

LRPC codes with multiple syndromes: near ideal-size KEMs without ideals

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Monday, April 4, 2022

Plan

- 1 Background on rank metric and LRPC codes
- 2 Presentation of LRPC-MS
- 3 Analysis of the decoding failure rate
- 4 Bonus : advances in LRPC implementations
- 5 Conclusion and perspectives

Summary

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Rank metric codes

In rank metric, we consider \mathbb{F}_{q^m} -linear codes (\mathbb{F}_{q^m} is a field extension of \mathbb{F}_q of degree m).

Definition (Rank weight)

An element $\mathbf{x} = (x_1, \dots, x_n) \in (\mathbb{F}_{q^m})^n$ can be unfolded against an \mathbb{F}_q -basis of \mathbb{F}_{q^m} in a matrix

$$\mathcal{M}(\mathbf{x}) = \begin{pmatrix} x_{1,1} & \dots & x_{n,1} \\ \vdots & & \vdots \\ x_{1,m} & \dots & x_{n,m} \end{pmatrix} \in \mathcal{M}_{m,n}(\mathbb{F}_p)$$

The rank weight of \mathbf{x} is defined as the rank of this matrix (which does not depend on the choice of the basis).

$$w_r(\mathbf{x}) = \text{Rank } \mathcal{M}(\mathbf{x}) \in [0, \min(m, n)]$$

Example

Let $\mathbb{F}_8 = \mathbb{F}_{2^3}$ and let α such that $\mathbb{F}_8 \simeq \mathbb{F}_2[\alpha] = \text{Vect}(1, \alpha, \alpha^2)$.

Example

$$\mathbf{x} = (1, \alpha, \alpha^2 + 1, \alpha + 1) \in \mathbb{F}_8^4$$

$$\mathcal{M}(\mathbf{x}) = \begin{pmatrix} 1 & 0 & 1 & 1 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

$$w_r(\mathbf{x}) = 3$$

Support in rank metric

Definition (Rank support)

The support of a word $\mathbf{x} = (x_1, \dots, x_n) \in (\mathbb{F}_{q^m})^n$ is the subspace of \mathbb{F}_{q^m} generated by its coordinates :

$$\text{Supp}(\mathbf{x}) = \langle x_1, \dots, x_n \rangle_{\mathbb{F}_q} \subset \mathbb{F}_{q^m}$$

Hamming metric : $w_h(\mathbf{x}) = |\text{Supp}(\mathbf{x})|$

Rank metric : $w_r(\mathbf{x}) = \dim(\text{Supp}(\mathbf{x}))$

Ideal structure

To reduce the memory footprint of a generator matrix, we define ideal codes.

Definition (Double circulant code)

A double circulant code is a code $\mathcal{C}[2n, n]$ which admits a double circulating matrix as a generating matrix :

$$\mathbf{G} = \left(\begin{array}{cccc|cccc} a_0 & a_1 & \dots & a_{n-1} & b_0 & b_1 & \dots & b_{n-1} \\ a_{n-1} & a_0 & \ddots & a_{n-2} & b_{n-1} & b_0 & \ddots & b_{n-2} \\ \vdots & \ddots & \ddots & \vdots & \vdots & \ddots & \ddots & \vdots \\ a_1 & a_2 & \dots & a_0 & b_1 & b_2 & \dots & b_0 \end{array} \right)$$

Ideal structure

Definition (Ideal matrix)

Let $P(X)$ a polynomial in $\mathbb{F}_q[X]$ of degree n . A square matrix M of size $n \times n$ is ideal modulo P generated by $f(X)$ when it is of the form :

$$M = \begin{pmatrix} f(X) \bmod P \\ Xf(X) \bmod P \\ \vdots \\ X^{n-1}f(X) \bmod P \end{pmatrix}.$$

Definition (Ideal code)

An ideal code is a code $\mathcal{C}[2n, n]$ having $\mathbf{G} = (G_1|G_2)$ as a generator matrix where G_1 and G_2 are two ideal matrices.

Difficult problems in rank metric

Definition (Rank Syndrome Decoding $\text{RSD}(n, k, w)$)

Given a random parity check matrix $\mathbf{H} \in \mathcal{M}_{n-k, n}(\mathbb{F}_q)$ and a syndrome $\mathbf{s} = \mathbf{H}\mathbf{e}$ for \mathbf{e} an error of rank weight $w(\mathbf{e}) = w$, find \mathbf{e} .

