Performance of Sphere Decoding of Block Codes

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Abstract

A sphere decoder searches for the closest lattice point within a certain search radius. The search radius provides a tradeoff between performance and complexity. In this work, we focus on analyzing the performance of sphere decoding of block codes. The performance of soft-decision sphere decoding of linear block codes on AWGN channels and a variety of modulation schemes is studied. Hard-decision sphere decoding on binary and $q$-ary symmetric channels is analyzed. We show how sphere decoding of Reed Solomon codes can out perform popular decoding algorithms such as the hard decision Guruswami-Sudan algorithm and algebraic soft decision decoding. An upper bound on the performance of maximum likelihood decoding of Reed Solomon codes over $q$-ary symmetric channels is derived and used in the analysis. We then discuss sphere decoding of general block codes or lattices with arbitrary modulation schemes. The tradeoff between the performance and complexity of a sphere decoder is also discussed.

I. INTRODUCTION

Maximum likelihood (ML) decoding of linear block codes is known to be NP-hard [1]. A decoder that utilizes the soft output from the channel directly is called a soft-decision (SD) decoder. On the other hand, if hard decisions are made on the received bits before decoding, then such a decoder is called a hard-decision (HD) decoder. The optimum decoder is the corresponding
HD or SD maximum likelihood (ML) decoder. Berlekamp’s tangential bound is a tighter than the union bound for additive white Gaussian noise (AWGN) channels [2]. Poltyrev derived tight upper bounds on the performance of maximum likelihood decoding of linear block codes over AWGN channels and binary symmetric (BSC) channels. Bounds based on typical pairs decoding were derived by Aji et. al [3]. Other bounds such as the Divsalar simple bound and the variations on the Gallager bounds are tight for AWGN and fading channels [4], [5]. For a broad survey on bounds on the maximum likelihood decoding of linear codes, see [6].

Fincke and Pohst (FP) [7] described a sphere decoder algorithm which finds the closest lattice point without actually searching all the lattice points. A fast variation of it was given by Schnorr and Euchner [8]. Other efficient closest point search algorithms exist (for a survey see [9]). The sphere decoder algorithm was proposed for decoding lattice codes [10] and for detection in multiple antenna wireless systems [11], [12]. Vikalo and Hassibi proposed HD and SD sphere decoders for joint detection and decoding of linear block codes [13] [14]. On the other hand, one can think of a sphere decoder in a broader sense as any algorithm that returns the closest lattice point to the received word if it exists within a predetermined search radius. By this definition of a sphere decoder, the Berlekamp-Massey algorithm can be considered as a sphere decoder for Reed Solomon (RS) codes with a search radius equal to half the minimum distance of the code. Similarly, the algorithm recently proposed by Guruswami and Sudan for decoding RS codes is an algebraic sphere decoder whose search radius can be larger than half the minimum distance of the code [15].

There has been significant amount of research dedicated to the design of sphere decoders with smaller complexities, complexity analysis of sphere decoders and the application of sphere decoders to various settings and communication systems. However, little research focused on the performance analysis of sphere decoders. This paper sets down a framework for the analysis of the performance of sphere decoding of block codes over a variety of channels with various modulation schemes.

In this paper, we study the performance of soft decision sphere decoding of linear block codes on channels with additive white Gaussian noise and various modulation schemes as BPSK, M-PSK and QAM [16]. This is done in sections II and III respectively. Bounds on the performance of hard decision sphere decoding on BSC are derived in section IV. The application of these bounds to the binary image of Reed Solomon codes is also investigated. We then, in section
V derive bounds on the maximum likelihood performance of \( q \)-ary linear codes, such as Reed Solomon codes, over \( q \)-ary symmetric channels. This bound becomes handy when analyzing the performance of sphere decoding of Reed Solomon codes on \( q \)-ary symmetric channels. Furthermore, we show, in section III, how one can analyze the performance of a soft decision sphere decoder of a general block code with a general modulation scheme. In many settings, we support our analytic bounds by comparing them to numerical simulations. The tradeoff between performance and complexity is discussed in (VI). Finally, we conclude our work in section VII.

II. UPPER BOUNDS ON THE PERFORMANCE OF SOFT DECISION SPHERE DECODING OF BPSK AND M-PSK MODULATED BLOCK CODES.

In this section, we consider a sphere decoder when the modulation is binary or \( M \)-ary phase shift keying (PSK) \[16\]. Each transmitted codeword in the code has the same energy when mapped to the PSK constellation. For the case of MPSK modulation, complex sphere decoding algorithms which solve the closest point search problem were developed in \[17\].

A. Preliminaries

We will introduce some notation, so the bounds derived here are readily applicable for both \( M \)-ary and binary phase shift keying (PSK) modulation. We assume that \( C \) is an \((n, k)\) linear code. Each codeword of length \( n \) will be mapped to a word of \( M \)-PSK symbols. The number of channel symbols will be denoted by \( n_c \). If the code \( C \) is binary and of length \( n \), then \( n_c = \lceil n / \log_2(M) \rceil \).

For BPSK, \( n_c = n \). Note that the original code need not be binary. For example, an Reed Solomon (RS) code defined over \( \mathbb{F}_{2^m} \) could be mapped directly to an \( 2^m \)-ary PSK constellation by a one-to-one mapping from the symbols in \( \mathbb{F}_{2^m} \) to the \( 2^m \) points in the PSK constellation.

For PSK signaling, the code will have the property that all codewords are of equal energy and lie on a sphere of radius \( \sqrt{n_c} \) from the origin of space. Let \( n_d \) denote the dimension of the considered space (noise). For the case of BPSK modulation, the dimension of the Hamming space is the same as the number of channel symbols (bits) \( n_d = n_c \). On the other hand, for MPSK signaling, \( M > 2 \), each complex channel symbol has a real and an imaginary component. Thus the noise has \( 2n_c \) independent components and the dimension of the space is \( n_d = 2n_c \).

Assuming that a codeword \( c \in C \) is transmitted over a binary input AWGN channel, the received word is \( y = x + z \), where \( x = M(c) \) and \( M(c) \) is the mapping of the codeword \( c \).
under PSK modulation, i.e., for BPSK modulation $M(c) \triangleq 1 - 2c$. The additive white Gaussian noise (AWGN) is denoted by $z = [z_i]_{i=1}^{n_d}$ with variance $\sigma^2$. Let $G_w$ be the number of codewords which (after mapping) are at an Euclidian distance $\delta_w$ from each other. Note that for the case of BPSK modulation and a binary code $C$, the space is a Hamming space and the Euclidean distance is directly related to the Hamming distance, $\delta_w = 2^{\sqrt{w}}$, where $w$ is the Hamming distance. QPSK modulation and Gray encoding also result in a Hamming space [16] by $\delta_w = \sqrt{2^{w}}$, where $w$ is the (binary) Hamming distance between the codewords. For simplicity in the following analysis, we will assume that the modulated code is linear and the space is a Hamming space.

