

The Mattson-Solomon Polynomials

• $a = (a_0, a_1, \dots, a_{n-1}) \sim a(x) = a_0 + a_1x + \dots + a_{n-1}x^{n-1}$

• MS poly. associated with $a \rightarrow A(z)$

$a = (a_0, a_1, \dots, a_{n-1}), a_i \in GF(q^m), A(z) \in GF(q^m)[z]$

$A(z) = \sum_{j=1}^n A_j z^{n-j}$

$\alpha \in GF(q^m)$ is a primitive n -th root of unity.

where $A_j = a(\alpha^j) = \sum_{i=0}^{n-1} a_i \alpha^{ij}, j = 0, \pm 1, \pm 2, \dots$

($A(z)$ is not taken modulo $z^n - 1$)

• Alternative forms of $A(z)$ are:

$A(z) = \sum_{j=0}^{n-1} A_{-j} z^j$

$n-j = j'$

$= \sum_{i=0}^{n-1} a_i \sum_{j=0}^{n-1} (\alpha^{-i} z)^j$

• $A_j = A_{j \bmod n}$
• $A(z)$ sometimes called discrete Fourier Transform of a .

Remarks a) The coefficients A_j are given by

$$\begin{bmatrix} A_0 \\ A_1 \\ \vdots \\ A_{n-1} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ \alpha & \alpha & \alpha^2 & \dots & \alpha^{n-1} \\ 1 & \alpha^2 & \alpha^4 & \dots & \alpha^{2(n-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \alpha^{n-1} & \alpha^{2(n-1)} & \dots & \alpha^{(n-1)^2} \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{bmatrix}$$

$\delta = 2t+1$
or
 $\delta = 2t$

b) if $a_i \in GF(q)$, then $(A_j)^r = A_{jr \bmod n}$ (subscript mod n)

c) A narrow-sense BCH code of designated distance δ can now be defined as all vectors a for which

$A_1 = A_2 = \dots = A_{\delta-1} = 0$
 $\rightarrow g(x) = \text{lcm.} \{ M^{(1)}(x), M^{(2)}(x), \dots, M^{(\delta-1)}(x) \}$
 $(a(\alpha^1) = 0 = a(\alpha^2) = \dots = a(\alpha^{\delta-1}))$

Theorem (Inversion formula)

$$A(z) = \sum_{j=0}^{n-1} a_j \sum_{i=0}^{n-1} (\alpha^{-i} z)^j$$

The vector a is recovered from $A(z)$ by

$$a_i = \frac{1}{n} A(\alpha^i), i=0, 1, \dots, n-1$$

$$a = \frac{1}{n} (A(1), A(\alpha), \dots, A(\alpha^{n-1}))$$

$$a(x) = \frac{1}{n} \sum_{i=0}^{n-1} A(\alpha^i) x^i$$

if $C(x) = \sum_{i=0}^{n-1} C_i x^i$
 then $C_i = \frac{1}{n} \sum_{j=0}^{n-1} C(\alpha^j) \alpha^{-ji}$
 where α is a zero of $x^n - 1$

$$A(z) = \sum_{j=0}^{n-1} A_j z^j$$

$$A_j = \frac{1}{n} \sum_{i=0}^{n-1} A(\alpha^i) \alpha^{-ji}$$

$$A_j = a(\alpha^j) = \sum_{i=0}^{n-1} a_i (\alpha^j)^i$$

$$\therefore a_i = \frac{1}{n} \sum$$

Notation

$$f(y) \bmod y^n - 1$$

$$\rightarrow [f(y)]_n$$

(remainder)

$$f(y) = \sum_{i=0}^{n-1} f_i y^i, g(y) = \sum_{i=0}^{n-1} g_i y^i$$

$$f(y) * g(y) = \sum_{i=0}^{n-1} f_i g_i y^i$$

(Componentwise product)

Theorem (properties of MS polys). (omit proof)

i) if $c(x) = a(x) + b(x)$ then $C(z) = A(z) + B(z)$

$$C(z) \rightarrow \text{MS poly for } c(x)$$

$$A(z), B(z) \rightarrow \text{MS poly for } a(x), b(x) \text{ resp.}$$

ii) $c(x) = [a(x)b(x)]_n$ iff $C(z) = A(z) * B(z)$

iii) $c(x) = a(x) * b(x)$ iff $C(z) = \frac{1}{n} [A(z)B(z)]_n$

iv) The MS poly of a cyclic shift of a is $A(\alpha z)$

v) The MS poly of $0 \rightarrow 0, 1 \rightarrow 1$

vi) An overall parity check on a is given by $\sum_{i=0}^{n-1} a_i = A(0) = A_0$

• $A(z) \rightarrow$ MS poly $a = \sum_{i=0}^{n-1} a_i x^i \sim (a_0, a_1, \dots, a_{n-1})$