Difficult problems in rank metric

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Definition (Rank Support Learning $\text{RSL}(n, k, w, \ell)$)

Given a random parity check matrix $\mathbf{H} \in \mathcal{M}_{n-k, n}(\mathbb{F}_q)$ and ℓ syndromes $\mathbf{s}_i = \mathbf{H}\mathbf{e}_i$ for \mathbf{e}_i errors of same support E a subspace of dimension w , find E .

Difficult problems in rank metric

Definition (Ideal Rank Syndrome Decoding IRSD(n, k, w))

Given an ideal random parity check matrix $\mathbf{H} \in \mathcal{M}_{n-k, n}(\mathbb{F}_q)$ and a syndrome $\mathbf{s} = \mathbf{H}\mathbf{e}$ for \mathbf{e} an error of rank weight $w(\mathbf{e}) = w$, find \mathbf{e} .

Problematic with the structure :

- Quantum attacks¹
- Potential weaknesses

1. Ronald CRAMER, Léo DUCAS et Benjamin WESOŁOWSKI. “Mildly short vectors in cyclotomic ideal lattices in quantum polynomial time”. In : *Journal of the ACM (JACM)* 68.2 (2021), p. 1–26.

Low Rank Parity Check Codes

An LRPC code is a code which admits a parity check matrix whose coordinates belong to a subspace of \mathbb{F}_{q^m} of small dimension.

Definition (LRPC codes)

Let $\mathbf{H} = (h_{ij})_{\substack{1 \leq i \leq n-k \\ 1 \leq j \leq n}} \in \mathbb{F}_{q^m}^{(n-k) \times n}$ be a full-rank matrix such that its coordinates generate an \mathbb{F}_q -subspace F of small dimension d :

$$F = \langle h_{ij} \rangle_{\mathbb{F}_q}.$$

Let \mathcal{C} be the code with parity-check matrix \mathbf{H} . By definition, \mathcal{C} is an $[n, k]$ LRPC code of dual weight d . Such a matrix \mathbf{H} is called a homogeneous matrix of weight d and support F .

Example

Let us consider again the field $\mathbb{F}_8 = \text{Vect}(1, \alpha, \alpha^2)$

Example

$$H = \begin{pmatrix} 1 & \alpha & \alpha \\ \alpha & 0 & \alpha + 1 \\ \alpha & \alpha & \alpha \end{pmatrix}$$

is of rank 3 as an \mathbb{F}_{q^m} -matrix but the \mathbb{F}_q -subspace generated by its coordinates is of dimension 2.

$$(1, \alpha, \alpha, \alpha, 0, \alpha + 1, \alpha, \alpha, \alpha) \rightarrow \begin{pmatrix} 1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

LRPC decoding

Problem

Let $E = \langle e_1, \dots, e_r \rangle$ an (unknown) subspace of \mathbb{F}_{q^m} of dimension r and $F = \langle f_1, \dots, f_d \rangle$ a (given) subspace of \mathbb{F}_{q^m} of dimension d . Given an LRPC matrix $\mathbf{H} \in F^{n-k \times n}$ and $\mathbf{s} = \mathbf{H}\mathbf{e}$ where $\mathbf{e} \in E^n$, find E .

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The coordinates of \mathbf{s} belong to the product space

$$EF = \text{Vect}\{ef \mid e \in E, f \in F\} = \langle e_1 f_1, \dots, e_r f_1, \dots, e_1 f_d, \dots, e_r f_d \rangle.$$

LRPC decoding

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The coordinates of \mathbf{s} belong to the product space
 $EF = \text{Vect}\{ef \mid e \in E, f \in F\} = \langle e_1 f_1, \dots, e_r f_1, \dots, e_1 f_d, \dots, e_r f_d \rangle$.

The subspaces $f_i^{-1}EF$ all contain E since for example
 $f_1^{-1}EF = \langle e_1, \dots, e_r, \dots, f_1^{-1}e_1 f_d, \dots, f_1^{-1}e_r f_d \rangle$.