**B. Analysis of Soft Decision Sphere Decoding**

A soft-decision sphere decoder with an Euclidean radius $D$, denoted by SSD($D$), solves the following optimization problem,

$$\hat{c} = \arg \min_{v \in C} \|y - M(v)\|^2$$

subject to $\|y - M(v)\|^2 \leq D^2$,

where $\|x\|$ is the Euclidean norm of $x$. Such decoders include list-decoders that list all codewords whose modulated image is within an Euclidean distance $D$ from the received vector $y$ and choose the closest one. If no such codeword exists, a decoding failure is signaled. A decoding error is signaled if the decoded codeword is not the transmitted code word.

Let $E_D$ denote the event of error or failure of SSD($D$), then the error plus failure probability, $P(E_D)$ \(^1\) is

$$P(E_D) = P(E_D|E_{ML})P(E_{ML}) + P(E_D|S_{ML})P(S_{ML}),$$

where $E_{ML}$ and $S_{ML}$ denote the events of an ML error and an ML success respectively. Let $\epsilon = \|y - M(c)\|$, then an ML error results if there exists another codeword $\hat{c} \in C$ such that $\|y - M(\hat{c})\| \leq \epsilon$. Since limiting the decoding radius to $D$ will not do better than ML decoding, then $P(E_D|E_{ML}) = 1$. By observing that $P(S_{ML}) \leq 1$, it follows that an upper bound on the decoding performance is

$$P(E_D) \leq P(E_{ML}) + P(E_D|S_{ML}).$$

\(^1\)Throughout this paper, $P(X)$ will denote the probability that the event $X$ occurs.
Let $\Omega_D$ be the Euclidean sphere of radius $D$ centered around the transmitted codeword in the $n_d$ dimensional space. The probability that the added white Gaussian noise will not lie in the sphere $\Omega_D$ is

$$P(z \notin \Omega_D) = P(\chi_{n_d} > D^2) = 1 - \Gamma_r(n_d/2, D^2/2\sigma^2)$$

(4)

where $\chi_n = \sum_{i=1}^n z_i^2$ is a Chi-squared distributed random variable with $n$ degrees of freedom. Let $\Gamma(x)$ denote the Gamma function, then the cumulative distribution function (CDF) of $\chi_v$ is given by the regularized Gamma function $\Gamma_r$ [18],

$$\Gamma_r(v/2, w/2) = \begin{cases} \int_0^w \frac{t^{v/2-1}e^{-t/2}}{2^{v/2}\Gamma(v/2)} dt, & w \geq 0; \\ 0, & w < 0. \end{cases}$$

(5)

**Lemma 1:** A lower bound on $P(\mathcal{E}_D)$ is $P(\mathcal{E}_D) \geq P(z \notin \Omega_D)$.

**Proof:** The sphere decoder error plus failure probability could be written as

$$P(\mathcal{E}_D) = P(\mathcal{E}_D|z \in \Omega_D)P(z \in \Omega_D) + P(\mathcal{E}_D|z \notin \Omega_D)P(z \notin \Omega_D)$$

$$\geq P(\mathcal{E}_D|z \notin \Omega_D)P(z \notin \Omega_D)$$

$$= P(z \notin \Omega_D),$$

where the last inequality is because $P(\mathcal{E}_D|z \notin \Omega_D) = 1$ which follows from the definition of the sphere decoder (1).

Define $\bar{P}(\mathcal{E}_{ML})$ to be an upper bound on the SD-ML decoder error probability, then we have the following lemma,

**Lemma 2:** $P(\mathcal{E}_D) \leq \bar{P}(\mathcal{E}_{ML}) + P(z \notin \Omega_D)$.

**Proof:** Given an ML success, $\mathcal{E}_D$ will only be due to failures of the SSD(D) decoder, i.e.,

$$P(\mathcal{E}_D|\mathcal{S}_{ML}) = P(\|y - M(c)\| > D) = P(z \notin \Omega_D),$$

where the last equality follows from the linearity of the code and without loss of generality one could assume that the all zero codeword was transmitted. By definition, $P(\mathcal{E}_{ML}) \leq \bar{P}(\mathcal{E}_{ML})$. By substituting in (3) we are done.

Lemma 2 provides a way to bound the performance of sphere decoding of linear block codes on a variety of channels where additive white Gaussian noise is added and for a variety of modulation schemes. For example, it can be used in conjunction with the Divsalar bound [4] to give an upper bound on the performance of sphere decoding of linear block codes over
independent Rayleigh fading channels. If $P(\mathcal{E}_{ML})$ is the union upper bound on the codeword error probability [16, Ch.8] for BPSK modulation on an AWGN channel, then

$$P(\mathcal{E}_D) \leq \sum_{w \geq 1} G_w Q(\sqrt{2\gamma R w}) + P(z \notin \Omega_D),$$

(6)

where $G_w$ is the number of codewords with (binary) Hamming weight $w$, $\gamma$ is the bit signal to noise ratio (SNR) and $R$ is the rate of the code.

Lemma 1 implies that one could obtain a tighter upper bound on $P(\mathcal{E}_D)$ by tightening the bound on the ML error probability, $\bar{P}(\mathcal{E}_{ML})$. Shannon’s sphere packing bound [19] is a lower bound on the error probability where Shannon showed that the Voronoi region of a codeword can be bounded by a right circular $n_d$-dimensional cone with the codeword on its axis. Poltyrev’s tangential sphere bound (TSB) is one of the tightest bounds on the ML performance of soft decision decoding of linear codes on AWGN channels with BPSK or MPSK modulation [20], [21] and is calculated by,

$$P(\mathcal{E}_{ML}) \leq \min_{\theta} \{ P(\mathcal{E}_{ML}, z \in V_{\theta}) + P(z \notin V_{\theta}) \},$$

(7)

where $V_{\theta}$ is an $n_d$-dimensional right circular cone with a half angle $\theta$ whose central line passes through the transmitted codeword and whose apex is at an Euclidean distance $\sqrt{n_c}$ from the transmitted codeword. Let the minimum of the optimization problem in (7) be achieved at $\theta = \phi$, then by Lem. 2 we have the following upper bound (which is tighter than (6) in case of BPSK)

$$P(\mathcal{E}_D) \leq P(\mathcal{E}_{ML}, z \in V_{\phi}) + P(z \notin V_{\phi}) + P(z \notin \Omega_D).$$

(8)

For the TSB, the optimum angle $\phi$ is related to the radius $\sqrt{r_{\phi}}$ (see Fig. 1 or Fig. 2) by $\tan(\phi) = \sqrt{r_{\phi}/n_c}$, such that $r_{\phi}$ is the root of this equation [21]

$$\sum_{\delta_b > 0} G'_b(r_o) \int_{0}^{\theta_b(r_o)} \sin^{n_d-3}(\vartheta) d\vartheta = \frac{\sqrt{\pi}\Gamma\left(\frac{n_d-2}{2}\right)}{\Gamma\left(\frac{n_d-1}{2}\right)}$$

(9)

when solved for $r_o$, where $\theta_b(r_o) \triangleq \cos^{-1}\left(\frac{\delta_b/2}{\sqrt{r_o(1-\delta_b^2/4n_c)}}\right)$ and

$$G'_b(r_o) = \begin{cases} G_b, & \delta_b^2/4 < r_o(1-\delta_b^2/4n_c); \\ 0, & \text{otherwise.} \end{cases}$$

