

Lecture 39: GS Decoder Wrap-up

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In the last lecture, we introduced the Guruswami-Sudan (GS) list decoding algorithm for RS codes. We restate the algorithm here. Recall that the input is the received word, in the form  $(\alpha_i, y_i) \in \mathbb{F}^2$ , for  $1 \leq i \leq n$ .

**GS list decoding algorithm**

Inputs:  $(\alpha_i, y_i) \in \mathbb{F}^2$  for  $1 \leq i \leq n$ , agreement parameter  $0 \leq t \leq n$

**Step 1)** Compute a non-zero  $Q(X, Y)$ , such that

- i)  $(1, k)$  weighted degree of  $Q \leq D$ .
- ii)  $Q$  has  $r$  roots at  $(\alpha_i, y_i)$ ,  $1 \leq i \leq n$ .

**Step 2)** Output all degree  $\leq k$  polynomials  $P(X)$  such that

- i)  $Y - P(X)$  divides  $Q(X, Y)$
- ii)  $P(\alpha_i) = y_i$  for at least  $t$  positions  $i$ .

We analyzed the GS list decoding algorithm modulo two lemmas. In today's lecture, we are going to prove the two lemmas.

## 1 Proof of key lemmas

We now recall the two lemmas:

**Lemma 1.1.** *The condition that  $Q(X, Y)$  has  $r$  roots at  $(\alpha, \beta)$  implies  $\binom{r+1}{2}$  constraints on the coefficients of  $Q$ .*

**Lemma 1.2.** *If  $Q(X, Y)$  is output by **Step 1** and  $P(X)$  needs to be output in **Step 2**, then  $Y - P(X)$  divides  $Q(X, Y)$ .*

**Proof of Lemma 1.1.** Let

$$Q(X, Y) = \sum_{\substack{i,j \\ i+kj \leq D}} q_{i,j} X^i Y^j$$

and  $Q_{\alpha,\beta}(X, Y) = Q(X + \alpha, Y + \beta) = \sum_{i,j} q_{i,j}^{\alpha,\beta} X^i Y^j$ . We will show that

(i)  $q_{i,j}^{\alpha,\beta}$  are homogeneous linear combinations of  $q_{i,j}$ 's.

(ii) If  $Q_{\alpha,\beta}(X, Y)$  has no monomial of degree  $< r$ , then that implies  $\binom{r+1}{2}$  constraints on  $q_{i,j}^{\alpha,\beta}$ 's.

Note that (i) and (ii) prove the lemma. To prove (i), note that by the definition:

$$Q_{\alpha,\beta}(X, Y) = \sum_{i,j} q_{i,j}^{\alpha,\beta} X^i Y^j \quad (1)$$

$$= \sum_{\substack{i',j' \\ i'+kj' \leq D}} q_{i',j'} (X + \alpha)^{i'} (Y + \beta)^{j'} \quad (2)$$

Note that, if  $i > i'$  or  $j > j'$ , then  $q_{i,j}^{\alpha,\beta}$  doesn't depend on  $q_{i',j'}$ . By comparing coefficients of  $X^i Y^j$  from (1) and (2), we obtain

$$q_{i,j}^{\alpha,\beta} = \sum_{\substack{i'>i \\ j'>j}} q_{i',j'} \binom{i'}{i} \binom{j'}{j} \alpha^i \beta^j,$$

which proves (i). To prove (ii), recall that by definition  $Q_{\alpha,\beta}(X, Y)$  has no monomial of degree  $< r$ . In other words, we need to have constraints  $q_{i,j}^{\alpha,\beta} = 0$  if  $i + j \leq r - 1$ . The number of such constraints is

$$|\{(i, j) | i + j \leq r - 1, i, j \in \mathbb{Z}^{\geq 0}\}| = \binom{r+1}{2},$$

where the equality follows from the argument we used to bound the dimension of Reed-Muller codes.  $\square$

We now re-state Lemma 1.2 more precisely and then prove it.

**Lemma 1.3.** *Let  $Q(X, Y)$  be computed by **Step 1**. Let  $P(X)$  be a polynomial of degree  $\leq k$ , such that  $P(\alpha_i) = y_i$  for at least  $t > \frac{D}{r}$  many values of  $i$ , then  $Y - P(X)$  divides  $Q(X, Y)$ .*

*Proof.* Define

$$R(X) \triangleq Q(X, P(X)).$$

As usual, to prove the lemma, we will show that  $R(X) \equiv 0$ . To do this, we will need the following claim.