So one can hope :

$$\bigcap_{i=1}^d f_i^{-1}EF = E$$

LRPC decoding

Algorithm 1: Rank Support Recovery (RSR) algorithm

Data: $F = \langle f_1, \dots, f_d \rangle$ an \mathbb{F}_q -subspace of \mathbb{F}_{q^m} ,
 $s = (s_1, \dots, s_{n-k}) \in \mathbb{F}_{q^m}^{(n-k)}$ a syndrome of an error e of weight r and of support E

Result: A candidate for the vector space E

//Part 1: Compute the vector space EF

1 Compute $S = \langle s_1, \dots, s_{n-k} \rangle$

//Part 2: Recover the vector space E

2 $E \leftarrow \bigcap_{i=1}^d f_i^{-1} S$ **return** E

Failure probability

Two possible cases of failure :

- $S \subsetneq EF$, the coordinates of the syndrome do not generate the entire space EF , or

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- $E \subsetneq S_1 \cap \dots \cap S_d$, the chain of intersection generates a subspace strictly bigger than E .

Failure probability

Two possible cases of failure :

- $S \subsetneq EF$, the coordinates of the syndrome do not generate the entire space EF , or
- $E \subsetneq S_1 \cap \dots \cap S_d$, the chain of intersection generates a subspace strictly bigger than E .

Proposition

The Decoding Failure Rate of algorithm RSR is bounded from above by :

$$q^{rd-(n-k)-1} + q^{-(d-1)(m-rd-r)}$$

Application of LRPC to cryptography

Definition (Key generation)

Let $\mathbf{U} = (\mathbf{A}|\mathbf{B})$ an LRPC matrix of weight d .

$$\begin{cases} pk &= \mathbf{H} = (\mathbf{I}|\mathbf{A}^{-1}\mathbf{B}) \\ sk &= \mathbf{U} \end{cases}$$

Application of LRPC to cryptography

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Definition (Encaps)

Choose an error support E of dimension r . Pick a random error \mathbf{e} in E^n and send ciphertext $\mathbf{c} = \mathbf{H}\mathbf{e}$. The shared secret is $\text{Hash}(E)$.

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Definition (Decaps)

Compute $\mathbf{s} = \mathbf{A}\mathbf{c} = \mathbf{U}\mathbf{e}$ and use LRPC decoding to find E .

ROLLO-II parameters

Instance	q	n	m	r	d	Security	DFR
ROLLO-II-128	2	189	83	7	8	128	2^{-134}
ROLLO-II-192	2	193	97	8	8	192	2^{-130}
ROLLO-II-256	2	211	97	8	9	256	2^{-136}

FIGURE: Parameters for ROLLO-II.

Instance	pk size	sk size	ct size	Security
ROLLO-II-128	1941	40	2089	128
ROLLO-II-192	2341	40	2469	192
ROLLO-II-256	2559	40	2687	256

FIGURE: Resulting sizes in bytes for ROLLO-II using NIST seed expander initialized with 40 bytes long seeds. The security is expressed in bits.

ROLLO-I parameters

Instance	q	n	m	r	d	Security	DFR
ROLLO-I-128	2	83	67	7	8	128	2^{-28}
ROLLO-I-192	2	97	79	8	8	192	2^{-34}
ROLLO-I-256	2	113	97	9	9	256	2^{-33}

FIGURE: Parameters for ROLLO-I.

Instance	pk size	sk size	ct size	Security
ROLLO-I-128	696	40	696	128
ROLLO-I-192	958	40	958	192
ROLLO-I-256	1371	40	1371	256

FIGURE: Resulting sizes in bytes for ROLLO-I using NIST seed expander initialized with 40 bytes long seeds. The security is expressed in bits.

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Idea

Definition (Key generation)

Let $\mathbf{U} = (\mathbf{A}|\mathbf{B})$ an LRPC matrix of weight d .

$$\begin{cases} pk &= \mathbf{H} = (\mathbf{I}|\mathbf{A}^{-1}\mathbf{B}) \\ sk &= \mathbf{U} \end{cases}$$

Definition (Encaps)

Choose an error support E of dimension r . Pick ℓ random errors \mathbf{e}_i in E^n for $1 \leq i \leq \ell$ and send ciphertexts $\mathbf{c}_i = \mathbf{H}\mathbf{e}_i$. The shared secret is $\text{Hash}(E)$.