(10)

Let $z_1$ be the component of the noise along the central axis of the cone with a probability distribution function (PDF) $\mathcal{N}(z_1) = \frac{1}{\sqrt{2\pi}\sigma} e^{-z_1^2/2\sigma^2}$ and $z_2$ be the noise component orthogonal...
to \( z_1 \). Define \( \beta_{z_1}(w) \triangleq \frac{\sqrt{w - z_1}}{\sqrt{w_1}^{\frac{1}{2}}} \) and \( r_{z_1}(\phi) \triangleq \sqrt{\phi} \left( 1 - \frac{z_1}{\sqrt{n_c}} \right) \), then the ML error probability given that the noise \( z \) is in the cone \( V_\phi \) is [20]

\[
P(E_{ML}, z \in V_\phi) = \int_{-\infty}^{\infty} N(z_1) \left[ \sum_{b>0} G'_b(r_\phi) \int_{\beta_{z_1}(b)}^{r_{z_1}(\phi)} N(z_2) \Gamma_r \left( \frac{n_d - 2 r_{z_1}(\phi)^2 - z_2^2}{2\sigma^2} \right) dz_2 \right] dz_1.
\]

(11)

C. A Tight Upper Bound

We observe that instead of directly substituting the TSB of (7) for \( \bar{P}(E_{ML}) \) in Lem. 2 as we did in (8), one can find an upper bound which is tighter than (8) by noticing that the events \( \{ z \notin V_\theta \} \) and \( \{ z \notin \Omega_D \} \) are not in general mutually exclusive.

Lemma 3: \( P(E_D) \) is upper bounded by

\[
P(E_D) \leq P(E_{ML}, z \in V_\theta) + P(z \notin \Omega_D) + P(\{ z \notin V_\theta \} \cap \{ z \in \Omega_D \}).
\]

Proof: Using Bayes’ rule and defining the region \( \Lambda(\theta, D) \triangleq \{ V_\theta \cap \Omega_D \} \) we get

\[
P(E_D) \leq \min_{\theta} \{ P(E_D|z \in \Lambda(\theta, D))P(z \in \Lambda(\theta, D)) + P(E_D|z \notin \Lambda(\theta, D))P(z \notin \Lambda(\theta, D)) \}. \tag{12}
\]

From the definition of \( \Lambda(\theta, D) \), it follows that \( P(E_D, z \in \Lambda(\theta, D)) = P(E_{ML}, z \in \Lambda(\theta, D)) \leq P(E_{ML}, z \in V_\theta) \), where the last inequality follows from that \( \Lambda(\theta, D) \subseteq V_\theta \). Using \( P(E_D|z \notin \Lambda(\theta, D)) \leq 1 \), it follows that

\[
P(E_D) \leq \min_{\theta} \{ P(E_{ML}, z \in V_\theta) + P(z \notin \Lambda(\theta, D)) \}
\]

\[
\leq P(E_{ML}, z \in V_\theta) + P(z \notin \{ V_\theta \cap \Omega_D \}). \tag{13}
\]

The last inequality is due to the observation that \( \phi \) does not necessarily minimize (13). By de Morgan’s law, \( \{ V_\phi \cap \Omega_D \}^c = \{ \Omega_D \}^c \cup \{ V_\phi \}^c \cap \Omega_D \}, \{ . \}^c \) is the complement of \{ . \}. □

We consider two cases:

Case A: The sphere \( \Omega_D \) lies totally inside the cone \( V_\phi \). (See Fig. 1). This case is equivalent to the event \( \Delta \triangleq \{ D \leq D_\phi \} \), where

\[
D_\phi = \sqrt{n_c} \sin(\phi), \tag{14}
\]
and will be called the critical decoding radius. It follows that \( P(\{z \notin V_o\} \cap \{z \in \Omega_D\}) = 0 \), which could be substituted in Lem. 2. Furthermore, since \( \Lambda(\theta, D) = \Omega_D \), it follows from (12) that a tighter upper bound is

\[
P(\mathcal{E}_D|\hat{A}) \leq P(\mathcal{E}_{ML}, z \in \Omega_D) + P(z \notin \Omega_D).
\]

(15)

The joint probability of the added noise falling inside a sphere of Euclidean radius \( D \) and an ML error could be expressed as

\[
P(\mathcal{E}_{ML}, z \in \Omega_D) = \sum_{0 < \phi_D < D} G_b \int_{b_D/2}^{D} N(z_o) \Gamma_r \left( \frac{n_d - 1}{2}, \frac{D^2 - z_o^2}{2\sigma^2} \right) dz_o.
\]

(16)

Let \( \varphi \) be the half angle at which the cone \( V_{\varphi} \) is tangential to the sphere \( \Omega_D \), \( \varphi = \sin^{-1}(D/\sqrt{n}) \) (see Fig. 1), then another tight upper bound is

\[
P(\mathcal{E}_D|\hat{A}) \leq P(\mathcal{E}_{ML}, z \in V_{\varphi}) + P(z \notin \Omega_D).
\]

(17)

Theoretically, it is clear that the bound of (15) is tighter than that of (17), but numerically they are almost equivalent, since the integration over the region \( \Omega_D \cap V_{\varphi} \) is negligible. Note that \( P(\mathcal{E}_{ML}, z \in V_{\varphi}) \) is easily calculated using equation (11) where \( \tan(\varphi) = \sqrt{r_{\varphi}/n_c} \) and \( r_{z_1}(\varphi) = T_{\varphi} \left( 1 - \frac{z_1}{\sqrt{n_c}} \right) \).

\( \square \)

\textbf{Case B: The sphere } \Omega_D \textbf{ intersects the cone } V_o. \textbf{ (see Fig. 2).} We have two cases depending on the position of the apex of the cone. The first is when the apex of the cone does not lie in the sphere, \( \sqrt{n_c} \sin(\phi) < D < \sqrt{n_c} \) (see Fig. 2a) and the second is when the apex lies in the sphere, \( D \geq \sqrt{n_c} \) (see Fig. 2b). In both cases the following analysis holds. Let the origin, \( O \), of the \( n_d \)-dimensional space be at the transmitted codeword which is also the center of \( \Omega_D \). Since the cone and the sphere are symmetrical around the central axis, we project on a two dimensional plane as in Fig. 2. The radial component of the noise (along the axis of the cone) is \( z_1 \). The altitudes \( y_a(\phi) \) and \( y_b(\phi) \) at which the (double) cone intersects the sphere are found by substituting the line equation \( P = P_1 + U(P_2 - P_1) \), where \( P = (x, y) \), \( P_1 = (0, \sqrt{n_c}) \) and \( P_2 = (2\sqrt{n_c} \tan(\phi), -\sqrt{n_c}) \) into the quadratic equation of the sphere. It follows that \( y_{a,b}(\phi) = \sqrt{n_c}(1 - 2U_{a,b}(\phi, D)) \), where

\[
U_{a,b}(\theta, D) = \frac{4n_c \pm \sqrt{16n_c^2 - 16n_c \sec^2(\theta)(n_c - D^2)}}{8n_c \sec^2(\theta)}.
\]

It is easy to check that at \( D = \sqrt{n_c}, u_b = 0 \) and \( y_b \) is at the apex of \( V_o \). If \( D > \sqrt{n_c} \) then the intersection at \( y_b \) is in the lower nappe of the cone. It is also observed that \( V_o \) and \( \Omega_D \) do not
intersect \( (\Omega_D \subset V_\phi) \) if \( 16n_c^2 < 16n_c \sec^2(\phi)(n_c - D^2) \) or equivalently \( D < \sqrt{n_c \sin(\phi)} \) which is Case A.

