

Decoding error-correcting codes with Gröbner bases

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Outline

- Introduction
- Unknown syndromes and MDS bases
- Decoding up to half the minimum distance
- Complexity of the algorithm

The decoding of cyclic codes up to half the BCH distance is well-known by Peterson, Arimoto and Gorenstein-Zierler,

by means of the **syndromes** s_i of a received word and the **error-locator polynomial** with coefficients σ_i .

Suppose that the defining set of the cyclic code contains $2t$ consecutive elements.

The **generalized Newton identities**

$$s_1 + \sigma_1 s_{i-1} + \cdots + \sigma_t s_{i-t} = 0, \quad i = t + 1, \dots, 2t.$$

are t linear equations in the variables $\sigma_1, \dots, \sigma_t$ with the known syndromes s_1, \dots, s_{2t} as coefficients.

Gaussian elimination solves this system of linear equations with complexity $\mathcal{O}(n^3)$.

This complexity was improved by the algorithm of Berlekamp-Massey and a variant of the Euclidean algorithm due to Sugiyama et al.

Both these algorithms are more efficient and are basically equivalent, but they decode up to the BCH error-correcting capacity, which is often strictly smaller than the true capacity.

They **do not correct up to the true error-correcting capacity**.

Gröbner bases techniques were addressed to remedy this problem.

These methods can be divided into the following categories:

- **Unknown syndromes** by Berlekamp and Tzeng-Hartmann-Chien,
- **Power sums** by Cooper and Chen-Reed-Helleseth-Truong,
- **Newton identities** by Augot-Charpin-Sendrier.

Our method is a generalization of the first one.

The theory of **Gröbner basis** is about solving systems of polynomial equations in several variables

It is as a common generalization of

- Linear Algebra,
linear systems of equations in several variables,
- Euclidean Algorithm,
polynomial equations of arbitrary degree in one variable.

The polynomial equations are **linearized** by treating the monomials as new variables. The number of variables grows exponentially in the degree of the polynomials.

The **complexity** of computing a Gröbner basis is doubly exponential in general, and **exponential** in our case of a finite set of solutions.

The complexity of our algorithm is exponential.

The **complexity coefficient** is measured under the **assumption** that the over-determined system of **quadratic equations** is **semi-regular** using the results of Bardet et al. applied to algorithm F5 of Faugère.

Let $\mathbf{b}_1, \dots, \mathbf{b}_n$ be a basis of \mathbb{F}_q^n .

B is the $n \times n$ matrix with $\mathbf{b}_1, \dots, \mathbf{b}_n$ as rows.

The **(unknown) syndrome** of a word \mathbf{e} with respect to B is the column vector $\mathbf{u}(\mathbf{e}) = \mathbf{u}(B, \mathbf{e}) = B\mathbf{e}^T$.

with entries $u_i(\mathbf{e}) = u_i(B, \mathbf{e}) = \mathbf{b}_i \cdot \mathbf{e}$ for $i = 1, \dots, n$.

The matrix B is invertible.

So the syndrome $\mathbf{u}(B, \mathbf{e})$ determines the error vector \mathbf{e} uniquely:

$$B^{-1}\mathbf{u}(B, \mathbf{e}) = B^{-1}B\mathbf{e}^T = \mathbf{e}^T.$$

The coordinatewise **star product** of $\mathbf{x}, \mathbf{y} \in \mathbb{F}_q^n$ by

$$\mathbf{x} * \mathbf{y} = (x_1 y_1, \dots, x_n y_n).$$

Then $\mathbf{b}_i * \mathbf{b}_j$ is a linear combination of the basis $\mathbf{b}_1, \dots, \mathbf{b}_n$.

There are **structure constants** $\mu_{ijl} \in \mathbb{F}_q$ such that

$$\mathbf{b}_i * \mathbf{b}_j = \sum_{l=1}^n \mu_{ijl} \mathbf{b}_l.$$

$\mathcal{U}(\mathbf{e})$ is the $n \times n$ matrix of (unknown) syndromes of a word \mathbf{e} with entries

$$u_{ij}(\mathbf{e}) = (\mathbf{b}_i * \mathbf{b}_j) \cdot \mathbf{e}.$$

The entries of $\mathcal{U}(\mathbf{e})$ and $\mathbf{u}(\mathbf{e})$ are related by

$$u_{ij}(\mathbf{e}) = \sum_{l=1}^n \mu_{ijl} u_l(\mathbf{e}).$$

Lemma

The rank of $\mathcal{U}(\mathbf{e})$ is equal to the weight of \mathbf{e} .

