psi} & \mathbb{F}_q^{n \times m} \\ u & \mapsto & v = uG & \mapsto & V \end{array}$$

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$$d_R(\mathcal{C}) \leq n - \lceil \frac{k}{m} \rceil + 1 = n - \tilde{k} + 1$$

## Gabidulin code (*Delsarte'73, Gabidulin'85*)

Let  $q = 2, m = 3, n = 2, \tilde{k} = 1$  and  $\alpha = 1 + \alpha^3$  generator of  $\mathbb{F}_{2^3}$  over  $\mathbb{F}_2$

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$\mathcal{C}$  is the image

$$\begin{aligned} \varphi : \mathbb{F}_{q^m}^{\tilde{k}} &\xrightarrow{\gamma} \mathbb{F}_{q^m}^n &\xrightarrow{\psi} \mathbb{F}_q^{n \times m} \\ \tilde{u} &\mapsto \tilde{v} = \tilde{u}\tilde{G} &\mapsto V \end{aligned}$$

where  $\tilde{G} = [\tilde{g}_1 \ \tilde{g}_2] \in \mathbb{F}_{2^3}^{1 \times 2}$  is s.t.  $\tilde{g}_1$  and  $\tilde{g}_2$  are LI over  $\mathbb{F}_2$ ; the rows of  $V$  are the coordinates of the columns of  $\tilde{v} = [\tilde{v}_1 \ \tilde{v}_2]$  over  $(1, \alpha, \alpha^2)$

$$\tilde{u} = u_0 + u_1\alpha + u_2\alpha^2 \in \mathbb{F}_2^3, \quad u_i \in \mathbb{F}_2,$$

$$\tilde{G} = [\tilde{g}_1 \ \tilde{g}_2] \text{ with}$$

$$\tilde{g}_1 = g_{01} + g_{11}\alpha + g_{21}\alpha^2, \quad g_{i1} \in \mathbb{F}_2 \text{ and } \tilde{g}_2 = g_{02} + g_{12}\alpha + g_{22}\alpha^2, \quad g_{i2} \in \mathbb{F}_2$$

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$$\tilde{u}\tilde{G} = [\tilde{u}\tilde{g}_1 \ \tilde{u}\tilde{g}_2]$$

$$\begin{aligned} \tilde{u}\tilde{g}_1 &= (u_0 + u_1\alpha + u_2\alpha^2)(g_{01} + g_{11}\alpha + g_{21}\alpha^2) \\ &= (u_0g_{01}) + (u_1g_{01} + u_0g_{11})\alpha + (u_2g_{01} + u_1g_{11} + u_0g_{21})\alpha^2 + \\ &\quad (u_2g_{11} + u_1g_{21})\alpha^3 + u_2g_{21}\alpha^4 \\ &= (u_0g_{01} + u_1g_{21}) + (u_0g_{11} + u_1(g_{01} + g_{21}) + u_2(g_{11} + g_{21}))\alpha + \\ &\quad (u_0g_{21} + u_1g_{101} + u_2(g_{01} + g_{21}))\alpha^2 \end{aligned}$$

$$\tilde{u}\tilde{G} = uG$$

$$= [u_0 \ u_1 \ u_2] \left[ \begin{array}{ccc|ccc} g_{01} & g_{11} & g_{21} & g_{02} & g_{12} & g_{22} \\ g_{21} & g_{01} + g_{21} & g_{11} & g_{22} & g_{02} + g_{22} & g_{12} \\ g_{11} & g_{11} + g_{21} & g_{01} + g_{21} & g_{12} & g_{12} + g_{22} & g_{02} + g_{22} \end{array} \right]$$



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$\mathcal{C}$  is the image

$$\begin{array}{ccccc} \varphi : \mathbb{F}_2^3 & \xrightarrow{\gamma} & \mathbb{F}_2^6 & \xrightarrow{\psi} & \mathbb{F}_q^{3 \times 2} \\ u & \mapsto & v = uG & \mapsto & V \end{array}$$

where  $v = [v_{01} \ v_{11} \ v_{21} \ | \ v_{02} \ v_{12} \ v_{22}]$  and  $V = \left[ \begin{array}{c|c} v_{01} & v_{02} \\ v_{11} & v_{12} \\ v_{21} & v_{22} \end{array} \right]^T$

