A parametric approach to list decoding of Reed-Solomon codes using interpolation

Mortuza Ali and Margreta Kuijper

April 15, 2011

Abstract

In this paper we present a minimal list decoding algorithm for Reed-Solomon (RS) codes. Minimal list decoding for a code \( C \) refers to list decoding with radius \( L \), where \( L \) is the minimum of the distances between the received word \( r \) and any codeword in \( C \). We consider the problem of determining the value of \( L \) as well as determining all the codewords at distance \( L \). Our approach involves a parametrization of interpolating polynomials of a minimal Gröbner basis \( G \). We present two efficient ways to compute \( G \). We also show that so-called re-encoding can be used to further reduce the complexity. We then demonstrate how our parametric approach can be solved by a computationally feasible rational curve fitting solution from a recent paper by Wu. Besides, we present an algorithm to compute the minimum multiplicity as well as the optimal values of the parameters associated with this multiplicity which results in overall savings in both memory and computation.

1 Introduction

Reed-Solomon (RS) codes are important linear block codes that are of significant theoretical and practical interest. A \( (n,k) \) RS code \( C \), defined over a finite field \( \mathbb{F} \), is a \( k \) dimensional subspace of the \( n \) dimensional space \( \mathbb{F}^n \). For a message polynomial \( m(x) = m_0 + m_1 x + \cdots + m_{k-1} x^{k-1} \), the encoding operation is to evaluate \( m(x) \) at \( x_1, x_2, \ldots, x_n \), where the \( x_i \)'s are \( n \) distinct elements of \( \mathbb{F} \). The rich algebraic properties and geometric structures of RS codes lead to the invention of a number of efficient decoding algorithms such as Sugiyama algorithm [26], Berlekamp-Massey (BM) algorithm [4, 20], and Welch-Berlekamp (WB) algorithm [28]. These classical decoding algorithms guarantee correct decoding as long as the number of errors is upper bounded by \( t = \lfloor (d-1)/2 \rfloor \), where \( d = n - k + 1 \) is the minimum distance of the code.

In classical decoding, the error correcting radius of \( t = \lfloor (d-1)/2 \rfloor \) originates from the requirement of unique decoding since for \( t > \lfloor (d-1)/2 \rfloor \) multiple codewords within distance \( t \) from the received word \( r \) may exist. One way to circumvent this limitation is to increase the decoding radius beyond \( \lfloor (d-1)/2 \rfloor \) and allow the decoder to output a list of codewords rather than one single codeword. However, such list decoding is only feasible if there are few codewords in the list. In [3] Guruswami showed that for a code of relative distance \( \delta = d/n \), any Hamming sphere of radius \( \leq n(1 - \sqrt{1 - \delta}) \) around a received word \( r \) contains only a polynomial number of codewords. Therefore, a \( (n,k) \) RS code with \( d = n - k + 1 \) can be list decoded up to the error correcting radius of \( n - \sqrt{n(k-1)} \) which Guruswami named as the Johnson bound.

\*M. Ali and M. Kuijper are with the Department of Electrical and Electronic Engineering, University of Melbourne, VIC 3010, Australia mortuzaa@unimelb.edu.au; mkuijper@unimelb.edu.au
†This work was supported by the Australian Research Council(ARC).
A list decoding algorithm was first discovered for low rate RS codes by Sudan [25] and later improved and extended for all rates by Guruswami and Sudan [10]. The Guruswami-Sudan algorithm can correct errors up to the Johnson bound $n - \sqrt{n(k - 1)}$. Given a received word $r$, the essential idea of the algorithm is to find all the polynomials $m$ of degree less than $k$ such that $m(x_i) \neq r_i$ for at most $t$ values of $i \in \{1, 2, \ldots, n\}$. The Guruswami-Sudan algorithm finds these polynomials in two steps: the interpolation step and the factorization step. In the interpolation step, it computes a bivariate polynomial $Q(x, r)$ that passes through all the points $(x_1, r_1), (x_2, r_2), \ldots, (x_n, r_n)$ with a prescribed multiplicity $s$ satisfying a certain weighted degree constraint (see [10] for the definition of weighted degree). Then the bivariate polynomial $Q(x, r)$ is factorized to find all the factors of the form $r - m(x)$, where $m$ is a polynomial of degree less than $k$. Now a polynomial $m$ is a valid message polynomial if it is of degree less than $k$ and $m(x_i) \neq r_i$ for at most $t$ values of $i \in \{1, 2, \ldots, n\}$. The construction of $Q(x, r)$ with the prescribed multiplicity and weighted degree constraint ensures that for all valid message polynomials $m$, $r - m(x)$ appears as a factor of $Q(x, r)$. Even though the algorithm may produce implausible polynomials, the total number of polynomials $L$ in the list will satisfy the bound $L < (s + 0.5)\sqrt{n/(k - 1)}$, see [21].

The most computationally intensive operation in the Guruswami-Sudan algorithm is the construction of the bivariate polynomial $Q(x, r)$. Computation of $Q(x, r)$ involves solving a system of $O(ns^2)$ homogeneous equations which using Gaussian elimination can be done in time cubic in the number of equations [27]. Clearly the algorithmic complexity of the interpolation step is dominated by the multiplicity $s$. Recently Wu [29] transformed the interpolation problem to a ‘rational interpolation problem’ which involves smaller multiplicity. Given the received word $r$, Wu’s algorithm first computes the syndrome $s$ of $r$ followed by the computation of the error locator polynomial $\Lambda$ and error correction polynomial $B$ using the Berlekamp-Massey algorithm. Wu demonstrated that all valid error locator polynomials can be expressed as a parametrization of $\Lambda$ and $B$. More specifically, given a list decoding radius $t$, Wu’s algorithm aims at finding all polynomials $\Lambda$ and $\beta$ such that $\Lambda' = \lambda \Lambda + \beta B$ has at most $t$ distinct roots. Wu showed that similar to the Guruswami-Sudan approach, this problem can be reduced to a curve fitting problem but with significantly smaller multiplicity.

It may be observed that the set of all $Q(x, r) \in F[x, r]$ passing through the points $(x_i, r_i)$, for $i = 1, 2, \ldots, n$, with multiplicity $s$ is an ideal $I_s$. From this observation several authors including Alekhnovich [2], Nielsen and Høholdt [23], Kuijper and Polderman [16], O’Keeffe and Fitzpatrick [24], and Lee and O’Sullivan [19], formulated the interpolation step of the list decoding algorithm as the problem of finding the minimal weight polynomial from the ideal $I_s$. Clearly the minimal weight polynomial will appear as the minimal polynomial in a minimal Gröbner basis of $I_s$ computed with respect to the corresponding weighted term order. Lee and O’Sullivan also showed that the minimal polynomial in the ideal $I_s$ can be computed more efficiently from a minimal Gröbner basis of a submodule of $F[x]^q$ for a sufficiently large $q$. Let $F[x, r]^q = \{ f \in F[x, r] | r-deg(f) < q \}$. Then $F[x, r]^q$ can be viewed as a free module over $F[x]^q$ with a free basis $1, r, \ldots, r^{q-1}$. Then the essential observation of Lee and O’Sullivan is that the minimal polynomial of $I_s$ can be constructed from the minimal Gröbner basis of a submodule of $F[x]^q$ along with the free basis $1, r, \ldots, r^{q-1}$, for large enough $q$.

In this paper we employ the theory of minimal Gröbner bases to perform minimal list decoding. Given the received word $r$, let $L$ denote the value of $d_H(r, C)$ where

$$d_H(r, C) := \min_{c \in C} \{ d_H(r, c) \}.$$ 

Our main objective is to determine the value of $L$ as well as all codewords $c$ which are at a distance $L$ from the received word $r$. Clearly, if $L$ is larger than the classical error correcting radius $\lfloor (d - 1)/2 \rfloor$,

---

1 Here the integer $q$ is not related to the size of the field.
the task is a list decoding operation. Our algorithm, unlike the Lee and O'Sullivan approach, starts with computing a minimal Gröbner basis $G$ of a submodule of $\mathbb{F}[x]^2$, rather than $\mathbb{F}[x]^q$. We then demonstrate that all valid message polynomials can be extracted from a parametrization in terms of the elements of $G$. For computational feasibility, we show that this parametric approach, like Wu's algorithm, can be translated into a 'rational interpolation problem'. However, our approach has at least three features that distinguishes it from Wu's algorithm. Firstly, our parametric formulation of the problem of list decoding, without the detour of syndrome computation, is simpler than Wu’s formulation. Secondly, while Wu's algorithm, for each valid $\Lambda'$, resorts to Forney's formula to compute the error values, our algorithm immediately leads to a valid message polynomial. Finally, we provide an algorithm to compute the minimum multiplicity along with the optimal values of the associated parameters to be used in the rational interpolation step. Use of these optimal parameters in the rational interpolation step results in savings of both memory and computation as compared to Wu’s algorithm.