Let B_r be the $r \times n$ sub matrix of B with $\mathbf{b}_1, \dots, \mathbf{b}_r$ as rows.

$\mathbf{b}_1, \dots, \mathbf{b}_n$ is called an **MDS basis** and B an **MDS matrix** if all the $t \times t$ sub matrices of B_t have rank t for all $t = 1, \dots, n$.

Let C_t be the code with B_t as parity check matrix.

Proposition

B is an MDS matrix if and only if C_t is an $[n, n-t, t+1]$ code for all t .

MDS bases are known to exist if $n \leq q$.

Let $\mathbf{x} = (x_1, \dots, x_n)$ be n mutually distinct elements in \mathbb{F}_q .

Define

$$\mathbf{b}_i = (x_1^{i-1}, \dots, x_n^{i-1}).$$

Then $\mathbf{b}_1, \dots, \mathbf{b}_n$ with matrix $B(\mathbf{x})$ are MDS and are called a **Vandermonde** basis and matrix, resp.

If $\alpha \in \mathbb{F}_q^*$ is an element of order n and $x_j = \alpha^{j-1}$, then we get a **Reed-Solomon (RS)** basis and matrix with $b_i * b_j = b_{i+j-1}$ and $u_{ij}(\mathbf{e}) = u_{i+j-1}(\mathbf{e})$.

Proposition

Suppose that B is an MDS matrix.

Let $\mathcal{U}_{u,v}(\mathbf{e})$ be the $u \times v$ sub matrix of $\mathcal{U}(\mathbf{e})$ consisting of the first u rows and v columns.

Then

$$\text{rank}(\mathcal{U}_{uv}(\mathbf{e})) = \begin{cases} v & \text{if } v \leq \text{wt}(\mathbf{e}), \\ \text{wt}(\mathbf{e}) & \text{if } v > \text{wt}(\mathbf{e}). \end{cases}$$

Let C be an \mathbb{F}_q -linear code of length n , dimension k , minimum distance d , and redundancy $r = n - k$.

Choose a parity check matrix H of C .

Let $\mathbf{h}_1, \dots, \mathbf{h}_r$ be the rows of H .

There are constants $a_{ij} \in \mathbb{F}_q$ such that

$$\mathbf{h}_i = \sum_{j=1}^n a_{ij} \mathbf{b}_j.$$

Let A be the $r \times n$ matrix with entries a_{ij} . Then $H = AB$.

Let $\mathbf{y} = \mathbf{c} + \mathbf{e}$ be a **received word**
with $\mathbf{c} \in C$ a **code word** and \mathbf{e} an **error vector**.

The **syndromes** of \mathbf{y} and \mathbf{e} with respect to H are equal and **known**

$$s_i(\mathbf{y}) := \mathbf{h}_i \cdot \mathbf{y} = \mathbf{h}_i \cdot \mathbf{e} = s_i(\mathbf{e})$$

Expressed in the unknown syndromes of \mathbf{e} with respect to B :

$$s_i(\mathbf{y}) = \sum_{j=1}^n a_{ij} u_j(\mathbf{e}).$$

The system $E(\mathbf{y})$ of equations in the variables U_1, \dots, U_n is given by:

$$\sum_{l=1}^n a_{jl} U_l = s_j(\mathbf{y}) \text{ for } j = 1, \dots, r.$$

It consists of $n - k$ independent linear equations in n variables

The system $E(t)$ in the variables $U_1, \dots, U_n, V_1, \dots, V_t$ is given by:

$$\sum_{j=1}^t \sum_{l=1}^n \mu_{ijl} U_l V_j = \sum_{l=1}^n \mu_{it+1l} U_l \text{ for } i = 1, \dots, n.$$

It consists of n quadratic equations in $n + t$ variables.

The system of equations $E(t, \mathbf{y})$ is the union of $E(t)$ and $E(\mathbf{y})$.

It consists of $n - k$ linear equations in n variables and n quadratic equations in $n + t$ variables.

