CSS.413.1 Topics in Coding Theory

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Page 3 1 Targets

1 Targets

The content of this course will be the followings:

- Introduction to Coding Theory: Definitions, Basic Properties, Linear Codes
- Reed Solomon Codes, Reed Muller Codes
- Decoding algorithms for Reed Solomon Codes:
 - Barlekamp-Welch Algorithm
 - Sudan's List Decoding Algorithm
 - Guruswami-Sudan List Decoding Algorithm upto the Johnson Bound
- Univariate Multiplicity Codes Decoding upto the List Decoding Capacity
- Bounds on the list size
- Local Decoding (LDC), Local Correction (LCC) of Codes
- Local Correction of Reed Muller Codes
- High Variate Locally correctable/decodable codes
- Local Decoding with constant queries Matching Vector Codes
- Private Information Retrieval Definitions, constructions
- Lower Bounds for LDCs Lower Bound for 2-query/4-query/Kalz-Trevisan/Alrabiah-Guruswami
- Local Testing of Codes:
 - Low-Degree Testing
 - Polischuk-Speilman Test
 - Friedl-Sudan Test
 - Arora-Sudan Test
 - Raz-Safra Test
- Applications: Explicit constructions
 - Combinatorial Designs
 - Subspace Designs
 - Derandomization
 - Hardness vs Randomness

2 Basics of Coding Theory

3 Decoding of Reed-Solomon Codes

4 Locally Decodable Codes and Locally Correctable Codes

5 Local Correction of Reed-Müller Codes

6 Multiplicity Codes

Multiplicity codes are a family of recently-introduced algebraic error-correcting codes based on evaluations of polynomials and their derivatives. Specifically, a codeword of a multiplicity code is obtained by evaluating a polynomial of degree at most k, along with all its derivatives of order < s, at n points of a finite field \mathbb{F}_q^m . These codes were introduced by Kopparty, Saraf and Yekhanin in [KSY14]. Notice that when s=1 this is basically the Reed-Solomon code when m=1 and Reed-Muller code when m>1.

6.1 Construction

Let $s, k, m \in \mathbb{Z}_0$ and let q be a prime power. Let $\Sigma = \mathbb{F}_q^{\binom{s+m-1}{m}}$. For $P(X_1, \dots, X_m) \in \mathbb{F}_q[X_1, \dots, X_m]$ we define the order s evaluations of P at $\mathbf{a} \in \mathbb{F}_q$ to be the vector $(P^{(\mathbf{i})}(\mathbf{a}))_{w(\mathbf{i} < s)} \in \Sigma$ where $wt(\mathbf{i}) = \sum\limits_{j=1}^m i_j$. Let E be a subset of n points in \mathbb{F}_q^m .

Definition 6.1.1: Multiplicity Codes

The multiplicity code of order-s evaluations of degree k polynomials in m variables over all points in E^m is the code over alphabet Σ , and has length n and for each polynomial $P(X) \in \mathbb{F}_q[X]$ with $\deg(P) \leq k$ the corresponding codeword is

$$\operatorname{Enc}_{s,k,m,q}(P) = (P^{(< s)}(\mathbf{a}))_{\mathbf{a} \in E} \in \Sigma^{n^m}$$

Our current interest is in the case m = 1. So

$$\operatorname{Enc}_{s,k,1}(P) = \begin{pmatrix} f(a_1) \\ f^{(1)}(a_1) \\ \vdots \\ f^{(s-1)}(a_1) \end{pmatrix}, \begin{bmatrix} f(a_2) \\ f^{(1)}(a_2) \\ \vdots \\ f^{(s-1)}(a_2) \end{bmatrix}, \cdots, \begin{bmatrix} f(a_n) \\ f^{(1)}(a_n) \\ \vdots \\ f^{(s-1)}(a_n) \end{bmatrix}$$

Remark: The above encoding in not the encoding

$$\operatorname{Enc}_{s,k,1} = \left(f(a_1), f^{(1)}(a_1), \cdots, f^{(s-1)}(a_1), f(a_2), f^{(1)}(a_2), \cdots, f^{(s-1)}(a_2), \cdots, f^{(s-1)}(a_n), f^{(1)}(a_n), \cdots, f^{(s-1)}(a_n) \right)$$

Each alphabet of the codeword is a vector of size s. The same holds for the multivariate case

The above operation of treating a vector as a single alphabet is called *folding*.