Definition (Decaps)

Compute $\mathbf{s}_i = \mathbf{A}\mathbf{c}_i = \mathbf{U}\mathbf{e}_i$ and use LRPC decoding with multiple syndromes to find E .

LRPC decoding with multiple syndromes

The LRPC decoding algorithm has several syndromes as inputs

$$s_j = \mathbf{U}e_j.$$

$$\mathbf{S} = \mathbf{UV}$$

LRPC decoding with multiple syndromes

Algorithm 2: Rank Support Recovery (RSR) algorithm with multiple syndromes

Data: $F = \langle f_1, \dots, f_d \rangle$ an \mathbb{F}_q -subspace of \mathbb{F}_{q^m} , $\mathbf{S} = (s_{ij}) \in \mathbb{F}_{q^m}^{(n-k) \times \ell}$
 the ℓ syndromes of error vectors of weight r and support E

Result: A candidate for the vector space E

//Part 1: Compute the vector space EF

1 Compute $S = \langle s_{11}, \dots, s_{(n-k)\ell} \rangle$

//Part 2: Recover the vector space E

2 $E \leftarrow \bigcap_{i=1}^d f_i^{-1} S$

3 **return** E

New failure probability

Proposition

For $k \geq \ell$ and for \mathbf{U} and \mathbf{V} random variables chosen uniformly in $F^{(n-k) \times n}$ and $E^{n \times \ell}$ respectively, the Decoding Failure Rate of algorithm $RSR(F, \mathbf{UV}, r)$ is bounded from above by :

$$(n - k)q^{rd - (n-k)\ell} + q^{-(d-1)(m-rd-r)}$$

Parameters with an ideal structure

Instance	q	n	k	m	r	d	ℓ	Security	DFR
ILRPC-MS-128	2	94	47	83	7	8	4	128	2^{-126}
ILRPC-MS-192	2	134	67	101	8	8	4	192	2^{-198}

FIGURE: Parameters for ILRPC-MS

Instance	pk size	sk size	ct size	Security
ILRPC-MS-128	488	40	1,951	128
ILRPC-MS-192	846	40	3,384	192

FIGURE: Resulting sizes in bytes for ILRPC-MS using NIST seed expander initialized with 40 bytes long seeds. The security is expressed in bits.

Parameters without an ideal structure

Instance	q	n	k	m	r	d	ℓ	Security	DFR
LRPC-MS-128	2	34	17	113	9	10	13	128	2^{-126}
LRPC-MS-192	2	42	21	139	10	11	15	192	2^{-190}

FIGURE: Parameters for LRPC-MS

Instance	pk size	sk size	ct size	Security
LRPC-MS-128	4,083	40	3,122	128
LRPC-MS-192	7,663	40	5,474	192

FIGURE: Resulting sizes in bytes for LRPC-MS using NIST seed expander initialized with 40 bytes long seeds. The security is expressed in bits.

Comparison to other KEMs

Instance	128 bits	192 bits
LRPC-MS	7,205	12,445
Loong.CCAKEM-III	18,522	N/A
FrodoKEM	19,336	31,376
Loidreau cryptosystem	36,300	N/A
Classic McEliece	261,248	524,348

FIGURE: Comparison of sizes of unstructured post-quantum KEMs. The sizes represent the sum of public key and ciphertext expressed in bytes.

Instance	128 bits	192 bits
ILRPC-MS	2,439	4,230
BIKE	3,113	6,197
ROLLO-II	4,030	4,810
HQC	6,730	13,548

FIGURE: Comparison of sizes of structured code-based KEMs. The sizes represent the sum of public key and ciphertext expressed in bytes.

Specificity to rank metric

- Sending errors with the same support does not make sense in Hamming metric
- Additional information given by multiple syndromes can be specifically leveraged by LRPC decoding algorithm

IND-CPA proof

Definition (LRPC indistinguishability)

Given a matrix $\mathbf{H} \in \mathbb{F}_{q^m}^{(n-k) \times k}$, distinguish whether the code \mathcal{C} with the parity-check matrix $(\mathbf{I}_{n-k} | \mathbf{H})$ is a random code or an LRPC code of weight d .