Define \( B \) to be the event \( B \Delta \{ D > \sqrt{n_c \sin(\phi)} \} \), \( f_{n-1}(t) \) to be the PDF of \( \chi_{n-1} = \sum_{i=2}^{n} z_i^2 \), and \( \omega^2 = D^2 - z_1^2 \) (see Fig. 2). From Lem. 3, the error probability is upper bounded by

\[
P(\mathcal{E}_D|B) \leq P(\mathcal{E}_{ML}, z \in V_\phi) + P(z \notin \Omega_D) + P(\{z \notin V_\phi\} \cap \{z \in \Omega_D\}|B),
\]

where by Fig. 2

\[
P(\{z \notin V_\phi\} \cap \{z \in \Omega_D\}|B) = \int_{y_{\mu}(\phi)}^{y_{\nu}(\phi)} N(z_1) \int_{r_{21}(\phi)}^{\omega_1^2} f_{n_d-1}(t) dt dz_1.
\]

The tight upper bound is summarized in this theorem,

**Theorem 4:** The performance of soft decision sphere decoding with an Euclidean decoding radius \( D \) of a linear code with (Euclidean) weight spectrum \( G_b \) on an AWGN channel with noise variance \( \sigma^2 \) and (binary or M-ary) PSK modulation is upper bounded by:

\[
P(\mathcal{E}_D) \leq \begin{cases} 
\sum_{0 < \Phi < D} G_b \int_{\frac{D}{\sqrt{2\pi\sigma^2}}} e^{-z_1^2/(2\sigma^2)} \Gamma_r \left( \frac{n_d-1}{2}, \frac{D^2-z_1^2}{2\sigma^2} \right) dz_1 \\
+1 - \Gamma_r(n_d/2, D^2/2\sigma^2), & D \leq \sqrt{n_c \sin(\phi)}; \\
\int_{-\infty}^{\infty} N(z_1) \left[ \sum_{\delta_b > 0} G_b'(r_\phi) \int_{r_{21}(\phi)}^{r_{21}(\phi)} N(z_2) \Gamma_r \left( \frac{n_d-2}{2}, \frac{r_{21}(\phi)-z_1^2}{2\sigma^2} \right) dz_2 \right] dz_1 \\
+1 - \Gamma_r(n_d/2, D^2/2\sigma^2) \\
+ \int_{y_{\mu}(\phi)}^{y_{\nu}(\phi)} \Gamma_r \left( \frac{n_d-1}{2}, \frac{\omega_1^2}{2\sigma^2} \right) - \Gamma_r \left( \frac{n_d-1}{2}, \frac{r_{21}(\phi)}{2\sigma^2} \right), & D > \sqrt{n_c \sin(\phi)}, \end{cases}
\]

where \( \phi \) is the half angle of the cone \( V_\phi \) and is given by (9).

Following the proof of Lemma 3, the error plus failure probability of \( SSD(D) \) is upper bounded by

\[
P(\mathcal{E}_D) \leq P(\mathcal{E}_D, z \in \Lambda(\phi, D)) + P(z \notin \Lambda(\phi, D)).
\]

From the previous arguments in Cases A and B, the following theorem provides a slightly tighter upper bound than that of the previous theorem.

**Theorem 5:** The performance of \( SSD(D) \) for BPSK or MPSK modulation is upper bounded by

\[
P(\mathcal{E}_D) \leq \begin{cases} 
P(\mathcal{E}_{ML}, z \in \Omega_D) + P(z \notin \Omega_D), & D \leq \sqrt{n_c \sin(\phi)}; \\
P(\mathcal{E}_{ML}, z \in \Lambda(\phi, D)) + P(z \notin \Omega_D) + P(\{z \notin V_\phi\} \cap \{z \in \Omega_D\}), & D > \sqrt{n_c \sin(\phi)}. \end{cases}
\]
Observe that the difference from Theorem 4 is that the term \( P(E_{ML}, z \in \Lambda(\phi, D)) \) was upper bounded by \( P(E_{ML}, z \in V(\phi)) \) in Theorem 4. Consider a codeword at a distance \( \delta_w \), then the half angle of the cone bisecting this distance is \( \theta_w = \sin^{-1}(\delta_w/2\sqrt{n_c}) \) (c.f. Fig. 2). This cone will intersect the sphere \( \Omega_D \) at altitudes \( x_a(w) \) and \( x_b(w) \) given by \( x_{a,b}(w) = \sqrt{n_c}(1-2U_{a,b}(\theta_w, D)) \).

Now define the integrals

\[
\mathcal{I}(\gamma, w, z_1) \triangleq \mathcal{N}(z_1) \int_{\beta_{z_1}(w)}^{\gamma} \mathcal{N}(z_2) \Gamma_r \left( \frac{n_d - 2}{2}, \frac{\gamma^2 - z_2^2}{2\sigma^2} \right) dz_2,
\]

\[
\mathcal{I}_2(w) = \int_{x_{a}(w)}^{y_{a}(\phi)} \mathcal{I}(\omega_{z_1}, w, z_1) dz_1 + \int_{y_{a}(\phi)}^{y_{b}(\phi)} \mathcal{I}(r_{z_1}(\phi), w, z_1) dz_1 + \int_{x_{b}(w)}^{x_{b}(w)} \mathcal{I}(\omega_{z_1}, w, z_1) dz_1. 
\]

Taking the union over all codewords with non-zero Euclidean weights such that \( \theta_w < \phi \), it follows that for \( D > \sqrt{n_c \sin(\phi)} \),

\[
P(E_{ML}, z \in \Lambda(\phi, D)) = \sum_{\delta_{b}>0} G'_b(r_{\phi}) \mathcal{I}_2(w).
\]

### D. A Note on Reed-Solomon Codes

Consider the case when the binary image of an Reed-Solomon (RS) code, defined over \( \mathbb{F}_{2^m} \), is transmitted over an AWGN channel and the decoder is either a HD or SD sphere decoder. Tight upper bounds on the performance of HD and SD maximum likelihood decoding of the binary images of RS codes were developed by El-Khamy and McEliece [22] by averaging over all possible binary representations of the RS code. We use the same technique here to analyze the performance of the sphere decoders, where the average binary weight enumerator of the RS code (see [22]) is used as the weight spectrum \( G_b \) of the binary linear code.

