## Skew cyclic codes: Hamming distance and decoding algorithms ${ }^{1}$

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[^0]Based on:

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- GLN, A new perspective of cyclicity in convolutional codes, IEEE Transactions on Information Theory 62 (5) (2016), 2702-2706.
- GLN, A Sugiyama-like decoding algorithm for convolutional codes, 2016. arXiv:1607.07187.
- GLN, Ideal codes over separable ring extensions, IEEE Transactions on Information Theory 63 (5) (2017), 2796-2813.


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(2) Skew Reed-Solomon codes
(3) Decoding

## Convolutional codes

$\mathbb{F}$ a finite field, $k \leq n$ positive integers. A rate $k / n$ convolutional transducer $G$ transforms
into information sequences $\mathbf{u}=\ldots \mathbf{u}_{-1} \mathbf{u}_{\mathbf{0}} \mathbf{u}_{\mathbf{1}} \ldots\left(\mathbf{u}_{i} \in \mathbb{F}^{k}\right)$ code sequences $\mathbf{v}=\ldots \mathbf{v}_{-1} \mathbf{v}_{\mathbf{0}} \mathbf{v}_{\mathbf{1}} \ldots\left(\mathbf{v}_{i} \in \mathbb{F}^{n}\right)$.

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Requirements: Write $\mathbf{u}=\sum_{i=-i_{0}}^{\infty} \mathbf{u}_{i} t^{i} \in \mathbb{F}^{k}((t)) \cong \mathbb{F}((t))^{k}, \mathbf{v}=\sum_{j=-j_{0}}^{\infty} \mathbf{v}_{i} t^{j} \in \mathbb{F}^{n}((t)) \cong \mathbb{F}((t))^{n}$. Then

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\mathbf{v}=\mathbf{u} G,
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where $G$ is an $k \times n$ full rank matrix with entries in $\mathbb{F}(t)$.

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A rate $k / n$ convolutional code $\mathcal{D}$ over $\mathbb{F}$ is the image of a rate $k / n$ convolutional transducer $G$, that is

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## Definition (Vector space version)

A rate $k / n$ convolutional code over $\mathbb{F}$ is a $k$-dimensional vector subspace of $\mathbb{F}(t)^{n}$.

## Cyclic convolutional codes (module version)

## Lemma

The map $\mathcal{D} \mapsto \mathcal{D} \cap \mathbb{F}^{n}[t]$ is a bijection between the set of rate $k / n$ convolutional codes and the set of submodules of rank $k$ of $\mathbb{F}^{n}[t] \cong \mathbb{F}[t]^{n}$ that are direct summands.

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This idea has been further developed, (e.g. considering more general finite $\mathbb{F}$-algebras $A$ ) by Roos (1979), Gluesing-Luerssen/Schmale (2004), López-Permouth/Szabo (2013), GLN (2014, 2017), among others.

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Main difficulty/opportunity: Dealing with idempotents in $A[t ; \sigma]$.

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- The theory (including the algorithms) work for any field $L$.


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So, let $L$ be a field, and $\sigma: L \rightarrow L$ a field automorphism of finite order $n$. Set $K=L^{\sigma}$. Consider the ring $\mathcal{R}=L[x ; \sigma] /\left\langle x^{n}-1\right\rangle \cong M_{n}(K)$.
Note: The multiplication rule in $L[x ; \sigma]$ is $x a=\sigma(a) x$ for all $a \in L$.

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We have the isomorphism of $L$-vector spaces $\mathfrak{p}: L^{n} \rightarrow \mathcal{R}$ sending ( $c_{0}, c_{1}, \ldots, c_{n-1}$ ) onto $c_{0}+c_{1} x+\cdots+c_{n-1} x^{n-1}$.
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## Definition

A $k$-dimensional $L$-linear code $\mathcal{C} \subseteq L^{n}$ of dimension $n$ is said to be a skew cyclic code if $\mathfrak{p}(\mathcal{C})$ is a left ideal of $\mathcal{R}$.

Note: We will identify $\mathcal{C}$ with $\mathfrak{p}(\mathcal{C})$. Since every left ideal of $\mathcal{R}$ is principal, we will often speak of the "skew code generated by a polynomial", aggravating the abuse of language.