The organization of the rest of the paper is as follows. In Section 2, we briefly review the relevant theory of Gröbner bases. In Section 3, we develop the theory and present the main algorithm along with two ways to compute the minimal Gröbner basis. In this section we also explain how so-called re-encoding can be applied to the proposed approach. In Section 4, we translate the parametric approach into a 'rational interpolation problem' and present an efficient algorithm for the computation of the minimum value of the multiplicity and other parameters to be used in the rational interpolation step. We demonstrate that the use of these optimal values of the parameters results in less memory requirement as well as less computational requirement as compared to Wu’s approach. Finally we conclude the paper in Section 5.

2 Preliminaries

The theory of Gröbner bases for modules in $\mathbb{F}[x]^q$ is generally recognized as a powerful conceptual and computational tool that plays a role similar to Euclidean division for modules in $\mathbb{F}[x]$. More specifically, minimal Gröbner bases prove themselves as an effective tool for various types of interpolation problems. In recent papers [18, 17] this effectiveness was ascribed to a powerful property of minimal Gröbner bases, explicitly identified as the ‘Predictable Leading Monomial Property’. The proofs in this paper make use of this property. Before recalling the PLM property let us first recall some terminology on Gröbner bases.

Let $e_1, \ldots, e_q$ denote the unit vectors in $\mathbb{F}^q$. The elements $x^\alpha e_i$ with $i \in \{1, \ldots, l\}$ and $\alpha \in \mathbb{N}_0$ are called monomials. Let $n_1, \ldots, n_q$ be nonnegative integers. In this paper we define the following two types of monomial orders:

- The $(n_1, \ldots, n_q)$-weighted term over position (top) order, defined as
  \[ x^\alpha e_i < x^\beta e_j : \Leftrightarrow \alpha + n_i < \beta + n_j \text{ or } (\alpha + n_i = \beta + n_j \text{ and } i < j). \]

- The $(n_1, \ldots, n_q)$-weighted position over term (pot) order, defined as
  \[ x^\alpha e_i < x^\beta e_j : \Leftrightarrow i < j \text{ or } (i = j \text{ and } \alpha + n_i < \beta + n_j). \]

Clearly, whatever order is chosen, every nonzero element $f \in \mathbb{F}[x]^q$ can be written uniquely as

\[ f = \sum_{i=1}^{L} c_i X_i, \]

where $L \in \mathbb{N}$, the $c_i$'s are nonzero elements of $\mathbb{F}$ for $i = 1, \ldots, L$ and the polynomial vectors $X_1, \ldots, X_L$ are monomials, ordered as $X_1 > \cdots > X_L$. Using the terminology of [1] we define
• $\text{lm}(f) := X_1$ as the leading monomial of $f$

• $\text{lt}(f) := c_1 X_1$ as the leading term of $f$

• $\text{lc}(f) := c_1$ as the leading coefficient of $f$

Writing $X_1 = x^{\alpha_1} e_1$, where $\alpha_1 \in \mathbb{N}_0$ and $i_1 \in \{1, \ldots, l\}$, we define

• $\text{lpos}(f) := i_1$ as the leading position of $f$

• $\text{wdeg}(f) := \alpha_1 + n_i$ as the weighted degree of $f$.

Note that for zero weights $n_1 = \cdots = n_q = 0$ the above orders coincide with the reflected versions of the standard TOP order and POT order, respectively, as introduced in the textbook [1].

Also note that, unlike with TOP, the introduction of weights does not change the POT ordering of monomials. In this paper, weighted POT order is needed only because we need the associated notion of ‘weighted degree’.

We now recall some basic definitions and results on Gröbner bases, see [1]. Below we denote the submodule generated by a polynomial vector $f$ by \langle f \rangle.

**Definition 2.1** Let $F$ be a subset of $F[x]^q$. Then the submodule $\mathcal{L}(F)$, defined as

$$\mathcal{L}(F) := \langle \text{lt}(f) \mid f \in F \rangle$$

is called the leading term submodule of $F$.

**Definition 2.2** Let $M \subseteq F[x]^q$ be a module and $G \subseteq M$. Then $G$ is called a Gröbner basis of $M$ if

$$\mathcal{L}(G) = \mathcal{L}(M).$$

In order to define a concept of minimality we have the following definition.

**Definition 2.3** ([1, Def. 4.1.1]) Let $0 \neq f \in F[x]^q$ and let $F = \{f_1, \ldots, f_s\}$ be a set of nonzero elements of $F[x]^q$. Let $\alpha_{j_1}, \ldots, \alpha_{j_m} \in \mathbb{N}_0$ and $\beta_{j_1}, \ldots, \beta_{j_m}$ be nonzero elements of $F$, where $1 \leq m \leq s$ and $1 \leq j_i \leq s$ for $i = 1, \ldots, m$, such that

1. $\text{lm}(f) = x^{\alpha_{j_i}} \text{lm}(f_{j_i})$ for $i = 1, \ldots, m$ and

2. $\text{lt}(f) = \beta_{j_1} x^{\alpha_{j_1}} \text{lt}(f_{j_1}) + \cdots + \beta_{j_m} x^{\alpha_{j_m}} \text{lt}(f_{j_m})$.

Define

$$h := f - (\beta_{j_1} x^{\alpha_{j_1}} f_{j_1} + \cdots + \beta_{j_m} x^{\alpha_{j_m}} f_{j_m}).$$

Then we say that $f$ reduces to $h$ modulo $F$ in one step and we write

$$f \xrightarrow{F} h.$$

If $f$ cannot be reduced modulo $F$, we say that $f$ is minimal with respect to $F$.

**Lemma 2.4** ([1, Lemma 4.1.3]) Let $f, h$ and $F$ be as in the above definition. If $f \xrightarrow{F} h$ then $h = 0$ or $\text{lm}(h) < \text{lm}(f)$.

**Definition 2.5** ([1]) A Gröbner basis $G$ is called minimal if all its elements $g$ are minimal with respect to $G \setminus \{g\}$. 
It is well known [1, Exercise 4.1.9] that a minimal Gröbner basis exists for any module in $\mathbb{F}[x]^q$ and that all leading positions of its elements are different. In [18, 17] another important property of a minimal Gröbner basis is identified; the theorem below merely formulates a well known result.

**Theorem 2.6** ([18]) Let $G$ be a submodule of $\mathbb{F}[x]^q$ with minimal Gröbner basis $G = \{g_1, \ldots, g_m\}$. Then for any $0 \neq f \in M$, written as

$$f = a_1g_1 + \cdots + a_ng_n,$$

where $a_1, \ldots, a_n \in \mathbb{F}[x]$, we have

$$\text{lm}(f) = \max_{1 \leq i \leq n; a_i \neq 0} (\text{lm}(a_i) \text{lm}(g_i)).$$

The property outlined in the above theorem is called the Predictable Leading Monomial (PLM) property, as in [18]. Note that this property involves not only degree information (as in the ‘predictable degree property’ first introduced in [5]) but also leading position information. Most importantly, the above theorem holds irrespective of which monomial orders top or pot is chosen, for a proof see [18].

Clearly, in the above theorem $m = \text{rank } (M)$ and all minimal Gröbner bases of $M$ must have rank $(M)$ elements, no matter which monomial order is chosen. Furthermore, we have the following theorem.

**Theorem 2.7** Let $n_1, \ldots, n_q$ be nonnegative integers and let $M$ be a module in $\mathbb{F}[x]^q$. Let $G = \{g_1, \ldots, g_m\}$ be a minimal Gröbner basis of $M$ with respect to the $(n_1, \ldots, n_q)$-weighted top order; denote $\ell_i := \text{wdeg } g_i$ for $i = 1, \ldots, m$. Let $\tilde{G} = \{\tilde{g}_1, \ldots, \tilde{g}_m\}$ be a minimal Gröbner basis of $M$ with respect to the $(n_1, \ldots, n_q)$-weighted pot order; denote $\tilde{\ell}_i := \text{wdeg } \tilde{g}_i$ for $i = 1, \ldots, m$. Then

$$\sum_{i=1}^m \ell_i = \sum_{i=1}^m \tilde{\ell}_i.$$  

**Proof** We first prove the theorem for the case $m = q$. It follows easily from the fact that both $G$ and $\tilde{G}$ are bases for $M$ (in a linear algebraic sense) that there exists a unimodular polynomial matrix $U \in \mathbb{F}[x]^{q \times q}$ such that

$$\text{col } \{g_1, \ldots, g_q\} = U \text{col } \{\tilde{g}_1, \ldots, \tilde{g}_q\}. $$

Without restrictions we may assume that the leading positions within each Gröbner basis are strictly increasing. Clearly it follows from the above equation that also

$$V = UW,$$

where $V = \text{col } \{g_1, \ldots, g_q\} \text{diag } \{x^{n_1}, \ldots, x^{n_q}\}$ and $W = \text{col } \{\tilde{g}_1, \ldots, \tilde{g}_q\} \text{diag } \{x^{\ell_1}, \ldots, x^{\ell_q}\}$. Since $U$ is unimodular we must have $\text{deg } \text{det } V = \text{deg } \text{det } W$. Clearly $\text{deg } \text{det } V = \sum_{i=1}^m \ell_i$ and $\text{deg } \text{det } W = \sum_{i=1}^m \tilde{\ell}_i$ from which (3) follows. Next, we prove the general case $m \leq q$. For this, we note that it follows immediately from [4] that the maximum degree of all minors of $V$ equals the maximum degree of all minors of $W$. On the other hand, the maximum degree of all minors of $V$ clearly equals $\sum_{i=1}^m \ell_i$ and similarly the maximum degree of all minors of $W$ equals $\sum_{i=1}^m \tilde{\ell}_i$. The theorem now follows.