$$A(z) = \sum_{i=1}^n A_i z^{n-i}, \quad A_i = a(\alpha^i)$$

$$= \sum_{i=0}^{n-1} A_{-i} z^i$$

subscript mod n

• inversion formula

$$a_j = \frac{1}{n} A(\alpha^j), \quad j=0, 1, \dots, n-1.$$

proof

$$A(\alpha^j) = \sum_{i=1}^n a(\alpha^i) \alpha^{(n-i)j}$$

$$= \sum_{i=1}^n \sum_{k=0}^{n-1} a_k \alpha^{ik} \cdot \alpha^{(n-i)j}$$

$$= \sum_{k=0}^{n-1} a_k \sum_{i=1}^n \alpha^{ik + (n-i)j}$$

$k=j \rightarrow a_j \sum \alpha$

$\alpha \rightarrow n$ -th root of unity

$$\sum_{i=0}^{n-1} \alpha^i = \begin{cases} 0 & \text{if } d \neq 1 \\ n & \text{if } d=1 \end{cases}$$

$$A(\alpha^j) = \sum_{i=0}^{n-1} A_{-i} \alpha^{ji}$$

$$= \sum_{i=0}^{n-1} a(\alpha^{-i}) \alpha^{ji}$$

$$= \sum_{i=0}^{n-1} \sum_{k=0}^{n-1} a_k \alpha^{-ik} \cdot \alpha^{ji}$$

$$= \sum_{k=0}^{n-1} a_k \sum_{i=0}^{n-1} \alpha^{(j-k)i}$$

$$= a_j \sum_{i=0}^{n-1} \alpha^0 + \sum_{\substack{k=0 \\ k \neq j}}^{n-1} a_k \left(\sum_{i=0}^{n-1} \alpha^{(j-k)i} \right)$$

$$= na_j$$

$$\frac{1 - (\alpha^{j-k})^n}{1 - \alpha^{j-k}} = 0 \quad \text{as } \alpha^n = 1$$

$$\sum_{i=0}^{n-1} (\alpha^{j-k})^i$$

proof of (ii) X

if $C(x) = [a(x) b(x)]_n$ then $C_j = C(\alpha^j) = a(\alpha^j) b(\alpha^j) = A_j B_j$

$\therefore C(z) = A(z) * B(z)$
 $= \sum_{j=0}^{n-1} A_j B_j z^j$

Conversely, let $C_j = A_j B_j \forall j$ s.t. $C(z) = A(z) * B(z)$

Claim $\sum_{k=0}^{n-1} C_k x^k = \left(\sum_{i=0}^{n-1} a_i x^i \right) \left(\sum_{I=0}^{n-1} b_I x^I \right)$

RHS = $\left(\sum_{i=0}^{n-1} \left\{ \sum_{j=0}^{n-1} a(\alpha^j) \bar{\alpha}^{ij} \right\} x^i \right) \times \left(\sum_{I=0}^{n-1} \left\{ \sum_{J=0}^{n-1} b(\alpha^J) \bar{\alpha}^{IJ} \right\} x^I \right)$
 mod $x^n - 1$

$a(x) = \sum_{i=0}^{n-1} a_i x^i$
 $\Rightarrow a_i = \frac{1}{n} \sum_{j=0}^{n-1} a(\alpha^j) \bar{\alpha}^{ij}$
 where α is the n -th root of unity

Coefficient of x^k is

i.e. $\alpha \in GF(q^m)$ is a zero of $x^n - 1$

$\sum_{i=0}^{n-1} \left(\frac{1}{n} \sum_{j=0}^{n-1} a(\alpha^j) \bar{\alpha}^{ij} \right) \left(\sum_{J=0}^{n-1} b(\alpha^J) \bar{\alpha}^{(n+k-j)J} \right)$
 $= \frac{1}{n} \sum_{j=0}^{n-1} a(\alpha^j) \sum_{J=0}^{n-1} b(\alpha^J) \bar{\alpha}^{-(n+k)J}$
 $\times \sum_{i=0}^{n-1} \alpha^{i(J-j)}$

$\sum_{i=0}^{n-1} \alpha^{i(J-j)} = \begin{cases} n & \text{if } J=j \\ 0 & \text{otherwise} \end{cases}$
 $\sum_{i=0}^{n-1} \alpha^{i(J-j)} = \sum_{i=0}^{n-1} \alpha^{i(J-j)} = \sum_{i=0}^{n-1} \alpha^{i(J-j)} = \sum_{i=0}^{n-1} \alpha^{i(J-j)}$
 $\sum_{i=0}^{n-1} \alpha^{i(J-j)} = \sum_{i=0}^{n-1} \alpha^{i(J-j)} = \sum_{i=0}^{n-1} \alpha^{i(J-j)}$

