

Remark: The codewords of $RM(m, r)$ are thus the evaluations of $f: \deg(f) \leq r$ at all points of the hypercube $\{0, 1\}^m$, taken in some fixed order. A natural ordering of points of the hypercube is the **lexicographic ordering**:

000...000 \rightarrow 000...001 \rightarrow 000...010 \rightarrow 000...011 \rightarrow ...
 \rightarrow 111...111

The following properties thus hold:

(0) $RM(m, r)$ is a linear code, $\forall r, m$. [Why?]

(i) The blocklength of $RM(m, r)$, for any $r \in [0: m]$, is $n = 2^m$.

(ii) $\dim(RM(m, r)) = \sum_{i=0}^r \binom{m}{i}$, since the # monomials of degree i is exactly $\binom{m}{i}$ [Why? How does this give rise to the dimension?]

(iii) $d_{\min}(RM(m, r)) = 2^{m-r}$.

We now prove Property (iii) above. Observe that any polynomial $f(x_1, \dots, x_m) \in \mathbb{F}_2[x_1, \dots, x_m]$

can be decomposed as:

$$f(x_1, \dots, x_m) = g(x_1, \dots, x_{m-1}) + x_m \cdot h(x_1, \dots, x_{m-1}).$$

Thus, any codeword $\underline{c} = \text{Eval}(f)$ (ordered in the lexicographic order)

is such that $\underline{c} = (\underline{c}_1 \mid \underline{c}_2)$, where

$$\underline{c}_1 = \text{Eval}(g) \in RM(m-1, r)$$

and $\underline{c}_2 = \text{Eval}(g) + \text{Eval}(h)$, with $\text{Eval}(h) \in RM(m-1, r-1)$.

This observation is called the "Plotkin decomposition of RM codes".

Lemma 2: We have that $\underline{c} \in \text{RM}(m, r)$ is such that $\underline{c} = (\underline{u} | \underline{u} + \underline{v})$,
where $\underline{u} \in \text{RM}(m-1, r)$ and $\underline{v} \in \text{RM}(m-1, r-1)$.

Proof of Property (iii): We prove this by induction on m and r . The base case ($m=0$ and $r=0$) is easy to establish. Now, assume that the property holds for $\bar{m} = m-1$ and all $r \leq m-1$. We need to show that it holds at $\bar{m} = m$ and at order r .

In particular, we need to show that when $\underline{u} \in \text{RM}(m-1, r)$ and $\underline{v} \in \text{RM}(m-1, r-1)$, we have $w_H(\underline{u}) + w_H(\underline{u} + \underline{v}) \geq 2^{m-r}$, when \underline{u} & \underline{v} are not both all-zeros.

(a) Suppose that $\underline{v} = \underline{0}$: Clearly, then $w_H(\underline{u}) + w_H(\underline{u} + \underline{v})$
 $= 2w_H(\underline{u}) \geq 2 \cdot 2^{m-1-r} = 2^{m-r}$.

(b) Suppose that $\underline{v} \neq \underline{0}$: Then, by the Δ -ineq., we have
 $w_H(\underline{u}) + w_H(\underline{u} + \underline{v}) \geq w_H(\underline{v}) \geq 2^{m-1-(r-1)} = 2^{m-r}$.

Further, there exists a codeword $\underline{c} = \text{Eval}\left(\prod_{i=1}^r x_i\right)$ that has Hamming weight exactly 2^{m-r} . \square

Lecture: Decoding RM Codes and List Decoding

Last time, we discussed the RM family of codes and its basic properties.

We shall now take a look at a decoder for RM codes that can correct up to

$\lfloor \frac{d_{\min}-1}{2} \rfloor$ errors - Reed's decoder.