We call the sum in (3) the $(n_1, \ldots, n_q)$-weighted degree of $M$, denoted by $\text{wdeg } (M)$. For zero weights $n_1 = \cdots = n_q = 0$ the above result expresses that the sum of the degrees of a (reflected)
We introduce the following two polynomials in \( \mathbb{F}_3 \).

### 3 Minimal list decoding through division

Let us now consider a \((n, k)\) RS code and a nonnegative integer \( t \). The problem of ‘list decoding up to \( t \) errors’ is the following:

**List Decoding Problem:** Given a received word \((r_1, \ldots, r_n) \in \mathbb{F}^n\), find all polynomials \( m \in \mathbb{F}[x] \) of degree less than \( k \) such that

\[
m(x_i) = r_i \quad \text{for at least } n - t \text{ values of } i \in \{1, \ldots, n\}.
\]

#### 3.1 Main approach

We introduce the following two polynomials in \( \mathbb{F}[x] \):

\[
\Pi(x) = \prod_{i=1}^{n}(x - x_i),
\]

and \( \mathcal{L} \) as the Lagrange interpolating polynomial, i.e., the polynomial of least degree for which

\[
\mathcal{L}(x_i) = r_i \quad \text{for all } i \in \{1, \ldots, n\}.
\]

**Definition 3.1** Let \( r = (r_1, \ldots, r_n) \in \mathbb{F}^n \). The interpolation module \( M(r) \) is given by the module in \( \mathbb{F}[x]^2 \) that is spanned by the vectors \( \tilde{g}_1 := [\Pi(x) \quad 0] \) and \( \tilde{g}_2 := [\mathcal{L}(x) - 1] \).

Note that \( \{\tilde{g}_1, \tilde{g}_2\} \) is a minimal \textit{pot} Gröbner basis for \( M(r) \). The above defined interpolation module is crucial to our approach. With \( \tilde{g}_2 \) we associate the bivariate polynomial \( Q_2(x, y) = \mathcal{L}(x) - y \); clearly \( Q_2(x_i, r_i) = 0 \) for all \( i \in \{1, \ldots, n\} \). Similarly, with \( \tilde{g}_1 \) we associate the polynomial \( Q_1(x, y) = \Pi(x) \); trivially \( Q_1(x_i, r_i) = 0 \) for all \( i \in \{1, \ldots, n\} \). Now consider an arbitrary bivariate polynomial \( Q \) of the form \( Q(x, y) = N(x) - D(x)y \) for which \( Q(x_i, r_i) = 0 \) for all \( i \in \{1, \ldots, n\} \). It can be shown, see [10], that \([N - D] \in M(r)\). Recall that list decoding up to \( t \) errors amounts to finding all polynomials \( m \in \mathbb{F}[x] \) of degree less than \( k \) such that

\[
m(x_i) = r_i \quad \text{for all } i \in \{1, \ldots, n\} \text{ except } i = j_1, \ldots, j_L \text{ with } L \leq t.
\]

In our context this amounts to looking for an interpolating bivariate polynomial \( Q \) of the form \( Q(x, y) = D(x)m(x) - D(x)y \), where \( D(x) = \prod_{j=1}^{L}(x - x_{j}) \). Note that then indeed \( Q(x_i, r_i) = 0 \) for all \( i \in \{1, \ldots, n\} \). Thus, to solve the above list decoding problem we are looking for particular vectors \([N - D] \in M(r)\) of weighted \((0, k - 1)\)-degree \( \leq t + k - 1 \), that satisfy
1. \( N \) is a multiple of \( D \) and

2. \( D \) has \( L \) distinct zeros in \( \mathbb{F} \), where \( L \) denotes \( \text{deg} \ D \).

In this paper we are interested in finding the smallest value \( L = d_H(r, C) \) for which list decoding is possible as well as performing the associated list decoding. Thus we occupy ourselves with maximum likelihood list decoding. We have the following theorem.

**Theorem 3.2** Let \( r = (r_1, \ldots, r_n) \in \mathbb{F}^n \) be a received word and let \( M(r) \) be the corresponding interpolation module. Let \( f = \begin{bmatrix} f^{(1)} & f^{(2)} \end{bmatrix} \in \mathbb{F}[x]^2 \) be a vector in \( M(r) \) of weighted \((0, k-1)\)-degree \( L \) that satisfies the following 3 requirements:

1. \( \text{lpos}(f) = 2 \),
2. \( f^{(1)} \) is a multiple of \( f^{(2)} \) and
3. there is no vector in \( M(r) \) of weighted \((0, k-1)\)-degree \(< L \) that satisfies requirements 1) and 2).

Then

\[
m := -\frac{f^{(1)}}{f^{(2)}}
\]

is a message polynomial corresponding to a minimal error pattern of \( L - k + 1 \) errors.

**Proof** From \( \text{lpos}(f) = 2 \) it follows immediately that \( \text{deg} \ m < k \) and \( \text{deg} \ f^{(2)} = L - k + 1 \). It remains to prove that \( f^{(2)} \) has \( L - k + 1 \) distinct zeros in \( \mathbb{F} \). Since \( f \in M(r) \) there exist polynomials \( \alpha \) and \( \beta \) such that

\[
f = \begin{bmatrix} \alpha & \beta \end{bmatrix} \begin{bmatrix} \Pi & 0 \\ \mathcal{L} & -1 \end{bmatrix}.
\]

(7)

Observe that \( \alpha \) and \( \beta \) do not have a common factor, otherwise the weighted degree of \( f \) would not be minimal (requirement 3). From (7) it follows that \( \alpha \Pi - f^{(2)} \mathcal{L} = f^{(1)} \) is a multiple of \( f^{(2)} \) by requirement 2. As a result, \( \alpha \Pi \) is a multiple of \( f^{(2)} \). Since \( \alpha \) and \( \beta = -f^{(2)} \) have no common factor it follows that \( \Pi \) must be a multiple of \( f^{(2)} \), i.e., \( f^{(2)} \) has \( L - k + 1 \) distinct zeros in \( \mathbb{F} \), which proves the theorem. \( \square \)

**Lemma 3.3** Let \( r = (r_1, \ldots, r_n) \in \mathbb{F}^n \) be a received word and let \( M(r) \) be the corresponding interpolation module. Let \( \{g_1, g_2\} \) be a \((0, k-1)\)-weighted top minimal Gröbner basis for \( M(r) \) with \( \text{lpos}(g_2) = 2 \). Denote \( \ell_1 := \text{wdeg} \ g_1 \) and \( \ell_2 := \text{wdeg} \ g_2 \). Let \( t \) be a nonnegative integer. Then a parametrization of all vectors \( f \in \mathbb{F}[x]^2 \) with \( \text{lpos}(f) = 2 \) and \( \text{wdeg} \ f = t + k - 1 \) (with respect to the \((0, k-1)\)-weighted top order) is given by

\[
f = ag_1 + bg_2,
\]

where \( a \in \mathbb{F}[x] \) with \( \text{deg} \ a \leq t + k - 1 - \ell_1 \) and \( b \) is a monic polynomial in \( \mathbb{F}[x] \) of degree \( t + k - 1 - \ell_2 \). In particular, there exist no such vectors \( f \) for \( t < \ell_2 - k + 1 \).
**Algorithm 1** Minimal list decoding of \((n,k)\) RS code

**Input:** Received word \(r = (r_1, \ldots, r_n)\)

**Output:** A list of polynomials \(m\) of degree < \(k\) such that \(d_H(c, r)\) is minimal, where \(c = (m(x_1), \ldots, m(x_n))\).