## Back to rank metric convolutional codes

Distance:  $A(D) = \sum_{i \in \mathbb{N}} A_i D^i, B(D) = \sum_{i \in \mathbb{N}} B_i D^i \in \mathbb{F}^{n \times m}[D]$

Rank weight:  $rkwt(A(D)) = \sum_{i \in \mathbb{N}} rk(A_i)$

Sum rank distance:  $d_{SR}(A(D) - B(D)) = rkwt(A(D) - B(D))$

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Sum rank distance of a rank metric convolutional code

$$\begin{aligned} d_{SR}(C) &= \min_{V(D), U(D) \in C, V(D) \neq U(D)} d_{SR}(V(D), U(D)) \\ &= \min_{0 \neq V(D) \in C} rkwt(V(D)). \end{aligned}$$

## Singleton bound

Theorem (Napp,P.,Rosenthal,Vettori,2016)

Let  $\mathcal{C}$  be an  $(n, k, \delta)$ -rank metric convolutional code. Then the sum rank distance of  $\mathcal{C}$  is upper bounded by

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

An  $(n \times m, k, \delta)$ -rank metric convolutional code whose sum rank distance attains the Singleton bound is called **Maximum Rank Distance (MRD)**.

## $j$ -th column rank distance of $\mathcal{C}$

### Definition

Let  $\mathcal{C}$  be an observable  $(n \times m, k, \delta)$ -rank metric convolutional code.

For  $j \in \mathbb{N}$  the  $j$ -th column rank distance of  $\mathcal{C}$  is given by

$$d_j^{cr} = \min\{rkwt(V(D)_{|[0,j]}) : V(D) \in \mathcal{C} \text{ and } V_0 \neq 0\},$$

where for  $V(D) = \sum_{i \in \mathbb{N}} V_i D^i$  we define  $V(D)_{|[0,j]} = \sum_{i=0}^j V_i D^i$ .

## $j$ -th column rank distance of $\mathcal{C}$

### Theorem

Let  $\mathcal{C}$  be an observable  $(n \times m, k, \delta)$ -rank metric convolutional code. Then the  $j$ -th column rank distance of  $\mathcal{C}$  is upper bounded by

$$d_j^{cr} \leq j \left( n - \left\lfloor \frac{k}{m} \right\rfloor \right) + n - \left\lfloor \frac{k-1}{m} \right\rfloor$$

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$$d_0^{cr} \leq d_1^{cr} \leq d_2^{cr} \leq \dots \leq \lim_{j \rightarrow \infty} d_j^{cr} = d_{SR}$$

## Theorem

Let  $\mathcal{C}$  be an MRD observable  $(n \times m, k, \delta)$ -rank metric convolutional code with column rank distances  $d_j^{cr}, j \in \mathbb{N}_0$ , and sum rank distance  $d_{sumrank}$ .

Let  $M = \min\{j \in \mathbb{N}_0, d_j^{cr} = d_{sumrank}\}$ . Then,

$$M = \left\lceil \frac{n \lfloor \frac{\delta}{k} \rfloor + \left\lfloor \frac{\delta - k \lfloor \frac{\delta}{k} \rfloor}{m} \right\rfloor}{n - \lfloor \frac{k}{m} \rfloor} \right\rceil$$



## Definition

An observable  $(n \times m, k, \delta)$ -rank metric convolutional code with column rank distances  $d_j^{cr}, j \in \mathbb{N}_0$ , is called strongly-MRD if

$$d_M^{cr} = n \left( \left\lfloor \frac{\delta}{k} \right\rfloor + 1 \right) - \left\lfloor \frac{k \left( \left\lfloor \frac{\delta}{k} \right\rfloor + 1 \right) - \delta}{m} \right\rfloor + 1$$

for

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Remark: Strongly-MRD  $\Rightarrow$  MRD.

Open problem: Do they exist strongly - MRD rank metric convolutional codes for all parameters  $n, m, k, \delta$ ?

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Yes, if  $m \geq \delta + k$ . In this case:

$$M = \left\lfloor \frac{\delta}{k} \right\rfloor \text{ and } d_j^{cr} = (j+1)n, j = 0, \dots, \left\lfloor \frac{\delta}{k} \right\rfloor$$

$$\text{Singleton bound: } \left( \left\lfloor \frac{\delta}{k} \right\rfloor + 1 \right) n = d_M^{cr}$$

## Constructions ( $m \geq \delta + k$ )

$A \in \mathbb{F}_q^{m \times m}$  with irreducible characteristic polynomial  $\chi(\lambda)$ ,

$$\mathbb{F}_q[A] = \left\{ \sum_{i=0}^{m-1} \alpha_i A^i : \alpha_i \in \mathbb{F}_q, i = 0, \dots, m-1 \right\}$$

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is a field.