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Note: When $L=\mathbb{F}$ is a finite field, these codes lie in the realm of the theory developed by Boucher/Chaussade/Geiselmann/Loidreau/Ulmer (2007, 2009), among others, where the name is taken.

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Construction method of Skew RS codes

- Choose a normal basis $\left\{\alpha, \sigma(\alpha), \ldots, \sigma^{n-1}(\alpha)\right\}$ of the field extension $L / K$.


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- For $1 \leq k<n$, the left ideal $\mathcal{C}$ of $\mathcal{R}$ generated by

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g=\left[x-\beta, x-\sigma(\beta), x-\sigma^{2}(\beta), \ldots, x-\sigma^{n-k-1}(\beta)\right]_{\ell}
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## Theorem

The Hamming minimum distance of $\mathcal{C}$ is $\delta=n-k+1$. Thus, it is an MDS code.

## Parity check matrix

A parity check matrix of the skew RS code is given by

$$
H=\left(\begin{array}{cccc}
N_{0}(\beta) & N_{0}(\sigma(\beta)) & \ldots & N_{0}\left(\sigma^{n-k-1}(\beta)\right) \\
N_{1}(\beta) & N_{1}(\sigma(\beta)) & \ldots & N_{1}\left(\sigma^{n-k-1}(\beta)\right) \\
N_{2}(\beta) & N_{2}(\sigma(\beta)) & \ldots & N_{2}\left(\sigma^{n-k-1}(\beta)\right) \\
\vdots & \vdots & \ddots & \vdots \\
N_{n-1}(\beta) & N_{n-1}(\sigma(\beta)) & \ldots & N_{n-1}\left(\sigma^{n-k-1}(\beta)\right)
\end{array}\right) .
$$

Here, for $\gamma \in L$,

$$
N_{j}(\gamma)=\gamma \sigma(\gamma) \ldots \sigma^{j-1}(\gamma)
$$

is the remainder of the left division of $x^{j}$ by $x-\gamma$, that is

$$
x^{j}=q(x)(x-\gamma)+N_{j}(\gamma) .
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## Transmission and syndromes

Our skew RS code $\mathcal{C}$ is generated by

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$$
\begin{array}{ccc}
m \longrightarrow c=m g & \longrightarrow & y=c+e \\
\sum_{i=0}^{k-1} m_{i} x^{i} & \sum_{i=0}^{n-1} c_{i} x^{i} & \sum_{i=0}^{n-1} c_{i} x^{i}+\sum_{j=1}^{\nu} e_{j} x^{k_{j}} \quad \nu \leq \tau \\
\text { message } & & \text { received }
\end{array}
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\sum_{i=0}^{k-1} m_{i} x^{i} & \sum_{i=0}^{n-1} c_{i} x^{i} & \sum_{i=0}^{n-1} c_{i} x^{i}+\sum_{j=1}^{\nu} e_{j} x^{k_{j}}
\end{array} \quad \nu \leq \tau
$$

From the received polynomial $y=\sum_{j=0}^{n-1} y_{j} x^{j}$ we can compute the syndromes

$$
S_{i}=\sum_{j=0}^{n-1} y_{j} N_{j}\left(\sigma^{i}(\beta)\right), \quad i=0, \ldots, n-1
$$

## Error locator and error evaluator

We thus know the syndrome polynomial, defined as

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For $1 \leq j \leq \nu$ we thus know that there is $p_{j}$ of degree $\nu-1$ such that $\lambda=\left(1-\sigma^{k_{j}}(\beta) x\right) p_{j}$.

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\begin{equation*}
\omega=\sum_{j=1}^{\nu} e_{j} \sigma^{k_{j}}(\alpha) p_{j} \tag{1}
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## Sugiyama's decoding scheme.

If $\lambda$ is computed, then the error positions $k_{1}, \ldots, k_{\nu}$ are derived. With these at hand, we can compute $p_{1}, \ldots, p_{\nu}$. Finally, the error values $e_{1}, \ldots, e_{\nu}$ are computed by solving a linear system from (1), whenever $\omega$ is known.