**Claim 1.4.** *If  $P(\alpha_i) = y_i$ , then  $(X - \alpha_i)^r$  divides  $R(X)$ , that is  $\alpha_i$  is a root of  $R(X)$  with multiplicity  $r$ .*

Note that by definition of  $Q(X, Y)$  and  $P(X)$ ,  $R(X)$  has degree  $\leq D$ . Assuming the above claim is correct,  $R(X)$  has at least  $tr$  roots. Therefore,  $R(X)$  is a zero polynomial as  $tr > D$ . We will now prove Claim 1.4. Define

$$P_{\alpha_i, y_i}(X) \triangleq P(X + \alpha_i) - y_i, \quad (3)$$

and

$$R_{\alpha_i, y_i}(X) \triangleq R(X + \alpha_i) \quad (4)$$

$$= Q(X + \alpha_i, P(X + \alpha_i)) \quad (5)$$

$$= Q(X + \alpha_i, P_{\alpha_i, y_i}(X) + y_i) \quad (6)$$

$$= Q_{\alpha_i, y_i}(X, P_{\alpha_i, y_i}(X)), \quad (7)$$

where the second, third and fourth equalities follow from the definitions of  $R(X)$ ,  $P_{\alpha_i, y_i}(X)$  and  $Q_{\alpha_i, y_i}(X, Y)$  respectively.

By (4) if  $R_{\alpha_i, y_i}(0) = 0$ , then  $R(\alpha_i) = 0$ . So if  $X$  divides  $R_{\alpha_i, y_i}(X)$ , then  $X - \alpha_i$  divides  $R(X)$ . Similarly, if  $X^r$  divides  $R_{\alpha_i, y_i}(X)$ , then  $(X - \alpha_i)^r$  divides  $R(X)$ . Thus, to prove the lemma, we will show that  $X^r$  divides  $R_{\alpha_i, y_i}(X)$ . Since  $P(\alpha_i) = y_i$  when  $\alpha_i$  agrees with  $y_i$ , we have  $P_{\alpha_i, y_i}(0) = 0$ . Therefore,  $X$  is a root of  $P_{\alpha_i, y_i}(X)$ , that is,  $P_{\alpha_i, y_i}(X) = X \cdot g(X)$  for some polynomial  $g(X)$  of degree at most  $k - 1$ . We can rewrite  $R_{\alpha_i, y_i}(X) = \sum_{i', j'} q_{i', j'}^{\alpha_i, y_i} X^{i'} (P_{\alpha_i, y_i}(X))^{j'} = \sum_{i', j'} q_{i', j'}^{\alpha_i, y_i} X^{i'} (Xg(X))^{j'}$ . Now for every  $i', j'$  such that  $q_{i', j'}^{\alpha_i, y_i} \neq 0$   $i' + j' \geq r$  as  $Q_{\alpha_i, y_i}(X, Y)$  has no monomial of degree  $< r$ . Thus  $R_{\alpha_i, y_i}(x)$  has no non-zero monomial  $X^\ell$ ,  $\ell < r$ . Thus  $X^r$  divides  $R_{\alpha_i, y_i}(X)$ , as desired.  $\square$

From the second property of **Step 1**,  $Q(X, Y)$  has  $r \geq 0$  roots at  $(\alpha_i, y_i)$ ,  $1 \leq i \leq n$ . However, our analysis did not explicitly use the fact that the multiplicity is same for every  $i$ . In particular, given non-zero integer multiplicities  $w_i \geq 0$ ,  $1 \leq i \leq n$ , the GS algorithm can output all polynomials  $P(X)$  of degree at most  $k$ , such that

$$\sum_{i: P(\alpha_i)=y_i} w_i > \sqrt{kn \sum_{i=0}^n \binom{w_i + 1}{2}}$$

Note that till now we have seen the special case  $w_i = r$ ,  $1 \leq i \leq n$ . Further, note that the  $\alpha_i$ 's need not be distinct for the all of the previous arguments to go through. In particular, we can further generalize the input which now has positive integer weights  $w_{i, \alpha}$  for every  $1 \leq i \leq n$  and  $\alpha \in \mathbb{F}$  and the GS algorithm will be able to output all  $P(X)$  of degree at most  $k$  such that

$$\sum_i w_{i, P(\alpha_i)} > \sqrt{kn \sum_{i=0}^n \sum_{\alpha \in \mathbb{F}} \binom{w_{i, \alpha} + 1}{2}}.$$

This will be useful to solve the following generalization of list decoding called soft decoding.

**Definition 1.5.** *Under soft decoding problem, the decoder is given as input a set of non-negative weights  $w_{i, d}$  ( $1 \leq i \leq n, \alpha \in \mathbb{F}_q$ ) and a threshold  $W \geq 0$ . The soft decoder needs to output all codewords  $(c_1, c_2, \dots, c_n)$  in  $q$ -ary code of block length  $n$  that satisfy:*

$$\sum_{i=1}^n w_{i, c_i} \geq W.$$

Consider the following special case of soft decoding where  $w_{i,y_i} = 1$  and  $w_{i,\alpha} = 0$  for  $\alpha \in \mathbb{F} \setminus \{y_i\}$  ( $1 \leq i \leq n$ ). Note that this is exactly the list decoding problem with the received word  $(y_1, \dots, y_n)$ . Thus, list decoding is indeed a special case of soft decoding. Soft decoding has practical application in setting where the channel is analog. In such a situation, the “quantizer” might not be able to pinpoint a received symbol  $y_i$  with 100% accuracy. Instead it can use the weight  $w_{i,\alpha}$  to denote its confidence level that  $i$ th received symbol was  $\alpha$ . Soft decoding (and its special case list recovery, which we will study in the next lecture) also has application in designing list decoding algorithm for concatenated codes.