6.2 Rate and Distance of Multiplicity Codes

We will now calculate the rate and the distance of the code. The block length is n^m . Since we are evaluating all the derivatives of order < s, the alphabet size is $q^{\binom{s+m-1}{m}}$. So the number of codewords is $\left(q^{\binom{s+m-1}{m}}\right)^{n^m} = q^{n^m\binom{s+m-1}{m}}$. The number of polynomials in m variables of degree at most k is $q^{\binom{k+m}{m}}$. So the rate of the code is

$$R = \frac{\binom{k+m}{m}}{n^m \binom{s+m-1}{m}} \approx \left(\frac{k}{ns}\right)^m$$

Now using the Multiplicity Schwartz-Zippel lemma we can calculate the distance of the code. We have the relative distance to be $\delta = 1 - \frac{k}{ns}$.

Theorem 6.2.1

The rate and the distance of the multiplicity code are $R = \frac{\binom{k+m}{m}}{n^m\binom{s+m-1}{m}} \approx \left(\frac{k}{ns}\right)^m$ and $\delta = 1 - \frac{k}{ns}$ respectively.

We usually think *m* and *s* to be large constant. So as multiplicity code achieves the Singleton bound asymptotically.

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6.3 List Decoding of Univariate Multiplicity Codes up to Capacity

Since we are interested in univariate multiplicity codes, we will set m=1. So we have three parameters k, s, n and the field size q. Therefore, as we have calculated before the rate and distance of the univariate multiplicity code are $R=\frac{k+1}{ns}\approx\frac{k}{sn}$ and $\delta=1-\frac{k}{ns}$ respectively.

6.3.1 Polynomial List Size up to Capacity

Theorem 6.3.1 [Kop15, GW11]

For every $\epsilon \in (0,1)$, there exists $s_0 \approx \frac{1}{\epsilon^2}$ such that the univariate multiplicity code with multiplicity parameter $s > s_0$ can be efficiently list decodable from $\left(1 - \frac{k}{ns} - \epsilon\right)$ fraction of errors.

We will give the proof in [GW11]. It uses polynomial method based arguments. This proof has two steps.

Step 1: Interpolation

Step 2: Reconstruction of close enough codewords

So assume the received word is $w = (\alpha_0, \beta_{i,0}, \beta_{i,1}, \dots, \beta_{i,s-1})_{i=1}^n$ and also consider the parameter $t = \sqrt{s} \approx \frac{1}{\epsilon}$. With this we will show the proof of the above theorem.

Proof: Step 1: Interpolation

In step 1 we will look for an m+1 variate polynomial $Q(X, Y_1, \ldots, Y_t)$ which is linear in Y_i 's i.e.

$$Q(X, Y_1, \dots, Y_m) = A_0(X) + A_1(X)Y_1 + \dots + A_t(X)Y_t$$

Let f is a close enough polynomial. Then define $R_f(X) = Q(X, f(X), f^{(1)}(X), \dots, f^{(t-1)}(X))$. Then we want $R_f(X) \equiv 0$. And also we want whenever f and the received word agree on some point $R_f(X)$ has a zero of high multiplicity at that point. So let f agrees with the received word at α . Then

$$R_f(\alpha_i) = Q(\alpha_i, f(\alpha_i), f^{(1)}(\alpha_i), \dots, f^{(t-1)}(\alpha_i)) = Q(\alpha_i, \beta_{i,0}, \beta_{i,1}, \dots, \beta_{i,t-1}) = 0$$

Now

$$R_f^{(1)}(X) = \frac{dA_0}{dX}(X) + \frac{d}{dX} \left(\sum_{i=1}^t A_i(X) f^{(i-1)}(X) \right) = \frac{dA_0}{dX}(X) + \sum_{i=1}^t \frac{dA_i}{dX}(X) \cdot f^{(i-1)}(X) + A_i(X) \cdot f^{(i)}(X)$$

Therefore

$$R_f^{(1)}(\alpha_i) = \frac{dA_0}{dX}(\alpha_0) + \sum_{i=1}^t \frac{dA_j}{dX}(\alpha_i) \cdot \beta_{i,j-1} + A_j(\alpha) \cdot \beta_{i,j}$$

So we want as many derivatives of R_f to be zero as possible.

Observation 1. In $R_f^{(k)}(X)$ we needed the evaluations of $\beta_{i,0}, \ldots, \beta_{i,t-1}, \beta_{i,t}, \ldots, \beta_{i,t+i-1}$.