### E. Numerical Results

In Fig. 3, we show how the bounds derived for M-ary modulated spherical codes are tight. The simulation curves and the analytical bounds will be labeled by ‘sim’ and ‘bnd’ respectively. A codeword in the \((24,12)\) Golay code is mapped into 12 QPSK symbols and transmitted over an AWGN channel. As observed, the simulated performance of the ML decoder and the SD sphere decoder [13] are tightly bounded by the bounds given in this section. The critical decoding radius in the \(2 \times 12\) dimensional space is \( D_\phi = 2.667 \).

In Fig. 4, the performance of SD sphere decoding of the binary image of the \((15,11)\) RS code, BPSK modulated over an AWGN channel, is investigated. The ML performance is simulated
by means of the MAP decoder, and it is observed that the averaged ML bound is tight [22]. We simulated the performance of SD sphere decoding when the decoding radius was 3 and 3.5 respectively. Our analytical bounds almost overlapped with the simulations. The critical decoding radius is \(D_\phi = 4.588\). A decoder with an Euclidean decoding radius of 5 has a near ML performance at an SNR of 5 dB. For reference purposes, we plot the performance of the hard-decision Berlekamp-Massey (BM) decoder and the algebraic soft decision decoder by Koetter and Vardy [23]. It is worth noting that algebraic soft decoding can also achieve near ML performance [24], [25].

III. SPHERE DECODING OF LATTICES

In this section, we consider the case of soft decision sphere decoding of a general lattice or code \(C\). In contrast to the case of section II the code is not constrained to be a linear code and the transmitted codewords are not constrained to have a fixed energy. The channel symbols of a transmitted codeword are also not required to have the same energy. Define \(G_w(i)\) to be the number of mapped codewords with an Euclidean distance \(\delta_w\) from the \(i\)th codeword. Given that \(c_i\) is transmitted, let the error probability of SSD(D) be upper bounded by \(P_i(\mathcal{E}_D)\). By taking the expectation over all codewords,

\[
P(\mathcal{E}_D) \leq \frac{1}{|C|} \sum_{c_i \in C} P_i(\mathcal{E}_D).
\]

(24)

Now, if we assume that \(P_i(\mathcal{E}_D)\) is of the union bound form; \(P_i(\mathcal{E}_D) = \sum_w G_w(i) P^{(w)}(\mathcal{E}_D)\), where \(P^{(w)}(\mathcal{E}_D)\) is the probability of a sphere decoder error due to incorrectly decoding a codeword at a distance \(\delta_w\) when \(c_i\) is transmitted. The error probability of SSD(D) can thus be upper bounded by \(P(\mathcal{E}_D) \leq \sum_{\delta_w > 0} \bar{G}_w P^{(w)}(\mathcal{E}_D)\), where \(P^{(w)}(\mathcal{E}_D)\) is the probability that the sphere decoder erroneously decodes a codeword at a distance \(\delta_w\) from the transmitted codeword and

\[
\bar{G}_w = \frac{1}{|C|} \sum_{c_i \in C} G_w(i),
\]

(25)

is the average number of codewords which are at an Euclidean distance \(\delta_w\) from another codeword. For an arbitrary finite code or lattice \(C\), using arguments from the previous sections, the error probability SSD(D) can be upper bounded by

\[
P(\mathcal{E}_D) \leq \min_{D' \leq D} \{ P(\mathcal{E}_{ML}, z \in \Omega_{D'}) + P(z \notin \Omega_{D'}) \},
\]

(26)
where $P(z \notin \Omega_D)$ is given by (4) and

$$P(\mathcal{E}_{ML}, z \in \Omega_D) = \sum_{0 < \frac{\delta w}{2} < D} \bar{G}_w \int_{\frac{\delta w}{2}}^{D} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-z^2/2\sigma^2} \frac{1}{\Gamma_r (\frac{n_d-1}{2}, \frac{D^2-z^2}{2\sigma^2})}dz. \quad (27)$$

The Hughes upper bound on the ML error probability is $P(\mathcal{E}_{ML}) \leq \min_D P(\Psi(D))$ [26], where

$$\Psi(D) \triangleq P(\mathcal{E}_{ML}, z \in \Omega_D) + P(z \notin \Omega_D). \quad (28)$$

The radius $D_o$ that minimizes this error probability is the root of the equation [27]

$$\sum_{0 < \frac{\delta w}{2} < D} \bar{G}_w \int_{0}^{\theta_{w,D}} \sin(\theta)^{n_d-2}d\theta = \frac{\sqrt{\pi} \Gamma (\frac{n_d-1}{2})}{\Gamma (\frac{n_d}{2})}, \quad (29)$$

where $\theta_{w,D} = \cos^{-1}(\delta w/2D)$. From (26), the upper bound on the sphere decoding error probability is given by

$$P(\mathcal{E}_D) \leq \begin{cases} 
\Psi(D), & D < D_o; \\
\Psi(D_o), & D \geq D_o.
\end{cases}$$

Furthermore, the optimum radius $D_o$ does not depend on the channel and can be the radius of choice for near maximum likelihood decoding. The bound developed here is universal in the sense that also applies for the case of a linear code with equal energy codewords. However, it is to be noted that the Hughes bound on ML decoding is not tighter than the Poltyrev tangential sphere bound [28].

For the case of $M$-PSK modulation of a linear code, the constellation may not result in a Hamming space if $M > 4$. In such a case the ensemble average weight enumerator $\bar{G}_w$ can be used with the bounds of Sec. II to analyze the performance. (The same technique can also be used with the results in next sections.)

**Example 6:** Assume an $(15,3)$ RS code over $F_{16}$ and assume a one-to-one mapping from the symbols of $F_{16}$ to the points of an 16-QAM modulation [16], whose average energy per symbol is 10. The ensemble weight enumerator $\bar{G}_w$ was numerically computed to evaluate the bounds. The radius that minimizes the bound on the ML error probability is $D_o = 12.9$. In Fig. 5, we confirm that the bounds on the sphere decoder error probability agree with the simulations for the case of $D = 10$. We also compare the simulated performance of ML error probability $P(\mathcal{E}_{ML}, z \in \Omega_D)$ to that of the analytic performance in both cases. At low SNRs this probability is low as the probability of the received word falling inside the sphere is relatively low. As more received words fall inside the sphere, the ML error probability increases as the SNR increases. At
a certain SNR, the probability of the ML error starts decreasing due to the improved reliability of the received word.

IV. PERFORMANCE OF SPHERE DECODING ON BINARY SYMMETRIC CHANNELS

In this section, an upper bound on the performance of the hard-decision sphere decoder, when the code is transmitted over the BSC, is derived. Transmitting a binary codeword over a binary input AWGN channel followed by hard decisions is equivalent to transmitting it on a BSC with a cross over probability \( p = Q(\sqrt{2R\gamma}) \) where \( \gamma \) is the bit signal to noise ratio. In case of M-PSK signaling with gray encoding, \( p \approx \frac{p_c}{\log_2(M)} \) where \( p_c = 2Q(\sqrt{2k\gamma} \sin \frac{\pi}{M}) \) [16].