$= \frac{1}{n} \sum_{j=0}^{n-1} a(\alpha^j) b(\alpha^j) \bar{\alpha}^{-(n+k)j}$
 $= \frac{1}{n} \sum_{j=0}^{n-1} a(\alpha^j) b(\alpha^j) \bar{\alpha}^{-kj} = \frac{1}{n} \sum_{j=0}^{n-1} A_j B_j \bar{\alpha}^{-kj}$
 $= \frac{1}{n} \sum_{j=0}^{n-1} C_j \bar{\alpha}^{-kj} = \frac{1}{n} \sum_{j=0}^{n-1} C_j \alpha^{-kj}$
 $\sum_{i=0}^{n-1} \alpha^i = \begin{cases} 0 & \text{if } \alpha \neq 1 \\ n & \text{if } \alpha = 1 \end{cases}$

locator polynomial

- $a = (a_0, a_1, \dots, a_{n-1})$, $a_i \in F$ has non-zero components $\nearrow = GF(q)$
 $a_{i_1}, a_{i_2}, \dots, a_{i_w}$ & no others.

$\therefore \text{wt}(a) = w$

- We associate with a the following elements of $GF(q^m)$:

$$X_1 = \alpha^{a_{i_1}}, X_2 = \alpha^{a_{i_2}}, \dots, X_w = \alpha^{a_{i_w}}$$

called the locators of a and following elements of $GF(q)$

$$Y_1 = a_{i_1}, Y_2 = a_{i_2}, \dots, Y_w = a_{i_w},$$

giving the values of the non-zero components.

- Thus a is completely specified by $(X_1, Y_1), (X_2, Y_2), \dots, (X_w, Y_w)$.

- In case a is a binary vector, all Y_i 's are 1.

Note that $a(\alpha^j) = A_j = \sum_{i=1}^w Y_i X_i^j$

$$a_0 + a_1 \alpha^j + a_2 \alpha^{2j} + \dots + a_{n-1} \alpha^{(n-1)j}$$

$$a_{i_1} (\alpha^j)^{i_1} + a_{i_2} (\alpha^j)^{i_2} + \dots + a_{i_w} (\alpha^j)^{i_w}$$

$$Y_1 X_1^j + Y_2 X_2^j + \dots + Y_w X_w^j$$

Defⁿ

The locator polynomial of the vector a is

$$\Delta(z) = \prod_{i=1}^w (1 - x_i z) = \sum_{i=0}^w \delta_i z^i, \quad \delta_0 = 1$$

- The roots of $\Delta(z)$ are the reciprocals of the locators
- The coefficients δ_i are the elementary symmetric functions of the x_i :

$$\delta_1 = -(x_1 + x_2 + \dots + x_w)$$

$$\delta_2 = x_1 x_2 + x_1 x_3 + \dots + x_{w-1} x_w$$

$$\delta_w = (-1)^w x_1 \dots x_w.$$

Generalized Newton identities

The A_i 's and the δ_i 's are related by a set of simultaneous linear equations.

Theorem $\forall j$, the A_i 's satisfy the linear recurrence

$$\textcircled{1} - \boxed{A_{j+w} + \delta_1 A_{j+w-1} + \dots + \delta_w A_j = 0.}$$

In particular, taking $j = 1, 2, \dots, w$

$$\begin{pmatrix} A_w & A_{w-1} & \dots & A_1 \\ A_{w+1} & A_w & \dots & A_2 \\ \vdots & \vdots & \ddots & \vdots \\ A_{2w-1} & A_{2w-2} & \dots & A_w \end{pmatrix} \begin{pmatrix} \delta_1 \\ \delta_2 \\ \vdots \\ \delta_w \end{pmatrix} = - \begin{pmatrix} A_{w+1} \\ A_{w+2} \\ \vdots \\ A_{2w} \end{pmatrix}$$

Proof of (i)

$$\Delta(z) = \prod_{i=1}^w (1 - x_i z) = \sigma_0 + \sigma_1 z + \sigma_2 z^2 + \dots + \sigma_w z^w$$

put $z = \frac{1}{x_i}$

$\times Y_i$ $0 = Y_i \sigma_0 + \frac{\sigma_1 Y_i}{x_i} + \frac{\sigma_2 Y_i}{x_i^2} + \dots + \frac{\sigma_w Y_i}{x_i^w} \times \frac{Y_i x_i^{j+w}}{\sum_{i=1}^w}$

$\times \frac{x_i^{j+w}}{\sum}$ $0 = \sum_{i=1}^w Y_i x_i^{j+w} \sigma_0 + \sum_{i=1}^w Y_i x_i^{j+w-1} \sigma_1 + \sum_{i=1}^w Y_i x_i^{j+w-2} \sigma_2 + \dots + \sum_{i=1}^w Y_i x_i^j \sigma_w$

$$= \sigma_0 A_{j+w} + \sigma_1 A_{j+w-1} + \sigma_2 A_{j+w-2} + \dots + \sigma_w A_j \quad \left| \begin{array}{l} \sum_{i=1}^w Y_i x_i^j = A_j \\ \sigma_0 = 1 \end{array} \right.$$

Exercise Let a be a vector of weight w .