Definition (Rank Support Learning RSL(n, k, w, ℓ))

Given a random parity check matrix $\mathbf{H} \in \mathcal{M}_{n-k,n}(\mathbb{F}_q)$ and ℓ syndromes $\mathbf{s}_i = \mathbf{H}\mathbf{e}_i$ for \mathbf{e}_i errors of same support E a subspace of dimension w , find E .

\Rightarrow considered difficult as long as $\ell \leq k(r-3)$ (without ideal structure) or $\ell \leq r-3$ (with ideal structure).

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Objective

We fix E and F subspaces of \mathbb{F}_{q^m} of dimension r and d respectively such that EF is of dimension rd . We also impose $q = 2$.

Theorem

For $n_1 + n_2 \leq n$ and for \mathbf{U} and \mathbf{V} random variables chosen uniformly in $F^{n_1 \times n}$ and $E^{n \times n_2}$ (respectively),

$$\mathbb{P}(\text{Supp}(\mathbf{UV}) \neq EF) \leq n_1 q^{rd - n_1 n_2}$$

Product of matrices

$$\begin{pmatrix} * & \dots & \dots & \dots & * \\ \vdots & & \mathbf{U} & & \vdots \\ * & \dots & \dots & \dots & * \end{pmatrix}
 \begin{pmatrix} * & \dots & * \\ \vdots & & \vdots \\ \vdots & \mathbf{V} & \vdots \\ \vdots & & \vdots \\ \vdots & & \vdots \\ * & \dots & * \end{pmatrix}
 \begin{pmatrix} * & \dots & * \\ \vdots & & \vdots \\ \vdots & \mathbf{UV} & \vdots \\ \vdots & & \vdots \\ * & \dots & * \end{pmatrix}$$

Impossible to use Leftover Hash Lemma

Lemma (Leftover Hash Lemma)

Let $\{\Phi_r\}_{r \in R}$ be a $(1 + \alpha)/m$ -almost universal family of hash functions from S to T , where $m := |T|$. Let H and X be independent random variables, where H is uniformly distributed over R , and X takes values in S . If β is the collision probability of X , and δ is the distance of $(H, \Phi_H(X))$ from uniform on $R \times T$, then $\delta \leq 1/2\sqrt{m\beta + \alpha}$.

$$\begin{array}{ccc}
 \mathbf{U} \times & & \mathbf{V} \rightarrow & & \mathbf{UV} \\
 d^{n_1 n} \text{ possibilities} & & r^{n_2 n} \text{ possibilities} \ll & & (rd)^{n_1 n_2} \text{ possibilities}
 \end{array}$$

Main proof idea

We fix ϕ a linear form on EF and we study the probability to have $\phi(\mathbf{U}\mathbf{v}) = \mathbf{0}$.

Lemma

We denote \mathbf{v} a column vector in E^n .

For a fixed \mathbf{U} , $\varphi_{\mathbf{U}} : \mathbf{v} \mapsto \phi(\mathbf{U}\mathbf{v})$ is a linear map from E^n to $\mathbb{F}_q^{n_1}$ so the distribution of $\phi(\mathbf{U}\mathbf{v})$ is uniform in $\text{Im}(\varphi_{\mathbf{U}}) \subset \mathbb{F}_q^{n_1}$.

We shall study the rank of $\varphi_{\mathbf{U}}$.

Indeed, when $\text{Rank}(\varphi_{\mathbf{U}}) = i$,

$$\mathbb{P}(\phi(\mathbf{U}\mathbf{v}) = \mathbf{0}) = q^{-i}$$

Main proof idea

We fix ϕ a linear form on EF and we study the probability to have $\phi(\mathbf{UV}) = \mathbf{0}$.

Lemma

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For a fixed \mathbf{U} , $\varphi_{\mathbf{U}} : \mathbf{v} \mapsto \phi(\mathbf{Uv})$ is a linear map from E^n to $\mathbb{F}_q^{n_1}$ so the distribution of $\phi(\mathbf{Uv})$ is uniform in $\text{Im}(\varphi_{\mathbf{U}}) \subset \mathbb{F}_q^{n_1}$.

We shall study the rank of $\varphi_{\mathbf{U}}$.