## The non-commutative key equation

## Theorem

These polynomials satisfy the non-commutative key equation

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\omega=S \lambda+x^{2 \tau} u
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The non-commutative key equation

$$
\begin{equation*}
x^{2 \tau} u+S \lambda=\omega \tag{2}
\end{equation*}
$$

is a right multiple of the equation

$$
\begin{equation*}
x^{2 \tau} u_{I}+S v_{I}=r_{I} \tag{3}
\end{equation*}
$$

where $u_{I}, v_{I}$ and $r_{I}$ are the Bezout coefficients returned by the REEA with input $x^{2 \tau}$ and $S$, and $I$ is the index determined by the conditions $\operatorname{deg} r_{I-1} \geq \tau$ and $\operatorname{deg} r_{I}<\tau$. In particular, $\lambda=v_{I} g$ and $\omega=r_{I} g$ for some $g \in L[x ; \sigma]$.

## Sugiyama-like decoding

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## Sugiyama-like decoding

- If $(\lambda, \omega)_{r}=1$, then the last theorem gives that the REEA serves for the computation of the locator polynomial $\lambda$ and the evaluator polynomial $\omega$.
- As mentioned above, once these polynomials are computed, the error polynomial is computed by a scheme similar to the Sugiyama algorithm for (commutative) RS codes.
- The condition $(\lambda, \omega)_{r}=1$ is equivalent to

$$
\operatorname{deg} v_{I}=\operatorname{Cardinal}\left\{0 \leq i \leq n-1: 1-\sigma^{i}(\beta) x \text { left divides } v_{I}\right\}
$$

## Sugiyama-like decoding algorithm

Input: A polynomial $y=\sum_{i=0}^{n-1} y_{i} x^{i}$ received from the transmission of a codeword $c$ in a skew RS code $\mathcal{C}$ generated by $g=\left[\left\{x-\sigma^{i}(\beta)\right\}_{i=0, \ldots, n-k-1}\right]_{\ell}$ of error-correcting capacity $\tau=\left\lfloor\frac{n-k}{2}\right\rfloor$.
Output: A codeword $c^{\prime}$, or key equation failure.
1: for $0 \leq i \leq 2 \tau-1$ do
2: $\quad S_{i} \leftarrow \sum_{j=0}^{n-1} y_{j} N_{j}\left(\sigma^{i}(\beta)\right)$
$S \leftarrow \sum_{i=0}^{2 \tau-1} \sigma^{i}(\alpha) S_{i} x^{i}$
4: If $S=0$ then
Return $y$
$\left\{u_{i}, v_{i}, r_{i}\right\}_{i=0, \ldots, l} \leftarrow \operatorname{REEA}\left(x^{2 \tau}, S\right)$
$I \leftarrow$ first iteration in REEA with deg $r_{i}<\tau$, pos $\leftarrow \emptyset$
for $0 \leq i \leq n-1$ do
If $1-\sigma^{i}(\beta) x$ is a left divisor of $v_{I}$ then

$$
\text { pos }=p o s \cup\{i\}
$$

If $\operatorname{deg} v_{I}>\operatorname{Cardinal}(p o s)$ then
12: Return key equation failure
13: for $j \in \operatorname{pos}$ do
14: $\quad p_{j} \leftarrow \operatorname{right-quotient}\left(v_{I}, 1-\sigma^{j}(\beta) x\right)$
15: Solve the linear system $r_{I}=\sum_{j \in p o s} e_{j} \sigma^{j}(\alpha) p_{j}$
16: $e \leftarrow \sum_{j \in \text { pos }} e_{j} x^{j}$
17: Return $y-e$

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## Theorem

The key equation failure happens if and only if $e_{1}, \ldots, e_{\nu}$ are linearly dependent over $K=L^{\sigma}$.

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How often occurs the key equation failure? Well, it never happens if the error values are in "general position". More precisely,

## Theorem

The key equation failure happens if and only if $e_{1}, \ldots, e_{\nu}$ are linearly dependent over $K=L^{\sigma}$.
On the other hand, we have an algorithm to solve the key equation failure, if we insist.


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