We first make some observations about the structure of RM codewords. For a

given subset $S \subseteq [m]$, let \bar{S} denote $[m] \setminus S$ and let

$$V_S \triangleq \{ \underline{z} \in \{0,1\}^m : z_i = 0, \forall i \in \bar{S} \}$$

be the vector space [why?] that consists of points (or coordinates) \underline{z} that are 0 outside S . Clearly, $\dim(V_S) = |S|$ and there are hence $2^{m-|S|}$ cosets $(V_S + \underline{b})$ of V_S .

Now, note that for any $S \subseteq [m]$ and any $\underline{b} \in \{0,1\}^m$, we always have

$$\sum_{\underline{z} \in V_S + \underline{b}} \text{Eval}_{\underline{z}}(x_S) = 1 \quad \text{--- (1)}$$

To see why (1) is true, note that $\text{Eval}_{\underline{z}}(x_S) = 1$ iff $z_i = 1, \forall i \in S$, which happens at exactly one point $\underline{z} \in V_S + \underline{b}$.

Furthermore, we have for all $T \neq S$,

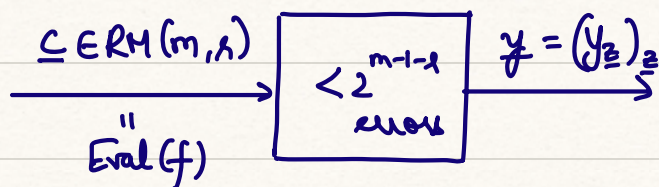
$$\sum_{\underline{z} \in V_S + \underline{b}} \text{Eval}_{\underline{z}}(x_T) = 0 \quad \text{--- (2)}$$

To see (2), note that since $T \neq S$, there exists $i \in S \setminus T$, with $z_i = 0$ or $z_i = 1$ not affecting $\text{Eval}_z(x_T)$.

Let $f(x_1, \dots, x_m) = \sum_{S \subseteq [m]: |S| \leq r} a_S x_S$ be s.t. $\text{Eval}(f) \in \text{RM}(m, r)$.

Reed's algorithm uses properties (1) & (2) to first decode a_S , for S s.t. $|S| = r$, and then $a_{S'}$, for $|S'| = r-1$, and so on.

Let $y = (y_z : z \in \{0, 1\}^m)$ be the received sequence when $\underline{c} = \text{Eval}(f)$ is corrupted by $< \frac{d_{\min}}{2} = 2^{m-1-r}$ errors.



Claim: We have that $\text{MAJ}_{\underline{b}} \left(\sum_{z \in V_S + \underline{b}} y_z \right) = a_S$, for $|S| = r$.

In other words, taking a majority vote among $\left(\sum_{z \in V_S + \underline{b}} y_z : \underline{b} \in \{0, 1\}^m \right)$ yields a_S .

Proof: Suppose first that $y = \text{Eval}(f)$. In this case, we have that for any $\underline{b} \in \{0, 1\}^m$,

$$\sum_{z \in V_S + \underline{b}} y_z = \sum_{z \in V_S + \underline{b}} \text{Eval}_z(f) = \sum_{T: |T| \leq r} a_T \sum_{z \in V_S + \underline{b}} \text{Eval}_z(x_T)$$

Note that from (1) and (2), we have $\sum_{z \in V_s + \underline{b}} \text{Eval}_z(x_T) = 1$ iff $S = T$.

$$\text{Hence, } \sum_{z \in V_s + \underline{b}} y_z = a_s, \quad \forall \underline{b} \in \{0, 1\}^m.$$

Now, if there are at most $2^{m-\lambda-1}$ errors in y , then at most $2^{m-\lambda-1}$ cosets $V_s + \underline{b}$ will be such that

$$\sum_{z \in V_s + \underline{b}} y_z \neq a_s. \quad [\text{why?}]$$

Hence, taking a majority vote $\text{MAJ}_{\underline{b}} \left(\sum_{z \in V_s + \underline{b}} y_z \right)$ yields a_s . \square