Indeed, when $\text{Rank}(\varphi_{\mathbf{U}}) = i$,

$$\mathbb{P}(\phi(\mathbf{Uv}) = \mathbf{0}) = q^{-i}$$

$$\mathbb{P}(\phi(\mathbf{UV}) = \mathbf{0}) = q^{-in_2}$$

Defining basis

By duality, the linear form ϕ is associated to a vector τ in EF such that $\phi(x) = \langle \tau, x \rangle$.

τ can be written :

$$\tau = \sum_{i=1}^s e_i f_i$$

where (e_1, \dots, e_s) and (f_1, \dots, f_s) are linearly independent elements in E and F . s is also called the tensor rank of τ .

These tuples can be completed to form basis $(e_1, \dots, e_s, \dots, e_r)$ and $(f_1, \dots, f_s, \dots, f_d)$.

Rank of $\varphi_{\mathbf{U}}$

We denote $\mathbf{U}_{ij}^{(k)}$ the coordinates of \mathbf{U}_{ij} in the basis we chose previously.

$$\begin{aligned}
 \varphi_{\mathbf{U}}((0, \dots, 0, \overset{j\text{-th}}{\underset{|}{e_k}}, 0, \dots, 0)) &= \phi(\mathbf{U}(0, \dots, 0, \overset{j\text{-th}}{\underset{|}{e_k}}, 0, \dots, 0)) \\
 &= (\phi(e_k \mathbf{U}_{ij}))_{1 \leq i \leq m_1} \\
 &= (\langle \tau, \sum \mathbf{U}_{ij}^{(l)} e_k f_l \rangle)_{1 \leq i \leq m_1} \\
 &= \begin{cases} (\mathbf{U}_{ij}^{(k)})_{1 \leq i \leq m_1} & k \leq s \\ \mathbf{0} & k > s \end{cases}
 \end{aligned}$$

Rank of $\varphi_{\mathbf{U}}$

So the matrix of $\varphi_{\mathbf{U}}$ looks like

$$\begin{array}{c}
 \begin{array}{c} \longleftarrow ns \qquad \qquad \qquad \longleftarrow n(d-s) \qquad \qquad \qquad \longrightarrow \\
 \begin{array}{c} \uparrow \\
 n_1 \\
 \downarrow \end{array} \\
 \begin{pmatrix} * & \cdots & * & 0 & \cdots & 0 \\
 * & \cdots & * & 0 & \cdots & 0 \\
 \vdots & & \vdots & \vdots & & \vdots \\
 \vdots & & \vdots & \vdots & & \vdots \\
 * & \cdots & * & 0 & \cdots & 0 \\
 * & \cdots & * & 0 & \cdots & 0 \end{pmatrix}
 \end{array}
 \end{array}$$

where each $*$ is an independent uniform random variable.

End of the proof

The rank $\varphi_{\mathbf{U}}$ thus follows the law of a random variable R_s .

$$\begin{aligned}\mathbb{P}(\text{Supp}(\mathbf{UV}) \subset \ker(\phi_{\tau})) &= \sum_{i=0}^{n_1} \mathbb{P}(\text{Supp}(\mathbf{UV}) \subset \ker(\phi_{\tau}) \mid \text{Rank}(\varphi_{\mathbf{U}}) = i) \\ &\quad \mathbb{P}(\text{Rank}(\varphi_{\mathbf{U}}) = i) \\ &= \sum_{i=0}^{n_1} q^{-in_2} \mathbb{P}(\text{Rank}(\varphi_{\mathbf{U}}) = i) \\ &= \sum_{i=0}^{n_1} q^{-in_2} \mathbb{P}(R_s = i) \\ &= \mathbb{E}(q^{-n_2 R_s}) \\ &\leq n_1 q^{-n_1 n_2}\end{aligned}$$

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Implementations

- Efficient
- Easy to use
- Isochronous (or constant-time) \Rightarrow no conditional branching on a secret expression

Long computations in LRPC codes cryptography

Definition (Key generation)

Let $U = (A|B)$ an LRPC matrix of weight d .

$$\begin{cases} pk &= H = (I|A^{-1}B) \\ sk &= U \end{cases}$$

Long computations in LRPC codes cryptography

Definition (Key generation)

Let $U = (x|y)$ an ideal LRPC matrix of weight d .

$$\begin{cases} pk &= H = (I|x^{-1}y) \\ sk &= U \end{cases}$$

Inversion in the field $\mathbb{K} := \mathbb{F}_{2^m}[X]/(P) \approx \mathbb{F}_{(2^m)^n}$

P is an irreducible polynomial of degree n with coefficients in \mathbb{F}_{2^m} .