The linear equations are independent and used to eliminate $n - k$ variables.

Thus we get a system of n quadratic equations in $k + t$ variables.

Theorem

Let B be an MDS matrix with structure constants μ_{ijl} .

Let H be a parity check matrix of the code C such that $H = AB$.

Let $\mathbf{y} = \mathbf{c} + \mathbf{e}$ be a received word

with \mathbf{c} in C the codeword sent and \mathbf{e} the error vector.

Suppose that the weight of \mathbf{e} is not zero and at most $(d - 1)/2$.

Let t be the smallest positive integer such that $E(t, \mathbf{y})$ has a solution (\mathbf{u}, \mathbf{v}) over some extension \mathbb{F}_{q^m} of \mathbb{F}_q .

Then $\text{wt}(\mathbf{e}) = t$ and the solution is unique satisfying $\mathbf{u} = \mathbf{u}(\mathbf{e})$.

Experiments were done on an
AMD Athlon 64 Processor 2800+ (1.8MHz), 512MB RAM under Linux.

The computations of Gröbner bases were realized in SINGULAR 3-0-1.

Code	err. cap.	mindist.	GB dec.	no. of rec.	average
[25,11,4]	1	2.99	1.10	300	0.0037
[25,11,5]	2	21.58	2.89	300	0.0096
[25,8,5]	2	0.99	1.84	300	0.0061
[25,8,6]	2	3.38	1.79	300	0.0060
[25,8,7]	3	12.26	6.94	300	0.0231
[31,15]	2	-	10.76	300	0.0359
[31,15]	3	-	11.19	10	1.119

no. of err.	[120,40]	[120,30]	[120,20]	[120,10]	[150,10]
2	1	1	1	1	1
3	13	1	1	1	1
4	313	9	1	1	1
5	-	62	1	1	1
6	-	200	5	1	3
7	-	933	14	1	4
8	-	-	32	1	4
9	-	-	74	1	4
10	-	-	183	2	6
11	-	-	633	3	6
12	-	-	-	4	6
13	-	-	-	5	8
14	-	-	-	6	8
15	-	-	-	14	10
16	-	-	-	20	11
17	-	-	-	29	16
18	-	-	-	71	16
19	-	-	-	139	34
20	-	-	-	327	53
21	-	-	-	483	84
22	-	-	-	-	133
23	-	-	-	-	241
24	-	-	-	-	513

Given a decoding algorithm for a code C of rate R over \mathbb{F}_q of complexity $Compl(C)$,

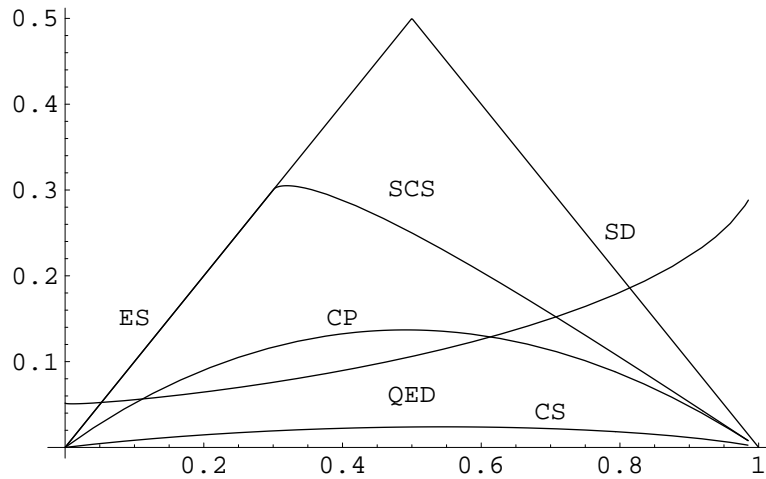
the **complexity coefficient** $CC(R)$ is defined as

$$CC(R) = \frac{1}{n} \log_q(Compl(C)).$$

In the binary case the complexity of our method is worse than exhaustive search.

But with increasing alphabet our method is better.
The following figure compares the complexity coefficients
for $q = 2^{10}$ of

- exhaustive search (ES),
- syndrome decoding (SD),
- systematic coset search (SCS),
- covering polynomials (CP),
- covering sets (CD) and
- our method using quadratic equations (QED).



Questions?