### Error Correcting Codes: Combinatorics, Algorithms and Applications(Fall 2007)

### Lecture 27: Berlekamp-Welch Algorithm

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In the last lecture, we discussed unique decoding of RS codes and briefly went through the Berlekamp-Welch algorithm. In today's lecture we will study Berlekamp-Welch algorithm in more detail.

Recall that the  $[n, k, n-k+1]_q$  Reed-Solomon code regards a message as a polynomial P(X) of a degree at most k-1, and the encoding of a message  $\mathbf{m} = (m_0, \ldots, m_{k-1})$  is  $(P(\alpha_1), \ldots, P(\alpha_n))$ . Here,  $m_i \in \mathbb{F}_q$ ,  $k \leq n \leq q$ , and  $P(X) = \sum_{i=0}^{k-1} m_i X^i$ . Now let us look at the decoding problem of Reed-Solomon codes. Suppose we are given distinct values  $\alpha_1, \ldots, \alpha_n$  where  $\alpha_i \in \mathbb{F}_q$  with received word  $\mathbf{y} = (y_1, \ldots, y_n) \in \mathbb{F}_q^n$  and parameters k and  $e < \frac{n-k+1}{2}$ , where e is an upper bound on the number of errors which occurred during transmission. Our goal is to find a polynomial  $P(X) \in \mathbb{F}_q[X]$  of degree at most k-1, such that  $P(\alpha_i) \neq y_i$  for at most e values of  $i \in [n]$ (assuming such a P(X) exists). Although this problem is quite non-trivial one, a polynomial time solution can be found for this problem. This solution dates back to 1960 when Peterson [2] came up with a decoding algorithm for the more general BCH code that runs in time  $O(n^3)$ . Later Berlekamp and Massey sped up this algorithm so that it runs in  $O(n^2)$ . There is an implementation using Fast Fourier Transform that runs in time  $O(n \log n)$ . We will not discuss these faster algorithms, but will study another algorithm due to Berlekamp and Welch. More precisely, we will use the Gemmell-Sudan description of the Berlekamp-Welch algorithm[1].

# **1** Berlekamp-Welch Algorithm

We start by describing the Berlekamp-Welch algorithm.

**Input:**  $n \ge k \ge 1, 0 < e < \frac{n-k+1}{2}$  and n pairs  $\{(\alpha_i, y_i)\}_{i=1}^n$  with  $\alpha_i$  distinct. Let  $\mathbf{y} = (y_1, \dots, y_n)$ .

**Output:** Polynomial P(X) of degree at most k - 1 or fail.

**Step 1:** Compute a non-zero polynomial E(X) of degree exactly e, and a polynomial Q(X) of degree at most e + k - 1 such that

$$y_i E(\alpha_i) = Q(\alpha_i) \quad 1 \le i \le n.$$
(1)

If such polynomials do not exist output fail.

**Step 2:** If E(X) does not divide Q(X) output fail else compute  $P(X) = \frac{Q(X)}{E(X)}$ . If  $\Delta(\mathbf{y}, (P(\alpha_i))_i) > e$  then output fail else output P(X).

**Remark 1.1.** Notice that computing E(X) a as hard as finding the solution polynomial P(X). If one knew E(X), then one can use erasure decoding for RS codes to recover P(X). Also in some cases, the polynomial Q(X) is as hard to find as E(X). E.g., given Q(X) and  $\mathbf{y}$  (such that  $y_i \neq 0$ for  $1 \leq i \leq n$ ) one can find the error locations by checking positions where  $Q(\alpha_i) = 0$ . While each of these polynomials e(X), Q(X) is hard to find individually, together they are easier to find.

Note that if the algorithm does not output fail, then the algorithm produces a correct output. Thus, to prove the correctness of the Berlekamp-Welch algorithm, we just need the following result.

**Theorem 1.2.** If  $(P(\alpha_i))_i$  is transmitted (where P(X) is a polynomial of degree at most k - 1) and at most e errors occur, then the Berlekamp-Welch algorithm outputs P(X).

The proof of the theorem above follows from the subsequent claims.

**Claim 1.3.** There exist a pair of polynomials E(X) and Q(X) that satisfy **Step 1** such that  $\frac{Q(X)}{E(X)} = P(X)$ .