1. Compute the polynomials \(\Pi\) and \(L\) given by (5) and (6); define the interpolation module \(M(r) := \text{span} \{[\Pi 0], [L \ -1]\}\).
2. Compute a minimal Gröbner basis \(G = \{g_1, g_2\}\) of \(M(r)\) with respect to the \((0,k-1)\)-weighted \(\text{top}\) monomial order, with \(\text{lpos}(g_2) = 2\). Denote \(\ell_1 := \text{wdeg} g_1\) and \(\ell_2 := \text{wdeg} g_2\); set \(j = 0\).
3. Check requirement 2) of Theorem 3.2 for \(f = ag_1 + bg_2\), for all \(a \in F[x]\) with \(\text{deg} a \leq \ell_2 - \ell_1 + j\) and for all monic \(b \in F[x]\) with \(\text{deg} b = j\); write \(f = [f^{(1)} f^{(2)}]\).
4. Whenever step 3) is successful, output all obtained quotient polynomials, i.e., polynomials \(m\) of the form \(m = -f^{(1)}/f^{(2)}\). In case step 3) is not successful increase \(j\) by 1 and repeat step 3).

**Proof** According to Theorem 2.6, \(\{g_1, g_2\}\) has the PLM property with respect to the \((0,k-1)\)-weighted \(\text{top}\) order. The parametrization now follows immediately from this property. \(\square\)

Together, the above lemma and theorem give rise to the heuristic list decoding Algorithm 1. An important feature of the above algorithm is that we use \(\ell_2 = \text{wdeg} g_2\) to decide how many errors to decode. Indeed, it follows from the above lemma that it is not possible to perform list decoding for \(t < \ell_2 - k + 1\). We now present the main theorem of this section.

**Theorem 3.4** Let \(r = (r_1, \ldots, r_n) \in F^n\) be a received word and let \(M(r)\) be the corresponding interpolation module. Let \(\{g_1, g_2\}\) be a \((0,k-1)\)-weighted \(\text{top}\) minimal Gröbner basis for \(M(r)\) with \(\text{lpos}(g_2) = 2\). Write \(g_2 = [g_2^{(1)} g_2^{(2)}]\). Then Algorithm 1 yields a list of all message polynomials \(m\) such that
\[
d_H(c, r) \text{ is minimal, where } c = (m(x_1), \ldots, m(x_n)).
\] (8)

In particular, in case there exists an error pattern with only \(\lfloor (n - k)/2 \rfloor\) errors, the list consists of only
\[
m = -g_2^{(1)}/g_2^{(2)}.\] (9)

**Proof** Firstly, it follows immediately from Theorem 3.2 and Lemma 3.3 that any polynomial \(m\) that is output by Algorithm 1 has to have degree < \(k\) and satisfy (8). Vice versa, if \(m\) is a polynomial of degree < \(k\) that satisfies (8) then it follows from Lemma 3.3 that it must be in the output list of Algorithm 1. Finally, let us assume that there are only \(\lfloor (n - k)/2 \rfloor\) errors. This implies that there exists a vector \(f = [f^{(1)} f^{(2)}]\) in \(M(r)\) with \(\text{wdeg} f \leq \lfloor (n - k)/2 \rfloor + k - 1 < (n + k - 1)/2\) that satisfies the requirements of Theorem 3.2. Because of Lemma 3.3 it follows that \(\ell_2 < (n + k - 1)/2\). Now, since \(\ell_1 + \ell_2 = n + k - 1\) by Theorem 2.7 this implies that \(\ell_1 > \ell_2\). As a result, \(a = 0\) in step 3), so that step 4) immediately gives the unique solution for \(j = 0\) as (9). \(\square\)

Our next example illustrates the classical decoding scenario, showing that Algorithm 1 is an extension of existing classical interpolation-based algorithms as in [19, 7].

**Example 3.5** Consider the single-error correcting \((7,5)\) RS code over \(GF(7)\). The message polynomial \(m(x) = 2x^2 + x + 3\) is encoded as \(c = (m(0), m(1), \ldots, m(6)) = (3, -1, -1, 3, -3, 2, -3)\). Let
the received word be \( r = (3, 2, -1, 3, -3, 2, -3) \). Thus an error occurred at locator position 1. The polynomials \( L \) and \( \Pi \) are computed as \( L(x) = -3x^6 - 3x^5 - 3x^4 - 3x^3 - x^2 - 2x + 3 \) and \( \Pi(x) = x^7 - x \). Thus the module \( M(r) \) is spanned by the rows of the matrix

\[
\begin{pmatrix}
    x^7 - x \\
    -3x^6 - 3x^5 - 3x^4 - 3x^3 - x^2 - 2x + 3 \\
    0
\end{pmatrix}.
\]

A minimal Gröbner basis \( \{g_1, g_2\} \) of \( M(r) \) with respect to the \((0,4)\)-weighted top monomial order is computed as

\[
\text{col} \{g_1, g_2\} = \begin{pmatrix}
    -3x^6 - 3x^5 - 3x^4 - 3x^3 - x^2 - 2x + 3 \\
    2x^3 - x^2 + 2x - 3 \\
    -x + 1
\end{pmatrix}.
\]

Thus, in the terminology of Theorem 3.4 we have \( g_2^{(1)} = 2x^3 - x^2 + 2x - 3 \) and \( g_2^{(2)} = -x + 1 \). Applying Algorithm 1, we determine that \( g_2^{(1)} \) is a multiple of \( g_2^{(2)} \) and we recover

\[
m(x) = \frac{g_2^{(1)}}{g_2^{(2)}} = 2x^2 + x + 3.
\]

Let us now move on to an example of decoding beyond the classical error bound. Our approach is particularly feasible for the case that \( b = 1 \) and \( a \) is restricted to a constant, as illustrated in the next example. Note that the example is an instance of “one-step-ahead” list decoding [29].

Example 3.6 Consider the single-error correcting \((7,4)\) RS code over GF(7); let the message polynomial be \( m(x) = 2x^2 + x + 3 \) which is encoded as \( c = (m(0), m(1), \ldots, m(6)) = (3, -1, -1, 3, -3, 2, -3) \). Let the received word be \( r = (3, 2, -1, 3, 2, -3) \) which differs from \( c \) at locations 1 and 4. The polynomials \( L \) and \( \Pi \) are computed as \( L(x) = -x^6 - 2x^5 + x^4 - x^3 + 2x + 3 \) and \( \Pi(x) = x^7 - x \). The interpolation module \( M(r) \) is spanned by the rows of the matrix

\[
M(r) = \begin{pmatrix}
    x^7 - x \\
    -x^6 - 2x^5 + x^4 - x^3 + 2x + 3 \\
    0
\end{pmatrix}.
\]

A minimal Gröbner basis \( \{g_1, g_2\} \) of \( M(r) \) with respect to the \((0,3)\)-weighted top monomial ordering is computed as

\[
\text{col} \{g_1, g_2\} = \begin{pmatrix}
    x^5 - 2x^4 - x^3 - x^2 + x + 3 \\
    -2x^4 + 2x^3 + x^2 - 3x + 2 \\
    -3x - 1
\end{pmatrix}.
\]

Thus in this example \( \ell_1 = \ell_2 = 5 \), so that \( a \) is a constant. Applying Algorithm 1, we consider \( f = ag_1 + g_2 \) for \( a = 0, \ldots, 6 \). Writing \( f = [f^{(1)} \ f^{(2)}] \), we find that \( f^{(2)} \) divides \( f^{(1)} \) for \( a = 0, 2, \) and \( 4 \), giving a list of three message polynomials—we recover not only \( m(x) = 2x^2 + x + 3 \) (for \( a = 0 \)), but also the message polynomials \( 3x^3 - 2x^2 + 3x - 2 \) (for \( a = 2 \)), and \( -2x^3 - 2x^2 + 3x + 3 \) (for \( a = 4 \)).

3.2 Computation of \( g_1 \) and \( g_2 \)

There are various ways in which the required minimal Gröbner basis \( \{g_1, g_2\} \) of the interpolation module \( M(r) \) can be computed. One obvious way is to simply run an existing computer algebra system such as SINGULAR, specifying the required \((0, k - 1)\)-weighted top order. Because of the specific form of \( M(r) \) a more efficient way is to apply the Euclidean algorithm to the polynomials \( \Pi \) and \( L \). More specifically, we have the following algorithm.
Algorithm 2 Computation of $g_1$ and $g_2$ via Euclidean algorithm

**Input**: Received word $r = (r_1, \ldots, r_n)$; polynomials $\Pi$ and $\mathcal{L}$ given by (5) and (6).

**Output**: Polynomials $g_1$ and $g_2$ in $\mathbb{F}[x]^2$, such that $\{g_1, g_2\}$ is a minimal Gröbner basis of $M(r)$ with respect to the $(0, k-1)$-weighted top monomial order, with $\text{lpos}(g_2) = 2$.

1. Define polynomials $h_0$, $h_1$, $t_0$ and $t_1$ in $\mathbb{F}[x]$ as
   $$\begin{bmatrix} h_0 & t_0 \\ h_1 & t_1 \end{bmatrix} := \begin{bmatrix} \Pi & 0 \\ \mathcal{L} & -1 \end{bmatrix};$$

   set $j := 0$.