Since we have evaluations till $(s-1)^{th}$ order derivative we can only take derivative of R_f upto order (s-t). So we want $R_f^{(k)}(\alpha_i) \equiv 0$ for all $k \in \{0, \dots, s-t\}$. And Q follows the following properties:

- $deg(A_i) \leq D$
- For all $i \in [n]$, $R_f^{(k)}(\alpha_i) \equiv 0$ for all $k \in \{0, \dots, s-t\}$. To make it simple define the operator Ψ as

$$\Psi(Q) := A_0^{(1)}(X) + \sum_{i=0}^t (A_i^{(1)}(X)Y_i + A_i(X)Y_{i+1})$$

and
$$\Psi^i(Q) = \Psi(\Psi^{i-1}(Q)), \Psi^0(Q) = Q$$
. Then $\forall i \in [n], \forall j \in \{0, \dots, s-t\}, \Psi^j(Q)(\alpha, \overline{\beta}_i) = 0$.

Observation 2. Each point of agreement of f is a root of R_f of multiplicity at least s - t + 1.

Now for step 1 to return a Q successfully we need the number of variables to be more than the number of equations. The number of variables is (t + 1)(D + 1). The number of equations is n(s - t + 1). So we need

$$(t+1)(D+1) > n(s-t+1) \iff D+1 > \frac{n(s-t)}{t+1}$$

Hence enough to take $D = \frac{n(s-t+1)}{t+1}$. Then step 1 returns a nonzero Q.

Observation 3. If f has agreement $> \frac{D+k}{s-t+1}$ with the received word then $R_f(X) \equiv 0$ as $\deg(R_f) \leq D+k$ and each point of agreement is a zero of multiplicity at least s-t+1.

So the number of agreement $> \frac{D+k}{s-t+1} = \frac{\frac{n(s-t+1)}{t+1}+k}{s-t+1} = \frac{n}{t+1} + \frac{k}{s} \cdot \frac{s}{s-t+1} \approx \epsilon n + \frac{k}{s}$ since we take s to be constant.

Step 2: Reconstruction of close enough codewords

Find all degree k, f(X) such that

Step 2.1: $Q(X, f(X), f^{(1)}(X), \dots, f^{(t-1)}(X)) \equiv 0$ [This step looks like solving a differential equation]

Step 2.2: *f* has large agreement with the received word.

For the step 2.1 let f_1, \ldots, f_l are the solutions. Then any linear combination of them is also a solution. Hence the space of solutions of f is a vector space over the field. We need to argue that the dimension of this space is at small. Let S be the set of all $f \in \mathbb{F}_q[X]$ such that $\deg(f) \leq k$ and $Q(X, \overline{f}(X)) \equiv 0$. Then by Lemma 6.3.2 we have $\dim S \leq t$. Since s is constant, t is also constant. So the number of solutions is at most q^t . Hence the list size is at most $q^t = \operatorname{poly}(n)$.

Lemma 6.3.2

S is a subspace of dimension at most t.

Proof: WLOG we can assume $A_i(0) \neq 0$ for all $i \in \{0, ..., t\}$ otherwise we can do a random shift to make it nonzero. Let $f \in S$, then

$$Q(X, f(X), f^{(1)}(X), \dots, f^{(t-1)}(X)) = A_0(X) + A_1(X) \cdot f(X) + A_2(X) f^{(1)}(X) + \dots + A_t(X) \cdot f^{(t-1)}(X) \equiv 0$$

Let $A_i(X) = \sum_{j=0}^D A_{i,j} X^j$ for all $i \in \{0, \dots, t\}$ and $f(X) = \sum_{j=0}^k f_j X^j$. Therefore $f^{(i)}(X) = \sum_{j=0}^{k-i} \frac{(i+j)!}{j!} f_{i+j} X^j$. Then the coefficient of X^i in $Q(X, f(X), f^{(1)}(X), \dots, f^{(t-1)}(X))$ is

$$A_{0,i} + \left(\sum_{j=0}^{i} A_{1,j} \cdot f_{i-j}\right) + \left(\sum_{j=0}^{i} A_{2,j} \cdot (i+1-j)f_{i+1-j}\right) + \dots + \left(\sum_{j=0}^{i} A_{t,j} \cdot \frac{(t-1+i-j)!}{(i-j)!} f_{t-1+i-j}\right)$$

$$= A_{0,i} + \sum_{j=0}^{t} \sum_{i=0}^{i} A_{l,i-j} \cdot \frac{(l-1+j)!}{j!} f_{l-1+j}$$

Since f is a solution this coefficient is 0. Notice that in the above linear equation coefficient of X^i depends on f_j for all j < i + t and term multiplied to

6.3.2 Constant List Size up to Capacity

Theorem 6.3.3 [KRZSW18]

The list size above is of constant size only depends on ϵ and independent of the block length.

6.4 Local Correction of Multiplicity Codes

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7 References

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