Show that

$$\begin{pmatrix} A_{0v} & A_{v-1} & \dots & A_1 \\ A_{v+1} & A_v & \dots & A_2 \\ \dots & \dots & \dots & \dots \\ A_{2v-1} & A_{2v-2} & \dots & A_v \end{pmatrix}_{v \times v}$$

is nonsingular if $\underline{v=w}$, but singular if $v > w$.
 $w=v$ $w < v$

The usual form of Newton's identities

let $x_1, x_2, \dots, x_w$ be indeterminates, and

$$\Delta(z) = \prod_{i=1}^w (1 - x_i z) = \sum_{i=0}^w \sigma_i z^i$$

When σ_i 's are elementary symm. funs. of x_i 's, $\sigma_0 = 1$ & $\sigma_i = 0$ for $i > w$

Define the power sum

$$P_i = \sum_{r=1}^w x_r^i \quad \forall i.$$

~~Defn~~ (a) if $P(z) = \sum_{i=1}^{\infty} P_i z^i$, Show that—

$$\Delta(z)P(z) + z\Delta'(z) = 0.$$

(b) By equating coefficients, Show that

$$P_1 + \sigma_1 = 0$$

$$P_2 + \sigma_1 P_1 + 2\sigma_2 = 0$$

$$P_w + \sigma_1 P_{w-1} + \dots + \sigma_{w-1} P_1 + w\sigma_w = 0.$$

and, for $i > w$

$P_i + \sigma_1 P_{i-1} + \dots + \sigma_w P_{i-w} = 0$

 — (2)

② agrees with ① when all $y_i = 1$ (i.e. binomial coeff)

Exercis

In the binary case, the above equations imply

$$\begin{pmatrix}
 1 & 0 & 0 & 0 & 0 & \dots & 0 \\
 A_2 & A_1 & 1 & 0 & 0 & \dots & 0 \\
 A_4 & A_3 & A_2 & A_1 & 1 & \dots & 0 \\
 \dots & \dots & \dots & \dots & \dots & \dots & \dots \\
 A_{2w-2} & A_{2w-3} & \dots & \dots & A_{w-1} & \dots & 0
 \end{pmatrix}
 \begin{pmatrix}
 \sigma_1 \\
 \sigma_2 \\
 \vdots \\
 \sigma_w
 \end{pmatrix}
 =
 \begin{pmatrix}
 A_1 \\
 A_3 \\
 A_5 \\
 \vdots \\
 A_{2w-1}
 \end{pmatrix}$$

$\nwarrow$ MSB

Exercis MSB non-singular if $v \leq w$
 $w = v$ or $v-1$

MSB singular if $w < v-1$

→ (3)

Applications to decoding BCH codes

- BCH code with designated distance δ .
- $a = (a_0, a_1, \dots, a_{n-1}) \rightarrow$ error vector
- decoder can easily calculate MSB
 $A_1, A_2, \dots, A_{\delta-1}$ & finds $\sigma(z)$ using
 Newton identities. (hard)

$$y_1 = a_{i_1}, \quad y_2 = a_{i_2}, \quad \dots, \quad y_w = a_{i_w}$$

→ errors values

~~$$x_1 = \alpha^{a_{i_1}}, \quad x_2 = \alpha^{a_{i_2}}, \quad \dots, \quad x_w = \alpha^{a_{i_w}}$$~~

$$x_1 = \alpha^{i_1}, \quad x_2 = \alpha^{i_2}, \quad \dots, \quad x_w = \alpha^{i_w}$$

→ error locator
 $i_1, i_2, \dots, i_w \rightarrow$ positions where errors occur.

