

- Suppose we have filled out all the rows up to and including the μ -th row, we fill out the $(\mu + 1)$ -th row as follows:

- (1) If $d_\mu = 0$, then we set

$$\sigma^{(\mu+1)}(X) = \sigma^{(\mu)}(X).$$

- (2) If $d_\mu \neq 0$, we find a row prior to the μ -th row, say the ρ -th row, with partial solution $\sigma^{(\rho)}(X)$ such that $d_\rho > 0$ and ρ is maximal ~~$d_\rho \leq 1$ and ρ is the largest.~~ Then

$$\sigma^{(\mu+1)}(X) = \sigma^{(\mu)}(X) + d_\mu d_\rho^{-1} X^{2(\mu-\rho)} \sigma^{(\rho)}(X). \quad (5.37)$$

- (3) Compute the discrepancy

$$d_{\mu+1} = S_{2\mu+3} + \sigma_1^{(\mu+1)} S_{2\mu+2} + \sigma_2^{(\mu+1)} S_{2\mu+1} + \cdots + \sigma_{l_{\mu+1}}^{(\mu+1)} S_{2\mu+3-l_{\mu+1}}. \quad (5.38)$$

where $l_{\mu+1}$ is degree of $\sigma^{(\mu+1)}(X)$.

- If $d_\mu \neq 0$, then $\sigma^{(\mu)}(X)$ needs to be adjusted to obtain a new minimum-degree polynomial $\sigma^{(\mu+1)}(X)$ whose coefficients satisfy the first $\mu + 1$ Newton's identities.

- **Correction:** Go back to the steps prior the μ -th step and determine a step ρ at which the partial solution is $\sigma^{(\rho)}(X)$ such that ~~$d_\rho \neq 0$ and $\rho \leq l_\rho$~~ ^{$\rho_{p+1} > l_p$ and ρ is maximal,} ~~has the largest value,~~ where l_ρ is the degree of $\sigma^{(\rho)}(X)$. Then the solution at the $(\mu + 1)$ -th step of the iteration process for finding the error-location polynomial $\sigma(X)$ is

$$\sigma^{(\mu+1)}(X) = \sigma^{(\mu)}(X) + d_\mu d_\rho^{-1} X^{(\mu-\rho)} \sigma^{(\rho)}(X), \quad (5.32)$$

where

$$d_\mu d_\rho^{-1} X^{(\mu-\rho)} \sigma^{(\rho)}(X) \quad (5.33)$$

is the correction term with degree $\mu - (\rho - l_\rho)$.

- ~~Since ρ is chosen to maximize $\rho - l_\mu$, this choice of ρ is equivalent to minimize the degree $\mu - (\rho - l_\rho)$ of the correction term.~~

- **Correction:** Go back to the steps prior to the μ -th step and determine a polynomial $\sigma^{(\rho)}(X)$ such that ~~$d_{j+1} > d_j$ and j is maximum~~ ρ has the largest value, where l_ρ is the degree of $\sigma^{(\rho)}(X)$. Then

$$\sigma^{(\mu+1)}(X) = \sigma^{(\mu)}(X) - d_\mu d_\rho^{-1} X^{(\mu-\rho)} \sigma^{(\rho)}(X) \quad (6.19)$$

is the solution at the $(\mu + 1)$ -th step of the iteration process.

- Continue the above iterative process until $2t$ steps have been completed. At the $2t$ -th step, we have

$$\sigma(X) = \sigma^{(2t)}(X), \quad (6.20)$$

which is the minimum-degree polynomial whose coefficients satisfy the $2t - \nu$ generalized Newton's identities given by (6.13).

- If $\nu \leq t$ (the designed error-correcting capability), $\sigma^{(2t)}$ is unique and the true error-location polynomial with all its roots in $\text{GF}(q^m)$.

p is the maximum integer less than μ such that $\ell > p$.
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TABLE 7.2: Steps for finding the error-location polynomial of the (15,9) RS code over $GF(2^4)$.

μ	$\sigma^{(\mu)}(X)$	d_μ	l_μ	$\mu - l_\mu$
-1	1	1	0	-1
0	1	α^{12}	0	0
1	$1 + \alpha^{12}X$	α^7	1	0 (take $\rho = -1$)
2	$1 + \alpha^3X$	1	1	1 (take $\rho = 0$)
3	$1 + \alpha^3X + \alpha^3X^2$	α^7	2	1 (take $\rho = 0$)
4	$1 + \alpha^4X + \alpha^{12}X^2$	α^{10}	2	2 (take $\rho = 2$)
5	$1 + \alpha^7X + \alpha^4X^2 + \alpha^6X^3$	α^3	3	2 (take $\rho = 3$)
6	$1 + \alpha^7X + \alpha^4X^2 + \alpha^6X^3$	—	—	—

$$1 + \alpha^4X + \alpha^3X^2 + \alpha^3X^3$$

-1
0
6
2
2
4

Step 2. To find the error-location polynomial $\sigma(X)$, we fill out Table 7.1 and obtain Table 7.2. Thus, $\sigma(X) = 1 + \alpha^7X + \alpha^4X^2 + \alpha^6X^3$.

Step 3. By substituting $1, \alpha, \alpha^2, \dots, \alpha^{14}$ into $\sigma(X)$, we find that α^3, α^9 , and α^{12} are roots of $\sigma(X)$. The reciprocals of these roots are α^{12}, α^6 , and α^3 , which are the error-location numbers of the error pattern $e(X)$. Thus, errors occur at positions X^3, X^6 , and X^{12} .

Next, we need to determine the error values, by finding the error-value evaluator. We define the syndrome polynomial $S(X)$ as follows:

$$\begin{aligned} S(X) &\triangleq S_1 + S_2X + \dots + S_{2t}X^{2t-1} + S_{2t+1}X^{2t} + \dots \\ &= \sum_{j=1}^{\infty} S_j X^{j-1}. \end{aligned} \quad (7.21)$$

Note that only the coefficients of the first $2t$ terms are known. For $1 \leq j < \infty$, we also define

$$S_j = \sum_{l=1}^v \delta_l \beta_l^j. \quad (7.22)$$

The first $2t$ such S_j 's are simply the $2t$ equalities of (7.13). Combining (7.21) and (7.22), we can put $S(X)$ in the following form:

$$\begin{aligned} S(X) &= \sum_{j=1}^{\infty} X^{j-1} \sum_{l=1}^v \delta_l \beta_l^j \\ &= \sum_{l=1}^v \delta_l \beta_l \sum_{j=1}^{\infty} (\beta_l X)^{j-1}. \end{aligned} \quad (7.23)$$

Note that

$$\frac{1}{(1 - \beta_l X)} = \sum_{j=1}^{\infty} (\beta_l X)^{j-1}. \quad (7.24)$$