2. Check
   $$\text{deg } t_{j+1} + k - 1 \geq \text{deg } h_{j+1};$$

   if NO, go to Step 3. If YES, define $g_1 := [h_j \ t_j]$ and $g_2 := [h_{j+1} \ t_{j+1}]$ and STOP.

3. Apply the Euclidean algorithm to $h_j$ and $h_{j+1}$, yielding
   $$h_j = q_{j+1}h_{j+1} + h_{j+2}, \text{ where } \text{deg } h_{j+2} < \text{deg } h_{j+1}.$$  

4. Write
   $$\begin{bmatrix} h_{j+1} & t_{j+1} \\ h_{j+2} & t_{j+2} \end{bmatrix} := \begin{bmatrix} 0 & 1 \\ 1 & -q_{j+1} \end{bmatrix} \begin{bmatrix} h_j & t_j \\ h_{j+1} & t_{j+1} \end{bmatrix};$$

   increase $j$ by 1 and go back to Step 2.

---

**Theorem 3.7** Let $r = (r_1, \ldots, r_n) \in \mathbb{F}^n$ be a received word and let $M(r)$ be the corresponding interpolation module. Then Algorithm 2 yields a $(0, k-1)$-weighted top minimal Gröbner basis $\{g_1, g_2\}$ for $M(r)$ with $\text{lpos}(g_2) = 2$.

**Proof** First we note that the matrix

$$\begin{bmatrix} 0 & 1 \\ 1 & -q_{j+1} \end{bmatrix}$$

is unimodular, i.e., has a polynomial inverse. It then follows that, at each step $j$, the rows of the matrix

$$\begin{bmatrix} h_j & t_j \\ h_{j+1} & t_{j+1} \end{bmatrix}$$

are a pot minimal Gröbner basis for $M(r)$ whose $(0, k-1)$-weighted pot degrees add up to $n+k-1$. By definition, with respect to the $(0, k-1)$-weighted top order both these row vectors have leading position 1, until the stopping condition (10) is met. At this point the second row vector has leading position 2 and the sum of the $(0, k-1)$-weighted top degrees add up to $n+k-1$. It now follows from Corollary 2.8 that the rows of the matrix (11) must be a $(0, k-1)$-weighted minimal top Gröbner basis for $M(r)$. $$\square$$

Yet another alternative is to use an iterative method, interpolating the $x_i$’s step by step for $i = 1, \ldots, n$. This method has the advantage that the Lagrange polynomial $\mathcal{L}$ does not need to be computed upfront.

**Theorem 3.8** Let $r = (r_1, \ldots, r_n) \in \mathbb{F}^n$ be a received word and let $M(r)$ be the corresponding interpolation module. Then Algorithm 3 yields a $(0, k-1)$-weighted top minimal Gröbner basis $\{g_1, g_2\}$ for $M(r)$ with $\text{lpos}(g_2) = 2$. 
Algorithm 3 Computation of $g_1$ and $g_2$ via iterative algorithm

**Input:** Received word $r = (r_1, \ldots, r_n)$.

**Output:** Polynomials $g_1$ and $g_2$ in $\mathbb{F}[x]^2$, such that $\{g_1, g_2\}$ is a minimal Gröbner basis of $M(r)$ with respect to the $(0, k-1)$-weighted top monomial order, with $\text{lpos}(g_2) = 2$.

1. Initialize $L_0 := k - 1$ and $R_0 := I \in \mathbb{F}^{2 \times 2}$; denote $R_j := \begin{bmatrix} Q_j & -K_j \\ N_j & -D_j \end{bmatrix} \in \mathbb{F}[x]^{2 \times 2}$ for $j = 0, \ldots, n$.
2. Process the received values $r_j$ iteratively for $j = 1$ to $n$ as follows. For $j = 1$ to $n$ do
   1. compute $\Gamma_j := Q_{j-1}(x_j) - r_j K_{j-1}(x_j)$ and $\Delta_j := N_{j-1}(x_j) - r_j D_{j-1}(x_j)$
   2. define $R_j := V_j R_{j-1}$, where
      \begin{align*}
      V_j &:= \begin{bmatrix} \Delta_j & -\Gamma_j \\ 0 & x - x_j \end{bmatrix} \text{ and } L_j := L_{j-1} + 1 \text{ if } \Delta_j \neq 0 \text{ and } (L_{j-1} < (j + k - 1)/2 \text{ or } \Gamma_j = 0), \\
      \end{align*}
      \begin{align*}
      V_j &:= \begin{bmatrix} x - x_j \\ \Delta_j & 0 \end{bmatrix} \text{ and } L_j := L_{j-1} \text{ otherwise}
      \end{align*}
3. Define $g_1 := [Q_n \ - K_n]$ and $g_2 := [N_n \ - D_n]$.

**Proof** For $j = 1, \ldots, n$ denote the interpolation module associated with $r_1, \ldots, r_j$ by $M(r_1, \ldots, r_j)$. We show that the rows of $R_j$ are a Gröbner basis of $M(r_1, \ldots, r_j)$ of the required form for $j = 1, \ldots, n$. We interpret $L_j$ as the $(0, k-1)$-weighted top degree of the second row of $R_j$. Clearly this is true for $j = 1$. Let us now proceed by induction and assume that this is true for $j - 1 \in \{0, \ldots, n-1\}$. By definition of $V_j$ and the induction assumption the rows of $R_j$ are a Gröbner basis for $M(r_1, \ldots, r_j)$. Also, by construction, their $(0, k-1)$-weighted top degrees add up to 1 more than the $(0, k-1)$-weighted top degrees of $R_{j-1}$. Then, by induction, the $(0, k-1)$-weighted top degrees of $R_j$ add up to $j + k - 1 = \text{wdeg} (M(r_1, \ldots, r_j))$. It then follows from Corollary 2.8 that the rows of $R_j$ are a $(0, k-1)$-weighted top minimal Gröbner basis for $M(r_1, \ldots, r_j)$. Finally, by construction and the induction hypothesis, it is easily seen that the second row of $R_j$ has leading position 2. This proves the theorem.

3.3 The special case $r = (y_1, \ldots, y_{n-k}, 0, \ldots, 0)$

In this subsection we pay special attention to the case that the received word $r$ is of the form $(y_1, \ldots, y_{n-k}, 0, \ldots, 0) \in \mathbb{F}^n$. This comes about when so-called "re-encoding" is used in advance of RS decoding, see e.g., [13] [12].

First we introduce the polynomial $G \in \mathbb{F}[x]$ of degree $k - 1$ as

$$ G := \prod_{i=n-k+2}^{n} (x - x_i). $$

Clearly, the polynomials $\Pi$ and $\mathcal{L}$ of the previous subsection can be written as

$$ \Pi = \Pi_0 G $$

(13)
\[\mathcal{L} = \mathcal{L}_\varphi G,\]  

(14)

where \(\Pi\) and \(\mathcal{L}_\varphi\) are in \(\mathbb{F}[x]\). The following lemma is straightforward.

**Lemma 3.9** Let \((y_1,\ldots,y_{n-k}) \in \mathbb{F}^{n-k}\), \(\mathbf{r} = (y_1,\ldots,y_{n-k},0,\ldots,0) \in \mathbb{F}^n\) and let \(\Pi,\mathcal{L},G,\Pi_\varphi\) and \(\mathcal{L}_\varphi\) be defined as before. Let \(M(\mathbf{r}) := \text{span}\{[\Pi \ 0],[\mathcal{L} \ -1]\}\) as before and define \(M^*(\mathbf{y}) := \text{span}\{[\Pi_\varphi \ 0],[\mathcal{L}_\varphi \ -1]\}\). Then the following two statements are equivalent:

- \(\{g_1,g_2\}\) is a minimal Gröbner basis of \(M^*(\mathbf{y})\) with respect to the unweighted top order, with \(\text{lpos}(g_2) = 2\)
- \(\{g_1,g_2\}\) is a minimal Gröbner basis of \(M(\mathbf{r})\) with respect to the \((0,k-1)\)-weighted top order, with \(\text{lpos}(g_2) = 2\),

where \(g_i = \begin{bmatrix} g_i^{(1)} \\ g_i^{(2)} \end{bmatrix}\) and \(\tilde{g}_i = \begin{bmatrix} g_i^{(1)}G \\ g_i^{(2)} \end{bmatrix}\) for \(i = 1,2\).

Because of the above lemma it is now straightforward to modify Algorithm 1 into Algorithm 4.

**Algorithm 4** Minimal list decoding of \((n,k)\) RS code for re-encoded received word

- **Input:** Received word \(\mathbf{y} = (y_1,\ldots,y_{n-k})\) in \(\mathbb{F}^{n-k}\).
- **Output:** A list of polynomials \(m\) of degree \(<k\) such that \(d_H(\mathbf{c},\mathbf{r})\) is minimal, where \(\mathbf{c} = (m(x_1),\ldots,m(x_n))\) and \(\mathbf{r} = (y_1,\ldots,y_{n-k},0,\ldots,0)\) in \(\mathbb{F}^n\).