Then $A_j = \sum_{i=1}^w x_i y_i^j$, $\sigma(z) = \prod (1 - x_i z)$

$\nwarrow$ error locator poly. $\rightarrow$ find zeros

To determine the y_i 's, define the evaluator polynomial

$$\omega(z) = \sigma(z) + \sum_{k=1}^{\omega} z X_k Y_k \prod_{\substack{j=1 \\ j \neq k}}^{\omega} (1 - X_j z)$$

$$\omega(X_i^{-1}) = \underbrace{\sigma(X_i^{-1})}_0 + \underbrace{\sum_{\substack{k=1 \\ j \neq i}}^{\omega} X_i^{-1} X_k Y_k \prod_{\substack{j=1 \\ j \neq k}}^{\omega} (1 - X_j X_i^{-1})}_{\substack{j=1 \\ j \neq k} \rightarrow 1 - X_i X_i^{-1} = 0} \rightarrow 0$$

Once $\omega(z)$ is known, y_i is given by

$$y_i = \omega(X_i^{-1}) / \prod_{j \neq i} (1 - X_j X_i^{-1})$$

$$\sigma(z) = \prod_{j=1}^{\omega} (1 - X_j z)$$

$$\sigma'(X_i^{-1}) = \prod_{\substack{j=1 \\ j \neq i}}^{\omega} (1 - X_j X_i^{-1}) (-X_i)$$

$$= -X_i \omega(X_i^{-1}) / \sigma'(X_i^{-1})$$

Theorem $\omega(z) = (1 + S(z)) \sigma(z) \quad \text{--- (3)}$

Where $S(z) = \sum_{i=1}^{\omega} A_i z^i$

Note that $\deg \omega(z) \leq \deg \sigma(z) = \omega$, only $A_1, A_2, \dots, A_{\omega}$ are needed to determine $\omega(z)$ from (3).

Proof

$$\frac{\omega(z)}{\sigma(z)} = 1 + \sum_{i=1}^{\omega} \frac{z X_i Y_i}{1 - X_i z}$$

$$= 1 + \sum_{i=1}^{\omega} Y_i \sum_{j=1}^{\omega} (z X_i)^j$$

$$= 1 + \sum_{j=1}^{\omega} A_j z^j = 1 + S(z).$$

$$A_i = \sum_{r=1}^{\omega} X_r Y_r^i$$

Theorem The weight of a is $n-r$, where r is the no. of n th roots of unity which are zeros of the MS poly. $A(z)$.

Proof $a = \frac{1}{n} (A(1), A(\alpha), \dots, A(\alpha^{n-1}))$.

$\text{wt}(a) = n-r$ means $A(\alpha^i) = 0$

for $i = i_1, i_2, \dots, i_r$
 $\in \{0, 1, 2, \dots, n-1\}$

$\text{wt}(a) = n-r \geq n - \deg A(z)$ i.e. $A(z)$ has r many n th roots of unity as its zero.

$\Rightarrow \deg A(z) \geq r$
 $-r \geq -\deg A(z)$

Corollary

If a has MS poly $A(z)$, then

$\text{wt}(a) \geq n - \deg A(z)$.

* This gives another proof of the BCH bound.

* Theorem (BCH bound)

Let C be a cyclic code with generator poly

$g(x) = \prod_{i \in K} (x - \alpha^i)$, where K contains

a string of $d-1$ ~~successive~~ ^{consecutive} integers

$b, b+1, \dots, b+d-2$, for some b .

Then minimum weight of any non-zero codeword $a \in C$ is at least d .

Proof: Since $a \in \mathbb{C}$, $g(x) \mid a(x)$

$$\therefore a(\alpha^j) = 0, \text{ for } b \leq j \leq b+d-2$$

$$\therefore A(z) = \sum_{j=1}^n A_j z^{n-j} \quad A_j = a(\alpha^j)$$

$$= a(\alpha) z^{n-1} + a(\alpha^2) z^{n-2} + \dots + a(\alpha^{b+d-1}) z^{n-b-d+1} \\ = \underbrace{a(\alpha)}_{A_1''} z^{n-1} + \dots + \underbrace{a(\alpha^{b-1})}_{A_{b-1}''} z^{n-b+1} + \underbrace{a(\alpha^{b+d-1})}_{A_{b+d-1}''} z^{n-b-d+1} + \dots + a(\alpha^n)$$

$$\text{Let } \hat{A}(z) = z^{b-1} A(z) - \left\{ a(\alpha) z^{b-2} + \dots + a(\alpha^{b-1}) \right\} \{z^{n-1}\}$$

$$= \cancel{a(\alpha) z^{n+b-2}} + \cancel{a(\alpha^2) z^{n+b-1}} + \dots + \cancel{a(\alpha^{b-1}) z^n} \\ + a(\alpha^{b+d-1}) z^{n-d} + \dots + a(\alpha^n) z^{b-1}$$

$$- \left\{ \cancel{a(\alpha) z^{n+b-2}} + \dots + \cancel{a(\alpha^{b-1}) z^n} \right\} \\ + a(\alpha) z^{b-2} + \dots + a(\alpha^{b-1})$$

$$= \underbrace{a(\alpha^{b+d-1})}_{A_{b+d-1}''} z^{n-d} + \dots + \underbrace{a(\alpha^n)}_{A_n''} z^{b-1} + \underbrace{a(\alpha)}_{A_1''} z^{b-2} + \dots + \underbrace{a(\alpha^{b-1})}_{A_{b-1}''}$$

of n -th roots of unity which are zeros of $A(z)$
 = # of n -th roots of unity which are zeros of $\hat{A}(z)$
 $\leq \deg \hat{A}(z) = n-d$
 $a \rightarrow$ MS poly $A(z)$, $\deg A(z) \leq n-d$
 $\Rightarrow \text{wt}(a) \geq d$

$$\left. \begin{array}{l} \deg A(z) = r \\ \Rightarrow \text{wt}(a) = n-r \end{array} \right\}$$