Let \( y \) be the received word when the codeword \( c \) is transmitted over an BSC channel. The HD sphere decoder with radius \( m \), \( \text{HSD}(m) \), finds the codeword \( \hat{c} \), if it exists, such that

\[
\hat{c} = \arg \min_{v \in \mathcal{C}} d(y, v)
\]

subject to

\[
d(y, v) < m + 1,
\]

where \( d(y, v) \) is the Hamming distance between \( y \) and \( v \). Let \( \zeta = d(y, c) \), then, from the linearity of the code, the probability that the received word is outside a Hamming sphere (ball) of radius \( m - 1 \) centered around the transmitted codeword is

\[
P(\zeta \geq m) = \sum_{t=m}^{n} \binom{n}{t} p^t (1-p)^{n-t}.
\]

Poltyrev [20] derived a tight bound on the performance of the HD-ML decoder based on,

\[
P(\mathcal{E}_{ML}) \leq \min_m \{ P(\mathcal{E}_{ML}, \zeta < m) + P(\zeta \geq m) \}.
\]

The minimum of the above equation is at \( m_o \) where \( m_o \) is the smallest integer \( m \) such that [20]

\[
\sum_{b=1}^{2m} G_b \sum_{r=[\frac{b}{2}]}^{m} \binom{b}{r} \binom{n-b}{m-r} \geq \binom{n}{m}.
\]

We now turn our attention to the HD sphere decoder with an arbitrary decoding radius. Let \( P(\Sigma_m) \), be the error plus failure probability of the hard decision sphere decoder, \( \text{HSD}(m-1) \), then \( P(\Sigma_m) \) could be written as

\[
P(\Sigma_m) = P(\Sigma_m, \zeta < m) + P(\Sigma_m | \zeta \geq m) P(\zeta \geq m)
\]

\[
= P(\mathcal{E}_{ML}, \zeta < m) + P(\zeta \geq m),
\]
where we used the fact that \( P(\Sigma_m | \zeta \geq m) = 1 \) and the observation that given that \( \zeta < m \), the conditional error probability of the HSD\((m - 1)\) and the HD-ML decoders are the same. The last term in the above equation is a lower bound on the failure probability of the HSD\((m - 1)\) decoder. Similar to soft decision case, we have the following lemma,

**Lemma 7:** A lower bound on the performance of a hard decision sphere decoder, HSD\((m - 1)\), over a BSC with parameter \( p \) is \( P(\Sigma_m) \geq \sum_{t=m}^{n} \binom{n}{t} p^t (1-p)^{n-t} \).

To develop a tight upper bound on \( P(\Sigma_m) \), we consider two cases:

**Case I:** The decoding radius \( m \geq m_o \). Equation (34) could be written as

\[
P(\Sigma_m | m \geq m_o) = P(\mathcal{E}_{ML}, \zeta < m_o) + P(\mathcal{E}_{ML}, m_o \leq \zeta < m) + P(\zeta \geq m_o).
\]

It follows that \( P(\Sigma_m | m \geq m_o) \leq P(\mathcal{E}_{ML}, \zeta < m_o) + P(\zeta \geq m_o) \). We observe that the upper bound reduces to that of the HD-ML case (32). By recalling that the minimum of (32) is achieved at \( m_o \), the bound of (34) is looser than (7) when \( m > m_o \). The intuition behind this is that the performance of a sphere decoder with a decoding radius \( m_o - 1 \) or greater approaches that of the ML decoder.

**Case II:** The decoding radius \( m < m_o \). Noticing that the sphere \( \{ \zeta < m \} \subset \{ \zeta < m_o \} \), \( P(\Sigma_m | m < m_o) \) is indeed given by (34).

Thus, we have proved the following theorem,

**Theorem 8:** The performance of a hard-decision sphere decoder with a decoding radius \( m - 1 \) when used for decoding a linear code with a weight spectrum \( G_b \) over an BSC channel with a cross-over probability \( p \) is upper bounded by

\[
P(\Sigma_m) \leq \begin{cases} 
P(\mathcal{E}_{ML}, \zeta < m_o) + P(\zeta \geq m_o), & m \geq m_o; \\
P(\mathcal{E}_{ML}, \zeta < m) + P(\zeta \geq m), & m < m_o, 
\end{cases}
\]

where \( m_o \) is radius that minimizes (32) and is the solution of (33). \( P(\zeta \geq m) \) is given by (31) and the joint probability of an HD-ML error and \( d(y, c) < m \) is upper bounded by the union bound \[20\], \( P(\mathcal{E}_{ML}, \zeta < m) \leq \sum_{b=1}^{2(m-1)} G_b \sum_{r=\left[\frac{m}{2}\right]}^{m-1} \binom{m}{r} p^r (1-p)^{m-r} \sum_{s=0}^{m-r-1} \binom{m-r}{s} p^s (1-p)^{m-r-s} \).

**A. Numerical Examples**

In this subsection, the bounds developed for SD and HD sphere decoding are evaluated and compared with the performance of the corresponding sphere decoders, [13] and [14] respectively.

In Fig. 6, we compare the analytical bounds to simulations of sphere decoding of an \((15,7)\) BCH code BPSK modulated and transmitted over an AWGN channel. The minimum distance
of the BCH code is $5$. The critical decoding Euclidian radius of the soft decision decoder is $D_\phi = 3.17$ while the critical Hamming decoding radius of the hard decision decoder is $m_o = 3$.

We observe that the simulated performance is tightly upper bounded by the analytical bounds of theorems 4 and 8 for soft and hard decision sphere decoding respectively. The larger the decoding radius the nearer the performance is to maximum likelihood decoding.

V. PERFORMANCE OF SPHERE DECODING ON Q-ARY SYMMETRIC CHANNELS

Now consider an $(n, k, d)$ RS code and a hard-decision sphere decoder which can correct $\tau$ symbol errors, where the symbols are in $F_q$. The Berlekamp-Massey algorithm is a well known polynomial time algorithm that can correctly decode words which are at a (symbol) Hamming distance of $\tau_{BM} = \lfloor \frac{n-k}{2} \rfloor$ from the transmitted codeword. The error probability of bounded distance decoding of RS codes is well studied (cf. [29]). Recently, Guruswami and Sudan [15] developed a list decoding algorithm that can correct upto $\tau_{GS} = \lceil n - \sqrt{nk} - 1 \rceil$ symbol errors. To analyze this case, we first derive a bound on the performance of the corresponding ML decoder.

A. Bound on the Maximum Likelihood decoding of linear block codes on q-ary symmetric channels.

We will assume an $(n, k, d)$ linear code over $F_q$ transmitted over a $q$-ary symmetric channel. The probability that a symbol is correctly received will be denoted by $s$, while the probability that it is received as another symbol will be $p = (1 - s)/(q - 1)$. Transmitting a $q$-ary code over an AWGN channel followed by hard-decision can be modeled as transmitting it over a $q$-ary symmetric channel. Assume that $q = 2^m$, the channel alphabet size is $2^b$, $b \leq m$, and each $q$-ary symbol is mapped to $m/b$ channel symbols. Let $p_c$ be the probability that a channel symbol is incorrectly decoded, then $s = (1 - p_c)^{m/b}$. For example, if the channel is a BPSK channel with a bit signal to noise ration $\gamma$, $q = 2^m$ and the binary image of the RS code is transmitted, then a $q$-ary symbol is correctly received if all the $m$ bits in its binary image are correctly received, i.e. $s = \left(1 - Q\left(\sqrt{2\frac{k}{n}\gamma}\right)\right)^{m}$.