1. Compute the polynomials \(\Pi_\varphi\) and \(\mathcal{L}_\varphi\) given by (13) and (14) ; define the interpolation module \(M(\mathbf{y}) := \text{span}\{[\Pi_\varphi \ 0],[\mathcal{L}_\varphi \ -1]\}\).
2. Compute a minimal Gröbner basis \(G = \{g_1,g_2\}\) of \(M(\mathbf{y})\) with respect to the unweighted top monomial order, with \(\text{lpos}(g_2) = 2\); set \(j = 0\).
3. Compute \(f = ag_1 + bg_2\), for all \(a \in \mathbb{F}[x]\) with \(\deg a \leq \ell_2 - \ell_1 + j\) and for all monic \(b \in \mathbb{F}[x]\) with \(\deg b = j\); write \(f = \left[ f^{(1)} \ f^{(2)} \right] \). Check whether \(f^{(3)}G\) is a multiple of \(f^{(2)}\), where \(G\) is given by (12).
4. Whenever step 3) is successful, output all obtained quotient polynomials, i.e., polynomials \(m\) of the form \(m = -f^{(1)}G/f^{(2)}\). In case step 3) is not successful increase \(j\) by 1 and repeat step 3).

Again the Euclidean algorithm can be used to compute \(g_1\) and \(g_2\) for this, Algorithm 2 should be initialized by \(\Pi_\varphi\) and \(\mathcal{L}_\varphi\) instead of \(\Pi\) and \(\mathcal{L}\) and the stopping criterion (10) should be replaced by

\[\deg t_{j+1} \geq \deg h_{j+1},\]

instead of (10).

An alternative way to compute \(g_1\) and \(g_2\) is to employ an algorithm that processes the values of \(y_1,\ldots,y_{n-k}\) iteratively. For this, Algorithm 3 is modified into Algorithm 5 which essentially coincides with the well-known Welch-Berlekamp algorithm [28], see also [14], [15].

**4 Minimal list decoding through rational interpolation**

The most computationally intensive task in Algorithm 1 is Step 3. Recall that in Step 3, we need to determine all \(a\) and \(b\) of degree \(k_1 \leq \ell_2 - \ell_1 + j\) and \(k_2 = j\) such that \(f^{(1)}\) is a multiple of \(f^{(2)}\). A brute force approach may be to consider

\[f = \begin{bmatrix} f^{(1)} \\ f^{(2)} \end{bmatrix} = a \begin{bmatrix} g_1^{(1)} \\ g_1^{(2)} \end{bmatrix} + b \begin{bmatrix} g_2^{(1)} \\ g_2^{(2)} \end{bmatrix}\]

12
It follows from Theorem 3.2 that, in the context of Algorithm 1, which efficient polynomial time algorithms exist. We explain this approach in the following.

Then Step 3 of Algorithm 1 can be formulated as the following rational interpolation problem.

\[ f(x) \text{ has } t \text{ distinct roots.} \]

Therefore, an alternative approach to Step 3 is to determine all \( a \) and \( b \) of degree \( k_1 \leq t + k - \ell_1 - 1 \) and \( k_2 = t + k - \ell_2 - 1 \), respectively, such that

\[
  f^{(2)}(x) = a(x)g_1^{(2)}(x) + b(x)g_2^{(2)}(x)
\]

has \( t \) distinct roots. Now dividing both sides of (15) by \( g_1^{(2)}(x) \) we get

\[
  \frac{f^{(2)}(x)}{g_1^{(2)}(x)} = a(x) + b(x)\frac{g_2^{(2)}(x)}{g_1^{(2)}(x)}.
\]

Now let us define

\[
  z_i = -\frac{g_2^{(2)}(x_i)}{g_1^{(2)}(x_i)} \quad \text{for } i = 1, \ldots, n.
\]

Then Step 3 of Algorithm [4] can be formulated as the following rational interpolation problem.

**Rational Interpolation Problem:** Given \( n \) points \((x_1, z_1), (x_2, z_2) \cdots, (x_n, z_n)\) and a non-negative integer \( t \), determine all rational polynomials of the form \( z = a/b \), with \( a \) and \( b \) of degree \( k_1 \) and \( k_2 \), respectively, such that \( z \) passes through \( t \) of the \( n \) points \((x_1, z_1), (x_2, z_2) \cdots, (x_n, z_n)\).
This problem looks similar to the interpolation problem addressed by Guruswami and Sudan in [10]. However, it is complicated by the fact that now we look for a rational solution rather than a polynomial solution. Recently, this rational interpolation problem has been addressed by Wu in [29]. For the sake of completeness we briefly describe Wu’s formulation here.

4.1 Wu’s rational interpolation algorithm

In line with the Guruswami-Sudan approach, Wu’s algorithm first computes a bivariate polynomial \( Q(x, z) \), satisfying certain constraints, that passes through all the \( n \) points \((x_1, z_1), (x_2, z_2), \ldots, (x_n, z_n)\). Then the desired rational solutions \( z = \frac{a}{b} \) are obtained from the factorization of \( Q(x, z) \). Given the values of \( t, k_1, \) and \( k_2 \), let us determine the constraints that must be satisfied for the existence of such a \( Q(x, z) \).

Let us define the \((1, w)\) weighted degree of a bivariate polynomial \( Q(x, z) = \sum_{(i,j)\in I} a_{i,j} x^i z^j \) as

\[
\text{wdeg}_{1,w} Q(x, z) = \max_{(i,j)\in I} \{i + jw\}. \tag{17}
\]

Let \( w := k_1 - k_2, \rho := \text{wdeg}_{1,w} Q(x, z) \), and \( M := \text{wdeg}_{0,1} Q(x, z) \). Clearly, \( \text{deg}_{0,1} Q(x, z) \) is the \( z \)-degree of \( Q(x, z) \). Now if \( z = \frac{a}{b} \) passes through \( t \) points with multiplicity \( s \) then the polynomial \( b(x)^M Q(x, a(x)/b(x)) \) must have \( ts \) roots. On the other hand, \( b(x)z - a(x) \) will be a factor of \( Q(x, z) \) if \( b(x)^M Q(x, a(x)/b(x)) \) is identically zero. In turn, \( b(x)^M Q(x, a(x)/b(x)) \) will be identically zero if it has more roots than its degree. Now the degree of \( b(x)^M Q(x, a(x)/b(x)) \) is at most \( \rho + Mk_2 \).

Therefore, a necessary condition that must be satisfied is

\[
\rho + Mk_2 < ts. \tag{18}
\]

On the other hand, a necessary condition for the existence of \( Q(x, z) \) passing through the \( n \) points with multiplicity \( s \) is that its \((u, v)\)-th Hasse derivatives at all the \( n \) points are zero for all \( u + v \leq s \). Thus the requirement that \((x_i, z_i)\) be a zero of \( Q(x, z) \) with multiplicity \( s \), for all \( i = 1, 2, \ldots, n \), leads to \( N \) constraints in the form of \( N \) homogeneous equations where

\[
N = ns(s+1)/2. \tag{19}
\]

and unknown variables are the coefficients of \( Q(x, z) \). A nonzero solution to the system of homogeneous equations is guaranteed to exist if the number of equations is less than the number of unknowns. Now the number of coefficients in \( Q(x, z) \) with \( \text{wdeg}_{1,w} Q(x, z) = \rho \) and \( \text{wdeg}_{0,1} Q(x, z) = M \) is

\[
U = (\rho + 1)(M + 1) - \frac{w}{2} M(M + 1). \tag{20}
\]

Therefore, a sufficient condition for the existence of a \( Q(x, z) \), passing through all the \( n \) points with multiplicity \( s \), is

\[
(\rho + 1)(M + 1) - \frac{w}{2} M(M + 1) > \frac{ns(s+1)}{2}. \tag{21}
\]

Wu, in [29], has proposed suitable choices for the values of \( s, M, \) and \( \rho \) satisfying (18) and (21) as

\[
s = \frac{t(n - k + 1 - t)}{t^2 - n(2t - (n - k + 1))}, \tag{22}
\]

\[
M = \frac{st}{2t - (n - k + 1)}, \tag{23}
\]

\[
\rho = ts - Mk_2 - 1. \tag{24}
\]
For more details on Wu’s algorithm see [22]. It is worth noting that the multiplicity \( s \), computed using \([22]\), is not minimal. Although Wu suggested to first compute \( s \) according to \([22]\) and then greedily minimize it subject to a certain constraint, he did not give any explicit algorithm to compute the minimal value of \( s \). More importantly, given the minimum \( s \), the values of \( M \) and \( \rho \) computed in \([23]\) and \([24]\) are not necessarily optimal. In the next section, we present an algorithm that computes the minimum value of \( s \) as well as the associated optimal values of \( M \) and \( \rho \).