Decoding of BCH code (binary)

1. Calculate syndrome

$$S = Hy^T = \begin{pmatrix} 1 & \alpha & \alpha^2 & \dots & \alpha^{n-1} \\ 1 & \alpha^3 & \alpha^6 & \dots & \alpha^{3(n-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \alpha^{b-2} & \alpha^{2(b-2)} & \dots & \alpha^{(n-1)(b-2)} \end{pmatrix} \begin{pmatrix} y_0 \\ y_1 \\ \vdots \\ y_{n-1} \end{pmatrix}$$

Received vector

$$\underline{y} = (y_0, y_1, \dots, y_{n-1}) = \begin{pmatrix} y(\alpha) \\ y(\alpha^3) \\ \vdots \\ y(\alpha^{b-2}) \end{pmatrix} = \begin{pmatrix} A_1 \\ A_3 \\ \vdots \\ A_{b-2} \end{pmatrix}$$

$$y(x) = y_0 + y_1x + \dots + y_{n-1}x^{n-1}$$

where $A_i = y(\alpha^i)$

Note that $A_{2r} = y(\alpha^{2r}) = y(\alpha^r)^2 = A_r^2$

Alternatively, calculate A_i 's as follows:

$M^{(t)}(x) \rightarrow$ minimal poly. of α^t .

$$y(x) = Q(x) \underset{\substack{\uparrow \\ \alpha^t}}{M^{(t)}(x)} + R(x), \quad 0 \leq \deg R(x) < \deg M^{(t)}(x)$$

Then $A_t = y(\alpha^t) = R(\alpha^t)$

2. Find the error locator poly $\sigma(z)$

$e \rightarrow$ error vector, $\text{wt}(e) = w$

$$e = 0 \dots 0 e_{i_1} 0 \dots e_{i_2} \dots e_{i_w} 0 \dots 0$$

wt $\otimes$ $X_1 = \alpha^{i_1}, X_2 = \alpha^{i_2}, \dots, X_w = \alpha^{i_w}$

$$\sigma(z) = \prod_{i=1}^{\omega} (1 - x_i z) = \sum_{i=0}^{\omega} \sigma_i z^i \quad x_r = \alpha^{ir}$$

↓
much harder part

$$e(x) = e_1' x^{i_1} + e_2' x^{i_2} + \dots + e_w' x^{i_w}$$

$$y(x) = y_0 + y_1 x + \dots + y_{n-1} x^{n-1}$$

σ_i 's & A_i 's are
related by
Newton's identities

$$\begin{aligned} A_1 &= y(\alpha^1) = \cancel{y_0} + \cancel{y_1} \alpha^1 \\ &= c(\alpha^1) + e(\alpha^1) \\ &= e(\alpha^1) \end{aligned}$$

$$= \cancel{\sum x_i}$$

$$= e_1(\alpha^1)^{i_1} + e_2(\alpha^1)^{i_2} + \dots + e_w(\alpha^1)^{i_w}$$

$$= x_{\beta_1}^1 + x_{\beta_2}^1 + \dots + x_{\beta_w}^1$$

$$= \sum_{r=1}^{\omega} x_r^1$$

$\sigma(z)$ is not uniquely
determined.

the decoder must find
the vector e with least
weight ω , or the lowest
degree $\sigma(z)$ satisfying
Newton's identities.

This uncertainty in ω makes this stage
so difficult.

$$\begin{matrix} * \leftarrow \end{matrix} \begin{pmatrix} 1 & 0 & 0 & 0 & 0 & \dots & 0 \\ A_2 & A_1 & 1 & 0 & 0 & \dots & 0 \\ \dots & & & & & & \\ A_{2w-4} & A_{2w-5} & \dots & A_{w-3} & & & \\ A_{2w-2} & A_{2w-3} & \dots & A_{w-1} & & & \end{pmatrix} \begin{pmatrix} \sigma_1 \\ \sigma_2 \\ \vdots \\ \sigma_{w-1} \\ \sigma_w \end{pmatrix} = \begin{pmatrix} A_1 \\ A_3 \\ A_5 \\ \vdots \\ A_{2w-3} \\ A_{2w-1} \end{pmatrix}$$