Natural inversion algorithm

Find u, v such that $ux + vP = 1$.

i	quotient q_i	remainder r_i	u_i	v_i
0		P	1	0
1		x	0	1
...				
i	r_{i-2}/r_{i-1}	$r_{i-2} - q_i r_{i-1}$	$u_{i-2} - q_i u_{i-1}$	$v_{i-2} - q_i v_{i-1}$
...				
k	q_k	r_k	u_k	v_k
$k+1$	q_{k+1}	0		

TABLE: Extended Euclidean algorithm

⇒ can lead to **cache attacks**

A naive approach

Use Euclidean algorithm with naive isochronous techniques.

- Set the number of iterations to a constant.
- Make euclidean divisions isochronous \Rightarrow slow and difficult to implement.

Itoh-Tsuiji algorithm

Idea² : compute $x^{-1} = (x^r)^{-1} x^{r-1}$ where :

$$r = 1 + 2^m + 2^{2m} + \dots + 2^{(n-1)m} = \frac{2^{mn} - 1}{2^m - 1}$$

It is easy to prove that $x^r \in \mathbb{F}_{2^m}$.

This reduces the inversion in $\mathbb{F}_{(2^m)^n}$ to :

- The computation of x^{r-1} and x^r , which can easily be made isochronous ;
- An inversion in the smaller field \mathbb{F}_{2^m} ;
- n multiplications in \mathbb{F}_{2^m} .

2. Toshiya ITOH et Shigeo TSUJII. "A fast algorithm for computing multiplicative inverses in GF (2^m) using normal bases". In : *Information and computation* 78.3 (1988), p. 171–177.

Switching to normal basis

Usually, an element $x \in \mathbb{K} = \mathbb{F}_{(2^m)^n}$ is represented in the power basis $\{1, X, \dots, X^{n-1}\}$:

$$x = x_0 + x_1X + \dots + x_{n-1}X^{n-1}$$

with $x_i \in \mathbb{F}_{2^m}$.

But it can be more practical to use a normal basis :

$$x = x_0\alpha + x_1\alpha^{2^m} + \dots + x_{n-1}\alpha^{2^{(n-1)m}}$$

where α is chosen such that $(\alpha, \alpha^{2^m}, \alpha^{2^{2m}}, \dots, \alpha^{2^{(n-1)m}})$ is a basis of \mathbb{K} seen as a \mathbb{F}_{2^m} -vector space.

Characteristics of a normal basis

- Easy to perform operation $x \mapsto x^{2^m}$
- Multiplication : very expensive

$$\begin{aligned}x \cdot y &= \left(\sum_i x_i \alpha^{2^{im}} \right) \cdot \left(\sum_j y_j \alpha^{2^{jm}} \right) \\ &= \sum_{i,j} x_i y_j \alpha^{2^{im} + 2^{jm}} \\ &= \sum_{i,j,k} x_i y_j t_{i,j,k} \alpha^{2^{ik}}\end{aligned}$$

Except if you find an optimal normal basis

Optimal normal basis

Definition (Optimal normal basis)

A optimal normal basis is a basis $(\alpha, \alpha^{2^m}, \alpha^{2^{2m}}, \dots, \alpha^{2^{(n-1)m}})$ such that for all i , $\alpha\alpha^{2^{im}} = \alpha^{2^{a_i m}} + \alpha^{2^{b_i m}}$

Scheme	n	ONB?
ROLLO-I-128	83	✓
ROLLO-I-192	97	✗
ROLLO-I-256	113	✓
ROLLO-II-128	189	✓
ROLLO-II-192	193	✗
ROLLO-II-256	211	✗

TABLE: Existence of an optimal normal basis depending on the value n for each ROLLO set of parameters

Smarter square and multiply

$$r - 1 = 2^m + 2^{2m} + \dots + 2^{(n-1)m}$$

Square & Multiply $\Rightarrow n - 1$ multiplications.