Note that now it suffices to argue that  $\frac{Q_1(X)}{E_1(X)} = \frac{Q_2(X)}{E_2(X)}$  for any pair of solutions that satisfy **Step 1** since Claim 1.3 above can then be used to see that ratio must be P(X). Indeed, we will show this to be the case:

**Claim 1.4.** If any two distinct solutions  $(E_1(X), Q_1(X)) \neq (E_2(X), Q_2(X))$  satisfy **Step 1**, then they will satisfy

$$\frac{Q_1(X)}{E_1(X)} = \frac{Q_2(X)}{E_2(X)}.$$

**Proof of Claim 1.3.** We just take E(x) to be the error-locating polynomial for P(X) and let Q(X) = P(X)E(X) where  $\deg(Q(X)) \le \deg(P(X)) + \deg(E(X)) \le e + k - 1$ . In particular, define E(X) as the following polynomial of degree exactly e:

$$E(X) = X^{e-\Delta(\mathbf{y},(P(\alpha_i))_i)} \prod_{1 \le i \le n | y_i \ne P(\alpha_i)} (X - \alpha_i)$$
(2)

By definition, E(X) is a degree *e* polynomial with the following property:

$$E(\alpha_i) = 0$$
 iff  $y_i \neq P(\alpha_i)$ 

We now argue that E(X) and Q(X) satisfy (1). Note that if  $E(\alpha_i) = 0$ , then  $Q(\alpha_i) = P(\alpha_i)E(\alpha_i) = y_i E(\alpha_i) = 0$ . When  $E(\alpha_i) \neq 0$ , we know  $P(\alpha_i) = y_i$  and so we still have  $P(\alpha_i)E(\alpha_i) = y_iE(\alpha_i)$ , as desired.

**Proof of Claim 1.4.** Note that the degrees of the polynomials  $Q_1(X)E_2(X)$  and  $Q_2(X)E_1(X)$  are at most 2e + k - 1. Let us define polynomial R(X) with degree at most 2e + k - 1 as follows:

$$R(X) = Q_1(X)E_2(X) - Q_2(X)E_1(X).$$
(3)

Furthermore, from **Step 1** we have, for every  $i \in [n]$ ,

$$y_i E_1(\alpha_i) = Q_1(\alpha_i)$$
 and  $y_i E_2(\alpha_i) = Q_2(\alpha_i).$  (4)

Substituting (4) into (3) we get for  $1 \le i \le n$ :

$$R(\alpha_i) = (y_i E_1(\alpha_i)) E_2(\alpha_i) - (y_i E_2(\alpha_i)) E_1(\alpha_i)$$
  
= 0.

The polynomial R(X) has n roots and

$$\deg(R(X)) \leq e+k-1+e$$
  
= 2e+k-1  
< n,

Where the last inequality follows from the upper bound on e. Since  $\deg(R(X)) < n$ , then we get that the polynomials  $Q_1(X)E_2(X)$  and  $Q_2(X)E_1(X)$  agree on more points than their degree, and hence they are identical. Note that as  $E_1(X) \neq 0$  and  $E_2(X) \neq 0$ , this implies that  $\frac{Q_1(X)}{E_1(X)} = \frac{Q_2(X)}{E_2(X)}$ , as desired.

#### **1.1 Implementation of Berlekamp-Welch Algorithm**

In Step 1, Q(X) has e + k unknowns and E(X) has e + 1 unknowns. For each  $1 \le i \le n$ , the constraint in (1) is a linear equation in these unknowns.

Thus, we have a system of n linear equations in 2e + k + 1 < n + 2 unknowns. By claim 1.3, these system of equations have a solution. The only extra requirement is that the degree of polynomial E(X) should be exactly e. We have already shown E(X) in equation (2) to satisfy this requirement. So we add a constraint that the coefficient of  $X^e$  in E(X) is 1. We have n + 1 linear equation in at most n + 1 variables, which we can solve in time  $O(n^3)$ , e.g. by Gaussian elimination. Finally, note that **Step 2** can be implemented in time  $O(n^3)$  by "long division."

**Theorem 1.5.** For any  $[n, k]_q$  Reed-Solomon code it can be uniquely decoded in  $O(n^3)$  time up to  $\frac{D}{2} = \frac{n-k+1}{2}$  number of errors.

## 2 Errors And Erasures decoding of RS codes

We have seem that there exists polynomial time constructible codes that lie on the Zyablov bound and can be corrected up to 1/4 of their designed distance. Next, we look into codes that correct up to  $\frac{1}{2}$  design distance for explicit codes on the Zyablov bound. For this one we will need correction from *both* errors and erasures. Berlekamp-Welch algorithm corrects up to  $e < \frac{D}{2}$  errors for  $[N, K, D]_Q$  RS codes. We have already seen that we can correct s < D erasures.<sup>1</sup> We will prove the following more general result in the next lecture:

<sup>&</sup>lt;sup>1</sup>The number of unknown positions is s and the number of known position is  $N - s \ge N - D + 1 = K$ . Having K constrains we can recover polynomial of degree at most K - 1 by interpolation (K unknowns and K equations) in  $O(N^3)$  time.

**Proposition 2.1.** Reed-Solomon codes can be corrected from e errors and s erasures as long as 2e + S < D in  $O(N^3)$  time.

# References

- [1] P. Gemmell and M. Sudan. Highly resilient correctors for polynomials. *Information processing letters*, 43(4):169–174, September 1992.
- [2] W. Wesley Peterson. Encoding and error-correction procedures for bose-chadhuri codes. *IEEE Transactions on Information Theory*, 6:459–470, 1960.