Let $\zeta$ be the Hamming distance between the transmitted codeword and the received $q$-ary word. Then, similar to the binary case, the ML error probability can be upper bounded as follows,

$$P(E_{ML}) \leq \min_m \{ P(E_{ML}, \zeta < m) + P(\zeta \geq m) \}. \quad (36)$$
Assuming that the code is linear, the probability that the received $q$-ary word lies outside a Hamming sphere (ball) of radius $m - 1$ centered around the transmitted word is

$$P(\zeta \geq m) = \sum_{\alpha=m}^{n} \binom{n}{\alpha} (1 - s)^{\alpha} s^{n-\alpha}. \quad (37)$$

The above equation will also provide a lower bound on the performance of the sphere decoder.

The first term in (36) is upper bounded in the following lemma.

**Lemma 9:** For an $(n, k, d)$ linear code over $F_q$, with a weight enumerator $G_w$, transmitted over a $q$-ary symmetric channel with parameters $s$ and $p$,

$$P(\mathcal{E}_{ML}, \zeta < m) \leq \sum_{w=d}^{2(m-1)} G_w \sum_{\alpha=0}^{m-1} \sum_{\eta=\lceil \frac{w-\alpha}{2} \rceil}^{w-\alpha} \frac{w!}{\eta! \alpha!(w-\eta-\alpha)!} \sum_{\alpha=0}^{m-1-\eta-\alpha} \left( \sum_{\beta=0}^{n-w} \binom{n-w}{\beta} (1 - s)^{\beta} s^{n-w-\beta} \right) \right) \sum_{w=0}^{\eta+\alpha-1} \left( \sum_{m'=1}^{m' - \eta - \alpha} \frac{w!}{\eta! \alpha!(w-\eta-\alpha)!} \right) \sum_{s=0}^{\eta+\alpha-1} \left( \sum_{m'=1}^{m' - \eta - \alpha} \frac{w!}{\eta! \alpha!(w-\eta-\alpha)!} \right)$$. \quad (38)

**Proof:** We will assume that the all-zero codeword is transmitted. Now consider a codeword $c$ with Hamming weight $w$ and assume the received word $r$ has a Hamming weight $m' - 1$. Consider the $w$ nonzero symbols in $c$ and the corresponding coordinates in $r$. Let $r$ and $c$ have the same symbols in $\eta$ of these coordinates. Let $\alpha$ of these $w$ coordinates in $r$ be neither zero nor match those in $c$, and $w - \eta - \alpha$ of the remaining coordinates be zero. Since the Hamming weight of $r$ is $m' - 1$, there must be $m' - 1 - \eta - \alpha$ non-zero symbols in the remaining $n - w$ coordinates and the remaining symbols will be zero. The probability of receiving such a word is $\sum_{w=0}^{\eta+\alpha-1} \left( \sum_{m'=1}^{m' - \eta - \alpha} \frac{w!}{\eta! \alpha!(w-\eta-\alpha)!} \right) \sum_{s=0}^{\eta+\alpha-1} \left( \sum_{m'=1}^{m' - \eta - \alpha} \frac{w!}{\eta! \alpha!(w-\eta-\alpha)!} \right) \sum_{w=0}^{\eta+\alpha-1} \left( \sum_{m'=1}^{m' - \eta - \alpha} \frac{w!}{\eta! \alpha!(w-\eta-\alpha)!} \right)$.

In such a case, the Hamming distance between $r$ and $c$ is $w + m' - 1 - 2\eta - \alpha$. An ML error result if this is less than the weight of $r$, i.e., if $\eta \geq \lceil \frac{w-\alpha}{2} \rceil$. By summing over all possible combinations of $\eta$ and $\alpha$ and applying the union bound for all codewords that can be within a Hamming distance $m'$ from $r$, the error probability is upper bounded by $\sum_{w=0}^{2(m'-1)} G_w \sum_{\alpha=0}^{m'-1} \sum_{\eta=\lceil \frac{w-\alpha}{2} \rceil}^{w-\alpha} \left( \sum_{m'=1}^{m' - \eta - \alpha} \frac{w!}{\eta! \alpha!(w-\eta-\alpha)!} \right) \sum_{s=0}^{\eta+\alpha-1} \left( \sum_{m'=1}^{m' - \eta - \alpha} \frac{w!}{\eta! \alpha!(w-\eta-\alpha)!} \right) \sum_{s=0}^{\eta+\alpha-1} \left( \sum_{m'=1}^{m' - \eta - \alpha} \frac{w!}{\eta! \alpha!(w-\eta-\alpha)!} \right)$. Applying the union bound for all received words with Hamming weights less than $m$, $m' \leq m$, the result follows.

We are now ready to prove the following theorem,

**Theorem 10:** The ML error probability of an $(n, k, d)$-ary linear code on a $q$-ary symmetric
channel is upper bounded by
\[ P(\mathcal{E}_{ML}) \leq 2^{2(m_\alpha - 1)} \sum_{w=d} G_w \sum_{\alpha=0}^{m_\alpha-1} \sum_{\eta=\left\lceil \frac{w-\alpha}{2} \right\rceil} w! \eta! \alpha! (w-\eta-\alpha)! p^n (1-p-s)^\alpha s^{w-\eta-\alpha} \]
where \( m_\alpha \) is the smallest integer \( m \) such that
\[ \sum_{w=d} G_w \sum_{\alpha=0}^{m} \frac{q-2}{q-1} \sum_{\eta=\left\lceil \frac{w-\alpha}{2} \right\rceil} w! \eta! \alpha! (w-\eta-\alpha)! \left( \frac{1}{q-1} \right) \frac{w!}{\eta! \alpha! (w-\eta-\alpha)!} \left( \frac{n-w}{m-\eta-\alpha} \right) \geq \binom{n}{m}. \]  

**Proof:** The upper bound follows by substituting (38) and (37) in (36). Observe that the first term in (39) is increasing in \( m \) while the second is decreasing in \( m \). Optimizing over the radius \( m \), the minimum is achieved at the first integer \( m \) such that
\[ \sum_{w=d} G_w \sum_{\alpha=0}^{m} \frac{q-2}{q-1} \sum_{\eta=\left\lceil \frac{w-\alpha}{2} \right\rceil} w! \eta! \alpha! (w-\eta-\alpha)! \left( \frac{1}{q-1} \right) \frac{w!}{\eta! \alpha! (w-\eta-\alpha)!} \left( \frac{n-w}{m-\eta-\alpha} \right) \geq \binom{n}{m}. \]  

It is worth noting that the optimum radius \( m_\alpha \) which minimizes the bound on the ML error probability only depends on the weight enumerator of the code and the size of its finite field. Since the optimum radius does not depend on the SNR, it is valid for \( q \)-ary symmetric channels at any SNR. Similar to the binary case [20], we establish below a connection between \( m_\alpha \) and the covering radius of the code.