We find a way to do $\log(n)$.

$$r - 1 = 2^m \left(\sum_{i=1}^{\log(n-1)} (2^{m2^i} - 1) 2^{m(t \bmod 2^i)} \right)$$

Performance results

	ROLLO-I-128	ROLLO-I-256	ROLLO-II-128
Non-isochronous algorithm ¹	1,030,500	1,702,620	4,295,704
	ROLLO-I-128	ROLLO-I-256	ROLLO-II-128
Isochronous algorithm ²	11,204,649		
Isochronous algorithm (our work ³)	3,514,016	5,785,700	22,859,614

TABLE: Duration of the key generation in CPU cycles

1. Nicolas ARAGON et al. *Rank-Based Cryptography Library*. URL : <https://rbc-lib.org/>.

2. Carlos AGUILAR-MELCHOR et al. "Constant time algorithms for ROLLO-I-128". In : *SN Computer Science 2.5* (2021), p. 1–19.

3. Carlos AGUILAR-MELCHOR et al. "Fast and Secure Key Generation for Low Rank Parity Check Codes Cryptosystems". In : *2021 IEEE International Symposium on Information Theory (ISIT)*. IEEE. 2021, p. 1260–1265.

Further refinement of our work

	ROLLO-I-128	ROLLO-I-256	ROLLO-II-128
Non-isochronous algorithm	1,030,500	1,702,620	4,295,704

	ROLLO-I-128	ROLLO-I-256	ROLLO-II-128
Isochronous algorithm	11,204,649		
Isochronous algorithm (our work)	3,514,016	5,785,700	22,859,614
Isochronous algorithm ¹	851,823	1,477,519	4,663,096

TABLE: Duration of the key generation in CPU cycles

1. [Tung CHOU et Jin-Han LIOU](#). "A Constant-time AVX2 Implementation of a Variant of ROLLO". In : *IACR Transactions on Cryptographic Hardware and Embedded Systems* (2022), p. 152–174.

State of the art implementation of ROLLO

Table 4: Cycle counts for key generation, encapsulation, and decapsulation of the ROLLO-I implementations from [AMAB⁺21] (the paper did not implement ROLLO-II), our ROLLO⁺ implementation, and the BIKE implementation from [CCK21].

instance	key gen.	encap.	decap	level	reference
ROLLO-I-128	11034623	984432	9775241	1	[AMAB ⁺ 21]
	11204649	320835	9744693		
ROLLO ⁺ -I-128	851823	30361	673666	1	this paper
ROLLO ⁺ -I-192	980860	38748	878398	3	
ROLLO ⁺ -I-256	1477519	55353	1635966	5	
ROLLO ⁺ -II-128	4663096	70621	876533	1	this paper
ROLLO ⁺ -II-192	4058419	94138	1060271	3	
ROLLO ⁺ -II-256	4947630	90021	1497315	5	
bikel1	589625	114256	1643551	1	[CCK21]
bikel3	1668511	267644	5128078	3	

Summary

- 1 Background on rank metric and LRPC codes
- 2 Presentation of LRPC-MS
- 3 Analysis of the decoding failure rate
- 4 Bonus : advances in LRPC implementations
- 5 Conclusion and perspectives**

Conclusion

- New rank metric based cryptosystem with competitive parameters and no ideal structure
- Probabilistic result on the support of the product of two random matrices
- Additional idea to make m down by 10 %
- The approach can generalize to RQC but is less efficient in that case

Thank you for your attention !

ANNEX

- An explicit method to build an optimal normal basis of $\mathbb{F}_{(2^m)^n}$ over \mathbb{F}_{2^m} .

Theorem

Let n be an integer prime to m and such that $2n + 1$ is a prime and assume that either :

- 1 *2 is primitive in \mathbb{Z}_{2n+1} , or*
- 2 *$2n + 1 = 3 \pmod{4}$ and 2 generates the quadratic residues in \mathbb{Z}_{2n+1} .*

Then $\alpha = \gamma + \gamma^{-1}$ generates an optimal normal basis of \mathbb{K} over \mathbb{F}_{2^m} , where γ is a primitive $(2n + 1)$ -th root of unity.