Theorem let $A_l = \sum_{i=1}^{\omega} x_i^l$

$$|A_l = \sum_{i=1}^{\omega} \gamma_i x_i^l|$$

The $v \times v$ matrix

binary case
 $\gamma_i = 1 + i$

$$M_v = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & \dots & 0 \\ A_2 & A_1 & 1 & 0 & 0 & \dots & 0 \\ A_4 & A_3 & A_2 & A_1 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ A_{2v-4} & A_{2v-5} & \dots & \dots & \dots & \dots & A_{v-3} \\ A_{2v-2} & A_{2v-3} & \dots & \dots & \dots & \dots & A_{v-1} \end{bmatrix}$$

is non singular if $\omega = v$ or $\omega = v-1$, and
singular if $\omega < v-1$.

$$\det M_v = \det \bar{M}_{v-1}$$

$\downarrow$
 $v \times v$

X proof-

i) $\omega < v-1$

$$\text{Then } M_v \begin{bmatrix} 0 \\ \vdots \\ \delta_1 \\ \vdots \\ \delta_{v-2} \end{bmatrix} = 0$$

$\Rightarrow M_v$ is ~~non~~ singular

using Newton's
identity
(binary case)

ii) $\omega = v$

$$\text{Then } \det M_v = \prod_{i < j} (x_i + x_j)$$

? For if we put $x_i = x_j$, $\det M_v = 0$ by (i)

$$\therefore \det M_v = \text{const.} \prod_{i < j} (x_i + x_j)$$

& const. is easily found to be

$$A_1 \pm x_1 + x_2 + \dots + x_{\omega} = \begin{cases} 0 & \text{if } \omega \text{ even} \\ x_1 & \text{if } \omega \text{ odd} \end{cases}$$

iii) finally, if $\omega = v-1$, M_v is nonsingular from (ii).

Using this theorem we have the following iterative
algm. for finding $\sigma(z)$, for a BCH code
of designated distance $\delta = 2t+1$, assuming
 w errors occur where $w \leq t$.

• Assume t errors occurred & try to solve
eqn's (*) with w replaced by t .

• By the above theorem, if t or $t-1$ errors
have occurred, a $\sigma(z)$ exists & we go
to next stage.

(3) Finding the roots of $\sigma(z) = \prod_{i=1}^w (1 - X_i z)$
→ reciprocal of ~~$\sigma(z)$~~ are zeros of
 $\sigma(z)$ are $X_1 = \alpha^{a_{i_1}}, X_2 = \alpha^{a_{i_2}}, \dots, X_w = \alpha^{a_{i_w}}$
& errors occurred at co-ordinates
 $i_1, i_2, \dots, i_w$.

→ error values if $\deg \sigma(z) = 1$ or 2
the zeros can be found directly.

→ in general, simplest technique is just
to test each power of α in turn
to see if it is a zero of $\sigma(z)$.

(Chien search)

→ There is an error in co-ordinate i
iff $\sigma(\alpha^i) = 0$.

- But if fewer than $t-1$ errors occurred, the eqns. $(*)$ have no soln.

• In this case assume $t-2$ errors occurred, & again try to solve $(*)$ with w now replaced by $t-2$.

- Repeat until a soln. found.

Difficulty in finding $\sigma(z)$

→ requires repeated evaluation of a large determinant over $GF(2^m)$.

→ if t large (> 3 or 4), this method is not preferred.

→ instead, generalized Newton's identities are used along with Berlekamp algm.

Using the generalized Newton's identities - the Berlekamp algorithm:

Assuming that w errors occurred.