### 4.2 Optimizing the integer parameters

Given feasible values of \( s \), \( M \), and \( \rho \), the rational interpolation step involves two steps: (1) construction of \( Q(x, z) \) and (2) factorization of \( Q(x, z) \). The best known algorithm for the construction of the interpolating polynomial \( Q(x, z) \) is the Kötter algorithm \([11]\). The Kötter algorithm has a complexity of \( O(MN^2) \) \([12]\), where \( N \) is the number of constraints as defined in \([19]\). More precisely, it has memory complexity of \( O(MU) \) and time complexity of \( O(NMU) \) \([8]\), where \( U \) is the number of coefficients in \( Q(x, z) \) as defined in \([20]\) and \( M \) is the \( z \)-degree of the interpolating polynomial \( Q(x, z) \). On the other hand, the rational factorization step can be done in time \( O(n^{3/2}s^{7/2}) \) using Wu’s rational factorization procedure \([29]\). As analyzed in sub-section 4.4, it is the Kötter algorithm that dominates the overall memory and computational complexity of the proposed, as well as Wu’s, list decoding procedure. Therefore, to reduce the complexity of the Kötter algorithm, we take the following two step strategy. In the first step, we derive an explicit method to determine the minimum value of \( s \) for which there exist some \( M \) and \( \rho \) satisfying \([18]\) and \([21]\). Once the minimum multiplicity is determined, \( N \) becomes fixed. Then in the second step, we compute the optimal values of \( M \) and \( \rho \) such that \( MU \) is minimized.

The constraint \([18]\) can be geometrically interpreted as follows. Assume that \( t \) and \( s \) are fixed. With the requirement that all the values involved in \([18]\) are non-negative integers, all feasible values of \( \rho \) and \( M \) must be on or below the line \( L \) defined by the equation

\[
\rho + Mk_2 = ts - 1.
\]  

On the other hand, the constraint \([21]\) requires that all feasible values of \( \rho \) and \( M \) are above the curve \( C \) defined by the equation

\[
(\rho + 1)(M + 1) - \frac{w}{2}M(M + 1) = \frac{ns(s + 1)}{2}.
\]  

Therefore, a necessary condition for the existence of a feasible solution satisfying both the constraints \([18]\) and \([21]\) is that \( L \) intersects \( C \) at two different points \((M_1, \rho_1)\) and \((M_2, \rho_2)\) on the real plane. Now solving \([25]\) and \([26]\) for \( M \) we get

\[
M = \frac{(ts - k_0) \pm \sqrt{(ts - k_0)^2 - 4(N - ts)k_0}}{2k_0},
\]  

where \( k_0 = (k_1 + k_2)/2 \). According to Algorithm \([1]\) while correcting \( t = \ell_1 - k + 1 + j \) errors, we have \( k_1 = \ell_2 - \ell_1 + j \) and \( k_2 = j \). Using \( \ell_1 + \ell_2 = n + k - 1 \), we get \( k_0 = (t - t_0) \) where \( t_0 = d/2 \). Substituting \( k_0 = (t - t_0) \) in \([27]\) we get

\[
M = \frac{(ts - t + t_0) \pm \sqrt{(ts - t + t_0)^2 - 4(N - ts)(t - t_0)}}{2(t - t_0)}.
\]  

It follows from \([28]\) that the value of \( M \) and thus the choice of \( s \) is independent of \( k_1 \) and \( k_2 \). Now for a fixed \( s \), it can easily be verified if \( L \) and \( C \) intersect at two different points on the real plane by checking whether

\[
(ts - t + t_0)^2 > 4(N - ts)(t - t_0).
\]
According to (29) any feasible $s$ must satisfy the following inequality which was also derived in Wu [29]

$$s^2(t^2 - 2(t - t_0)n) - 2s(n - t)(t - t_0) + (t - t_0)^2 > 0. \tag{30}$$

This in turn implies that

$$s > \frac{(t - t_0)(n - t + \sqrt{n(n - d)})}{t^2 - 2n(t - t_0)}. \tag{31}$$

From (31) it also follows that a feasible value of $s$ will exists only if

$$t^2 - 2n(t - t_0) > 0, \tag{32}$$

which also leads to the same bound on the list decoding radius as derived in [10]

$$t < n - \sqrt{n(n - d)). \tag{33}$$

Also from (31) we get the lower bound on $s$ as

$$s_l = \left\lfloor \frac{(t - t_0)(n - t + \sqrt{n(n - d)})}{t^2 - 2n(t - t_0)} \right\rfloor + 1. \tag{34}$$

Moreover, an upper bound on $s$ was derived in [29] as

$$s_u = \left\lfloor \frac{t(2t_0 - t)}{t^2 - 2n(t - t_0)} \right\rfloor + 1. \tag{35}$$

Thus any $s$, such that $s_l \leq s \leq s_u$ will satisfy the condition (29). Now assume that for a particular $s$, the condition (29) is satisfied, i.e., $L$ and $C$ intersect at two different points $(M_1, \rho_1)$ and $(M_2, \rho_2)$
on the real plane. Without any loss of generality let us assume that \( M_1 < M_2 \). When \( L \) and \( C \) intersect at two different points on the real plane, there will exist a feasible solution if there is an integer \( M^* \) such that \( M_1 < M^* < M_2 \), i.e., if

\[
[M_1] + 1 < M_2.
\]  

(36)

Clearly if (36) is satisfied, then any \( M^* \in \left[ [M_1] + 1, [M_2] - 1 \right] \) is a feasible choice of \( M \). Now according to [20], for a feasible choice of \( M = M^* \), it is desirable to find the minimum value of \( \rho \) so that \( U \) is minimized. Let the line \( M = M^* \) intersect \( C \) and \( L \) at points \((M^*, \rho_l)\) and \((M^*, \rho_h)\) respectively. Since \( L \) intersects \( C \) from above, it must be the case that \( \rho_l < \rho_h \). Although \( \rho_h \) is a feasible choice for \( \rho \), as used by Wu, we choose the minimum possible value as

\[
\rho^* = [\rho_l] + 1.
\]  

(37)

We illustrate the method of computing the feasible values of the integer parameters, using a particular example, in Fig. 1.

Now to find the optimal value of \( M \) and \( \rho \) such that \( MU \) is minimized, we need to compute \( \rho^* \) and \( U^* \) for all \( M^* \in \left[ [M_1] + 1, [M_2] - 1 \right] \) and choose \( M^* \) and \( \rho^* \) that result in the minimum value of \( MU \). We summarize the above procedure in Algorithm 6 that computes the values of minimum multiplicity \( s_{\text{min}} \) and the associated optimal \( z \)-degree \( M_{\text{opt}} \) and weighted degree \( \rho_{\text{opt}} \).

**Algorithm 6** Compute optimal values of the integer parameters

**Input**: \( n, k, t, k_1, \) and \( k_2 \).

**Output**: Minimum multiplicity \( s_{\text{min}} \) and optimal \( z \)-degree \( M_{\text{opt}} \) and weighted degree \( \rho_{\text{opt}} \).

1. Compute \( w := k_1 - k_2, d := n - k + 1, t_0 := d/2 \).
2. Initialize \( s := \max(s_l = [(t - t_0)(n - t + \sqrt{n(n - d)})/(t^2 - 2n(t - t_0))] + 1, 1) \)
3. \( M_{\text{opt}} := \infty, \rho_{\text{opt}} := \infty, U_{\text{opt}} = \infty \)
4. **while** no feasible solution is found **do**
5. 1. Compute \( N := ns(s+1)/2 \).
6. 2. If \((ts - t + t_0)^2 > 4(N - ts)(t - t_0)\) then
7. 3. \((M_2, M_1) := ((ts - t + t_0) \pm \sqrt{(ts - t + t_0)^2 - 4(N - ts)(t - t_0)})/2(t - t_0)\)
8. 4. If \([M_1] + 1 < M_2\) then
9. 5. \(s_{\text{min}} := s\)
10. 6. For \( M = [M_1] + 1 \) to \([M_2] - 1\) do
11. 7. \( \rho := [N/M + 1] + w/2M - 1 + 1\)
12. 8. \( U := (\rho + 1)(M + 1) - w/2M(M + 1)\)
13. 9. If \( MU < M_{\text{opt}}U_{\text{opt}}\) then
14. 10. \( M_{\text{opt}} := M, \rho_{\text{opt}} := \rho, U_{\text{opt}} = U\)
15. 11. end if
16. 12. end for
17. 13. end if
18. 14. return \( s_{\text{min}}, M_{\text{opt}}, \rho_{\text{opt}} \)
19. 15. end if
20. 16. \( s := s + 1 \)
21. 17. end while

**Complexity of Algorithm 6** The complexity of the algorithm is dominated by the **while** loop and the **for** loop. Number of times the **while** loop is executed is bounded by \( s_{\text{min}} \). The **for** loop executes

\[
O(\sqrt{(ts - t + t_0)^2 - 4(N - ts)(t - t_0)/(t - t_0)}) = O(ts)
\]  

(38)
the minimal Gröbner basis in step 1 using Algorithm 3 takes \(O(n^2)\) operations. Algorithm 6 in step 2 takes \(O(ns^2)\) time. The Kötter algorithm used in step 3 involves \(O(MN^2) = O(Mn^s s^4)\) operations [12], where \(N\) is the number of constraints as defined in [19] and \(M\) is the \(z\)-degree of the interpolating polynomial \(Q(x, z)\). The rational factorization in step 4 can be done in time \(O(n^{3/2} s^{7/2})\) [29]. Thus the overall complexity of the proposed algorithm is \(O(MN^2)\). However, because of step 2, our list decoding algorithm optimizes \(MU\). Since, more precisely, the Kötter algorithm involves memory complexity of \(O(MU)\) and time complexity of \(O(NMU)\), our algorithm uses less memory as well as computation as compared to Wu’s method. The advantage of the proposed algorithm in terms of \(z\)-degree \(M\) and number of unknown coefficients \(U\) is illustrated in Example 4.1.