**Lemma 11:** The covering radius of a linear code on \( F_q \) is lower bounded by \( m_\alpha - 1 \), where \( m_\alpha \) is given by Th. 10.

**Proof:** Define \( L(m) \) to be the left hand side term in (39) and \( c_o \) to be the all zero codeword. Similar to the proof of Lem. 9, one can show that \( (q-1)^m L(m) = \{|r \in F_q^n : d(r, c_o) = m \text{ and } d(r, c_i) \leq m ; \text{ } c_i \in C \setminus c_o \} \}. \) Also, \( (q-1)^m \binom{n}{m} = \{|r \in F_q^n : d(r, c_o) = m \} \}. \) Since \( (q-1)^{m_o - 1} L(m_o - 1) < (q-1)^{m_o - 1} \binom{n}{m_o - 1} \), it follows that there exit words \( r \in F_q^n \) such that \( \min_{c \in C} d(r, c) = m_o - 1 \) and this minimum is achieved when \( c \) is the all zero codeword \( c_o \). By recalling that the covering radius is [30] \( R_c = \max_{r \in F_q^n} \min_{c \in C} d(r, c) \), it follows that \( R_c \geq m_o - 1 \).

**B. Hard Decision Sphere decoding of linear block codes on \( q \)-ary symmetric channels.**

Here, we consider the case when the decoder is a \( q \)-ary hard decision sphere decoder. As for the binary case, the HSD\((m-1)\) can correctly decode a codeword if the number of \( q \)-ary symbol
errors is \( m - 1 \) or less. Thus Th. 8 will give the bound on the error plus failure probability of the sphere decoder. However, in this case, \( P(\zeta \geq m) \), \( P(\mathcal{E}_{ML}, \zeta < m) \) and \( m_o \) are given by (37), (38) and (39) respectively.

C. Numerical Examples

In Fig. 7, we show bounds on the performance of HD decoding of the near half rate \((31, 15)\) RS code over \( F_{32} \) when its binary image is transmitted over an AWGN channel followed by hard-decisions. The optimum binary decoding radius is 18. Thus the closer the decoding radius is to 18, the better the performance of the sphere decoder. The HD-ML decoder has more than 2 dB coding gain over the Berlekamp Massey (BM) decoder, which can correct 8 symbol errors. It is observed that the average performance of an HD sphere decoder, with a (binary Hamming) radius 8, closely upper bounds that of the HD-BM decoder that can correct 8 symbol errors. The HD-GS decoder can correct one more symbol error than the BM decoder. The performance of the GS algorithm is analyzed by modeling it as 16-ary HD sphere decoder of radius 9. Consequently, one can observe that a hard-decision sphere decoder with a binary decoding radius of 10 outperforms the symbol based GS decoder.

In Fig. 8, the binary image of the \((15, 3)\) RS code is BPSK modulated over an AWGN channel. For 16-ary hard decisions, the channel is modeled as an QSC. The performance bound of the hard ML (H-ML) decoder is shown (Th. 10) and is the same as an HSD of radius 9. The bounds of (37) and (38) are also shown and labeled as \( F(9) \) and \( E(9) \) respectively. As seen, the three bounds (‘bnd’) are in close agreement with the simulation (‘sim’), for such a hypothetical sphere decoder. The error probability of the GS decoder with radius 8 is simulated and agrees with the bound of Th. 8. For reference proposes, we show the average error probability of the soft decision bit level ML (S-ML) decoder (cf [22]) which has about 4 dB gain over the symbol H-ML decoder.

VI. A Note on Complexity

In Fig. 9, the empirical complexity exponents of SSD of the \((24, 12)\) Golay code BPSK modulated over an AWGN channel are shown. It is clear that for a larger decoding radius there is a price paid in terms of the complexity. We also show the complexity of the SSD whose radius changes such that with a probability of 0.9 the transmitted word is inside the sphere centered
around the received one. At a slighter increase in average complexity one can achieve ML decoding, by gradually increasing the decoding radius until a word is found. The corresponding complexity is shown as \( r^{2.90 + \text{cumulative}} \). The variation of the radius versus the SNR is shown in Fig. 10. The expected complexity of sphere decoding was thoroughly analyzed in [31].

VII. CONCLUSIONS

Bounds on the error plus failure probability of hard-decision and soft-decision sphere decoding of block codes were derived. By comparing with the simulations of the corresponding decoders, we demonstrate that our bounds are tight. The ML performance of codes on \( q \)-ary symmetric channels is analyzed. The performance of sphere decoding of Reed Solomon codes and their binary images was analyzed. Moreover, the bounds are extremely useful in predicting the performance of the sphere decoders at the tail of error probability when simulations are prohibitive. The bounds allows one to pick the radius of the sphere decoder that best fits the performance, throughput and complexity requirements of the system.

ACKNOWLEDGMENT

This research was supported by NSF grant no. CCF-0514881 and grants from Sony, Qualcomm, and the Lee Center for Advanced Networking.

REFERENCES


[25] ——, “Iterative algebraic soft-decision list decoding of Reed-Solomon codes,” accepted for publication in IEEE Journal on Selected Areas in Communications.
Fig. 1. Case A: The sphere $\Omega_D$ lies totally inside the Cone $V_\phi (D \leq \sqrt{\pi c \sin(\phi)})$
Fig. 2. Case B: The sphere $\Omega_D$ intersects the cone $V_\phi$; the apex of the cone $V_\phi$ lies outside the sphere $\Omega_D$ ($\sqrt{n_c} \sin(\phi) < D < \sqrt{n_c}$). In case $D \geq \sqrt{n_c}$, the apex of the cone $V_\phi$ lies inside the sphere $\Omega_D$. 
Fig. 3. Bounds on the performance of soft-decision sphere decoding of the (24, 12) Golay code when QPSK modulated over an AWGN channel.
Fig. 4. Bounds on the performance of SSD of a binary image of the (15, 11) Reed Solomon code BPSK modulated on an AWGN channel.
Fig. 5. The (15, 3) RS code is 16-QAM modulated and transmitted over an AWGN channel. The sphere decoder is a soft decision sphere decoder with an Euclidean radius 10. The bounds are compared to simulations for a sphere decoding ML error and the error plus failure probability.
Fig. 6. Bounds on the codeword error rate of soft-decision and hard-decision sphere decoding of the (15, 7) BCH code BPSK modulated over an AWGN channel. The simulations (labeled by ‘sim’) are tightly upper bounded by the analytic bounds (labeled by ‘bnd’).
Fig. 7. Bounds on the performance of hard-decision sphere decoding of the (31, 15) RS code BPSK on an AWGN channel.
Fig. 8. The (15, 3) RS code is BPSK modulated and transmitted over an AWGN channel. For the 16-ary hard-decision decoder, the channel is an QSC. The bounds are compared to simulations for a sphere decoding ML error, sphere decoding failure, and their sum (error plus failure probability). The optimum radius for the ML bound is 9. The GS radius is 8.
Fig. 9. Complexity exponent for SSD of the (24, 12) Golay code.

Fig. 10. Statistical Decoding Radius vs Fixed Decoding Radius for the (24, 12) Golay code.