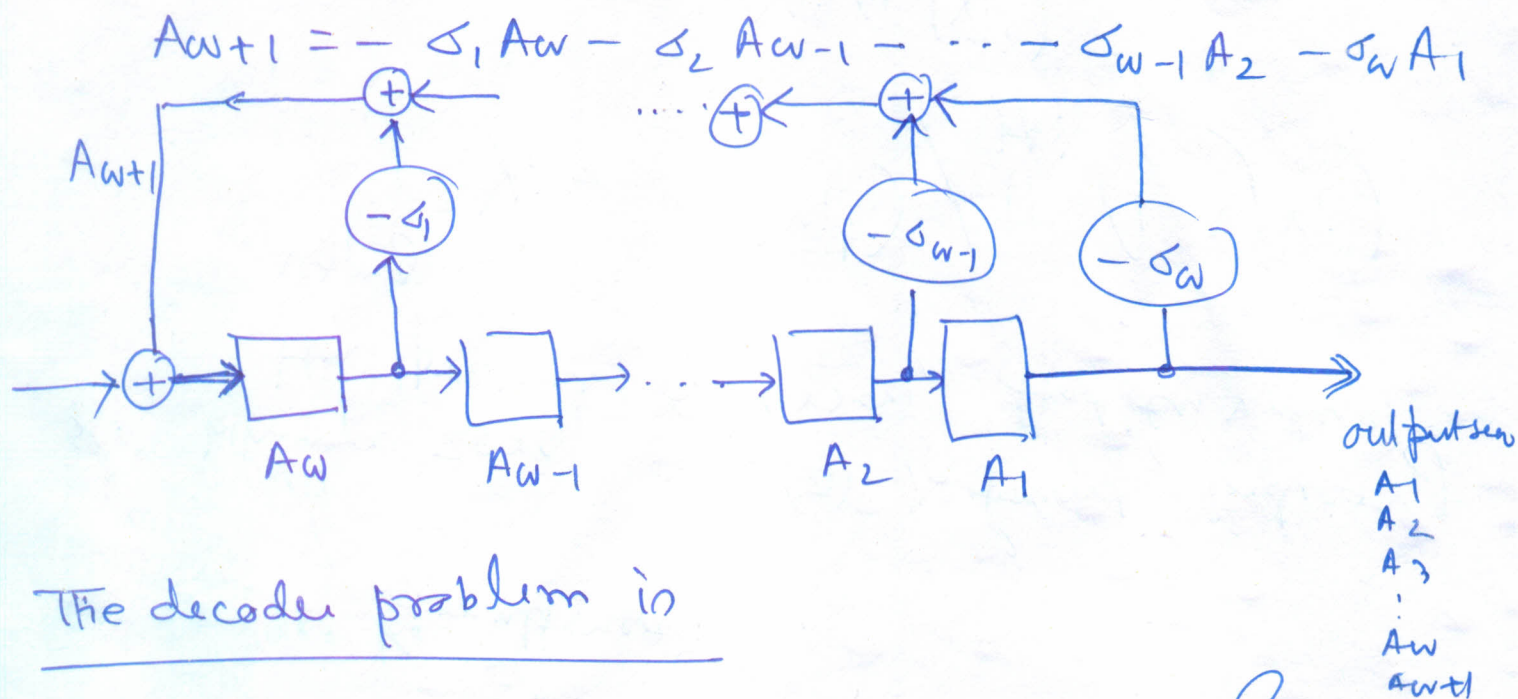
The σ_i 's & A_i 's are related by

$$A_{j+w} + \sigma_1 A_{j+w-1} + \dots + \sigma_w A_j = 0.$$

In particular, taking $j=1, 2, \dots, w$,

$$\begin{pmatrix} A_w & A_{w-1} & \dots & A_1 \\ A_{w+1} & A_w & \dots & A_2 \\ \dots & \dots & \dots & \dots \\ A_{2w-1} & A_{2w-2} & \dots & A_w \end{pmatrix} \begin{pmatrix} \sigma_1 \\ \sigma_2 \\ \vdots \\ \sigma_w \end{pmatrix} = - \begin{pmatrix} A_{w+1} \\ A_{w+2} \\ \vdots \\ A_{2w} \end{pmatrix}$$

- this can be interpreted as A_i 's are output from a LFSR of w stages, with initial contents $A_1, A_2, \dots, A_w$.



The decoder problem is

- Given the sequence $A_1, A_2, \dots, A_{s-1}$, find the LFSR of shortest length w which produces $A_1, \dots, A_{s-1}$ as output when initially loaded with $A_1, A_2, \dots, A_w$.
- There is an efficient algm. for finding such an LFSR (& hence error locator poly $\sigma(z)$) which is due to Berlekamp.

Decoding of non-binary BCH code

1. Decoder finds $A_1, \dots, A_{s-1}$.
2. The generalized Newton's identities are used to find $\sigma(z)$ & the error evaluator poly

$$e \quad \omega(z) = (1 + s(z)) / \sigma(z).$$

$$\text{where } \sigma(z) = \sum_{i=1}^s A_i z^i$$

Alternatively, Berlekamp's algm. can be used to efficiently find $\sigma(z)$ & $\omega(z)$ simultaneously.

3. When a zero of $\sigma(z)$ is found, indicating the presence of error, find the value of error using

$$y_i = \omega(\bar{x}_i^{-1}) / \prod_{j \neq i} (1 - x_j \bar{x}_i^{-1})$$

$$= - \frac{x_i \omega(\bar{x}_i^{-1})}{\sigma'(\bar{x}_i^{-1})}$$

Correcting more than t errors

The decoding algms. correct t or fewer errors in a BCH code of designated distance $2t+1$.

Complete decoding algms. are known for all double & some triple-error correcting codes.

Unsolved problem

Find a complete decoding algm. for all BCH codes.