Example 4.1 Consider the \((127, 24)\) RS code defined over \(GF(2^7)\) with \(d = 104\). Consider correcting \(t = 64\) errors when \(k_1 = 15\) and \(k_2 = 9\). For this instance, Wu’s algorithm using [22] computes
\( s = 2, \) which is also the minimum multiplicity. Now Wu’s algorithm computes \( M = 5 \) and \( \rho = 72 \) using (23) and (24), respectively. With these values, Wu’s algorithm requires solving a system of \( N = 381 \) homogeneous equations involving \( U = 408 \) unknowns. In contrast, in our algorithm we find that when \( s_{\text{min}} = 2 \), the line \( L \) intersect the curve \( C \) at points \((3.3241,*)\) and \((6.3426,*)\). Now for the feasible values of \( M^* = 4, 5, 6 \), we get \( \rho^* = 88, 78, 72 \) and \( U^* = 385, 384, 385 \), respectively. Finally we get the optimal values as \( M_{\text{opt}} = 4 \) and \( \rho_{\text{opt}} = 88 \) with \( U_{\text{opt}} = 385 \).

5 Conclusions

In this paper we have taken a parametric approach to the problem of minimal list decoding. The proposed algorithms have error correcting radius \( L \), where \( L \) is the minimum of the Hamming distances between the received word and any codeword in \( C \). There are several important features of the approach. Firstly, the minimality of \( L \) ensures that all solutions correspond to valid codewords and therefore we do not need to check for validity. The parameterization can also be used for general list decoding, however, then a check on the validity of the corresponding codewords needs to be carried out. Secondly, upon computation of a solution of the rational interpolation problem or, equivalently, of an error locator polynomial, we do not need to determine the error magnitudes via Forney’s formula. Instead, solutions to the rational interpolation problem directly lead to message polynomials. Thirdly, we provide a geometric approach to optimize the integer parameters associated with the problem of rational interpolation. Since the interpolation step is the most computationally intensive task in list decoding, optimization of the integer parameters results in less computational as well as memory requirements. Finally, by using re-encoding as in sub-section 3.3, the approach lends itself well to the type of distributed source coding (DSC) proposed in [3].

Acknowledgment

We thank Nikeeth Venkatraman Ramanathan for helping in implementing the computational examples.

References

[1] W. W. Adams and P. Loustaunau. An introduction to Gröbner Bases, volume 3 of Graduate Stud. Math. American Mathematical Society, 1994.

[2] M. Alekhnovich. Linear diophantine equations over polynomials and soft decoding of Reed-Solomon codes,, IEEE Trans. Inf. Th., 51(7):2257–2265, 2005.

[3] M. Ali and M. Kuijper. Source coding with side information using list decoding. In Proceedings IEEE International Symposium in Information Theory, pages 91–95, Austin, Texas, 2010.

[4] E.R. Berlekamp. Algebraic Coding Theory. McGraw-Hill, New York, 1968.

[5] G.D. Forney. Convolutional codes I: Algebraic structure. IEEE Trans. Inf. Th., 16:720–738, 1970. correction, vol. IT-17, p.360, 1971.

[6] G.D. Forney, Jr. Minimal bases of rational vector spaces, with applications to multivariable linear systems. SIAM J. Control, 13:493–520, 1975.

[7] S. Gao. A new algorithm for decoding of Reed-Solomon codes. In V. K. Bhargava, H. V. Poor, V. Tarokh, and S. Yoon, editors, Communications information and network security, pages 55–68. Kluwer, 2003.
[8] W. J. Gross, F. R. Kschischang, R. Kötter, and P. g. Gulak. Simulation results for algebraic soft-decision decoding of Reed-Solomon codes. In Proceedings of the 21st Biennial Symposium on Communications, pages 356–360, Kingston, Ontario, Canada, 2002.

[9] V. Guruswami. List decoding of error-correcting codes. volume 3282 of Lecture Notes in Computer Science. Springer, 2004.

[10] V. Guruswami and M. Sudan. Improved decoding of Reed-Solomon and algebraic-geometric codes. IEEE Trans. Inf. Th., 45:1757–1768, 1999.

[11] R. Kötter. Fast generalized minimum distance decoding of algebraic geometry and reed-solomon codes. IEEE Trans. Inf. Th., 42(3):721–737, 1996.

[12] R. Kötter, J. Ma, and A. Vardy. The re-encoding transformation in algebraic list-decoding of Reed-Solomon codes. IEEE Trans. Inf. Th. submitted (April 2010). Available: http://arxiv.org/abs/1005.5734.

[13] R. Kötter and A. Vardy. A complexity reducing transformation in algebraic list decoding of Reed-Solomon codes. In Proceedings ITW (Paris, France), 2003.

[14] M. Kuijper. A system-theoretic derivation of the Welch-Berlekamp algorithm. In Proceedings 2000 IEEE International Symposium in Information Theory, page 418, Sorrento, Italy, 2000.

[15] M. Kuijper. Algorithms for decoding and interpolation. In Brian Marcus and Joachim Rosenthal, editors, Codes, Systems, and Graphical Models, volume 123 of The IMA Volumes in Mathematics and its Applications, pages 265–282. Springer-Verlag, 2001.

[16] M. Kuijper and J.W. Polderman. Reed-Solomon list decoding from a system theoretic perspective. IEEE Trans. Inf. Th., IT-50:259–271, 2004.

[17] M. Kuijper and K. Schindelar. The predictable leading monomial property for polynomial vectors over a ring. In Proceedings IEEE International Symposium in Information Theory, pages 1133–1137, Austin, Texas, 2010.

[18] M. Kuijper and K. Schindelar. Minimal Gröbner bases and the predictable leading monomial property. Linear Alg. Appl., 434:104–116, 2011.

[19] K. Lee and M.E. O’Sullivan. List decoding of Reed-Solomon codes from a Gröbner basis perspective. J. Symbolic Comput., 43:645–658, 2008.

[20] J. L. Massey. Shift-register synthesis and BCH decoding. IEEE Trans. Inf. Th., IT-15:122–127, 1969.

[21] R. J. McEliece. The Guruswami-Sudan decoding algorithm for Reed-Solomon codes. Technical Report 42-153, Jet Propulsion Laboratory, California Institute of Technology, Pasadena, CA, May 2003.

[22] J. S. R. Nielsen. List-decoding of error correcting codes. Master’s thesis, Department of Mathematics, Technical University of Denmark, Denmark, 2010.

[23] R. Nielsen and T. Høholdt. Decoding Reed-Solomon codes beyond half the minimum distance. In J. Buchmann, T. Hoeholdt, T. Stichtenoth, and H. Tapia-Recillas, editors, Coding Theory, Cryptography and Related Areas, pages 221–236, Berlin, 2000. Springer-Verlag.

[24] H. O’Keeffe and P. Fitzpatrick. Grobner basis approach to list decoding of algebraic geometry codes. Applicable Algebra in Engineering, Communication and Computing, 8(5):445–466, 2007.
[25] M. Sudan. Decoding of Reed-Solomon codes beyond the error correction bound. *J. Compl.*, 13:180–193, 1997.

[26] Y. Sugiyama, M. Kasahara, S. Hirasawa, and T. Namekawa. A method for solving key equation for decoding goppa codes. *Information and Control*, 27:87–99, 1975.

[27] P. V. Trifonov. Interpolation in list decoding of Reed-Solomon codes. *Problems of Information Transmission*, 43(3):190–198, 2007.

[28] L. Welch and E. R. Berlekamp. Error correction of algebraic block code. *US Patent 4 633 470*, Dec 1986.

[29] Y. Wu. New list decoding algorithms for Reed-Solomon and BCH codes. *IEEE Trans. Inf. Th.*, 54:3611–3630, 2008.