Now, after decoding a_s , for all S s.t. $|S| = \lambda$, Reed's algorithm obtains

$$\tilde{y} \leftarrow y - \text{Eval} \left(\sum_{S: |S|=\lambda} a_S x_S \right),$$

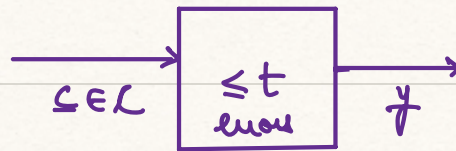
yielding a new "noisy" sequence \tilde{y} that can be seen as a corrupted version of a codeword $\underline{c}' \in \text{RM}(m, \lambda-1)$. The algorithm then applies the same procedure as outlined in Claim 1 above to decode a_s , for S s.t. $|S| = \lambda-1$, and so on.

Remark: Each step of majority voting is carried out on at most $2^{m-\lambda} \leq n$ cosets, and there are at most $\binom{m}{\lambda, m-\lambda} \leq 2^m = n$ monomials

of any fixed degree. Thus, Reed's algorithm works in $\text{poly}(n)$ time.

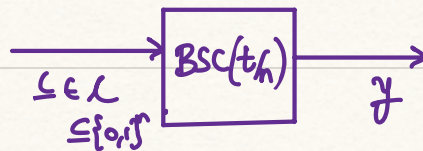
Further addenda:

We studied the family of RS codes, which we learnt were good since they achieve the Singleton bound [i.e., are Maximum Distance Separable (MDS) codes]. In other words, RS codes have the highest error-correcting capability, for errors introduced in an **adversarial manner**, i.e., when there is a bound $t \leq \lfloor \frac{n-k}{2} \rfloor$ on the # of errors a channel can introduce.



"Adversarial"/"Hamming" model of errors

In contrast, we have the stochastic/information-theoretic/**Shannon** model for errors:



Note that, "on average", one expects to see $\approx t$ errors in y [why?]. RM codes, on the other hand, are provably good for the above stochastic noise channel, i.e., RM codes achieve the capacity [Reeves-Pfister (2023), Abbe-Sandon (2023)] of the channel above, when $\lim_{n \rightarrow \infty} \frac{t}{n} = \tau \in (0, 1)$.

Note however that while the relative distance δ of RS codes of positive rate $R < 1$ is positive (if $R = \lim_n \frac{k}{n} \in (0, 1)$, then $\delta = \lim_n \frac{d}{n} = \lim_n \frac{n-k+1}{n} > 0$), the relative distance of RM codes of positive rate is asymptotically zero!

To see why, one must know that the rate

$$R(\text{RM}(m, r)) = \binom{m}{\leq r} / 2^m,$$

$$\text{with } \lim_{m \rightarrow \infty} R(\text{RM}(m, r_m)) > 0 \text{ iff } r_m \approx \frac{m}{2} \pm o(m).$$

$$\text{For such orders } (r_m)_{m \geq 1}, \text{ we have } \delta = \lim_{m \rightarrow \infty} \frac{2^{m-r_m}}{2^m} = 0.$$

Hence, it's NOT necessary for a code to have good performance over adversarial noise channels, for it to perform well over stochastic noise channels.

List decoding: An introduction

As seen above, there appears to be a "gap" between the fundamental limits of the adversarial and stochastic noise channels. More precisely, we make the following observation:

$$(i) \text{ [Adversarial model]} \quad \overbrace{\lim_n \frac{\# \text{ errors tolerated}}{n}}^{\triangleq \tau} \leq \frac{1-R}{2}. \text{ [Why?]}$$

$$\equiv R \leq 1 - 2\tau$$

(ii) [Shannon model] $R \leq 1 - h_b(\tau)$ [Recall Shannon's Theorem!]
 $\approx 1 - \tau$, for small τ .

The goal of list decoding is to serve as a bridge between the above fundamental limits, via the introduction of a new decoding paradigm, where the decoder is allowed to output a (small) list of candidate codewords. The decoder makes an error if the true input codeword is not present in this list.