$X \in \mathbb{F}^{n \times m}$  full row rank

$$G_0 = \begin{bmatrix} \psi^{-1}(XI) \\ \psi^{-1}(XA) \\ \vdots \\ \psi^{-1}(XA^{k-1}) \end{bmatrix}, \quad G_1 = \begin{bmatrix} \psi^{-1}(XA^k) \\ \psi^{-1}(XA^{k+1}) \\ \vdots \\ \psi^{-1}(XA^{2k-1}) \end{bmatrix}, \dots,$$

$$G_{\lfloor \frac{\delta}{k} \rfloor} = \begin{bmatrix} \psi^{-1}(XA^{\lfloor \frac{\delta}{k} \rfloor k}) \\ \psi^{-1}(XA^{\lfloor \frac{\delta}{k} \rfloor k+1}) \\ \vdots \\ \psi^{-1}(XA^{(\lfloor \frac{\delta}{k} \rfloor + 1)k-1}) \end{bmatrix}, \quad G_{\lfloor \frac{\delta}{k} \rfloor + 1} = \begin{bmatrix} \psi^{-1}(XA^{\lfloor \frac{\delta}{k} \rfloor k}) \\ \vdots \\ \psi^{-1}(XA^{(\lfloor \frac{\delta}{k} \rfloor + 1)k-1}) \\ 0 \end{bmatrix}$$

## Theorem

The  $(n \times m, k, \delta)$ -rank metric convolutional code  $\mathcal{C}$  with encoder

$$G(D) = G_0 + G_1 D + \cdots + G_{\lfloor \frac{\delta}{k} \rfloor}, \text{ if } k|\delta, \text{ and}$$

$$G(D) = G_0 + G_1 D + \cdots + G_{\lfloor \frac{\delta}{k} \rfloor} + G_{\lfloor \frac{\delta}{k} \rfloor + 1}, \text{ otherwise}$$

is strongly-MRD when  $m \geq \delta$ .



## Example

Let  $n = 3$ ,  $m = 4$ ,  $k = 2$ ,  $\delta = 2$ , the companion matrix  $A$  of the irreducible polynomial  $\chi(\lambda) = \lambda^4 + \lambda + 1 \in \mathbb{F}_2[\lambda]$  and the full row rank matrix  $X = [I_3 \ 0_{3 \times 1}]$ , i.e.,

$$A = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 \end{bmatrix} \in \mathbb{F}_2^{4 \times 4}, \quad X = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \in \mathbb{F}_2^{3 \times 4}.$$

$\mathbb{F}_2[A] = \{u_0I + u_1A + u_2A^2 + u_3A^3 : u_i \in \mathbb{F}_2\}$  is a field

$$XA = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}, \quad XA^2 = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 \end{bmatrix}, \quad XA^3 = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \end{bmatrix},$$

$$G_0 = \begin{bmatrix} \psi^{-1}(XI) \\ \psi^{-1}(XA) \end{bmatrix} = \left[ \begin{array}{cccc|cccc} 1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \end{array} \middle| \begin{array}{ccc} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{array} \right]$$

$$G_1 = \begin{bmatrix} \psi^{-1}(XA^2) \\ \psi^{-1}(XA^3) \end{bmatrix} = \left[ \begin{array}{cccc|cccc} 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 0 & 0 \end{array} \middle| \begin{array}{ccc} 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \end{array} \right].$$

The convolutional code with encoder  $G(D) = G_0 + G_1D$  has degree 2.

Consider  $u(D) = [u_0 \ u_1] + [u_2 \ u_3]D + \cdots \in \mathbb{F}_2[D]^2$ , with  $[u_0 \ u_1] \neq 0$ .

The first two coefficients of the codeword  $V(D) = \varphi(u(D))$  are

$$V_0 = X(u_0I + u_1A) \text{ and } V_1 = X(u_2I + u_3A + u_0A^2 + u_1A^3).$$

$rk(V_0) = rk(V_1) = 3 \Rightarrow d_0^{cr} = 3$  and  $d_1^{cr} = 6$ , which means that  $\mathcal{C}$  is strongly-MRD.

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*Thank you for your attention!*