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Reed-Muller Codes

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Reed-Muller Codes

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ABSTRACT

Reed-Muller (RM) codes are among the oldest, simplest and perhaps most ubiquitous family of codes. They are used in many areas of coding theory in both electrical engineering and computer science. Yet, many of their important properties are still under investigation. This work covers some of the developments regarding the weight enumerator and the capacity-achieving properties of RM codes, as well as some of the algorithmic developments. In particular, it discusses connections established between RM codes, thresholds of Boolean functions, polarization theory, hypercontractivity, and the techniques of approximating low weight codewords using lower degree polynomials (when codewords are viewed as evaluation vectors of degree r polynomials in m variables).

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It then overviews some of the algorithms for decoding RM codes, giving both algorithms with provable performance guarantees for every block length, as well as algorithms with state-of-the-art performances in practical regimes, which do not perform as well for large block length. Finally, some applications of RM codes in theoretical computer science and signal processing are given.

1

Introduction

A large variety of codes have been developed over the past 70 years. These were driven by various objectives, in particular, achieving efficiently the Shannon capacity [137], constructing perfect or good codes in the Hamming worst-case model [67], matching the performance of random codes, improving the decoding complexity, the weight enumerator, the scaling law, the universality, the local properties of the code [28], [78], [79], [98], [99], [123], [154], and more objectives in theoretical computer science such as in cryptography (e.g., secret sharing, private information retrieval), pseudorandomness, extractors, hardness amplification or probabilistic proof systems; see [1] for references. Among this large variety of code developments, one of the first, simplest and perhaps most ubiquitous code is the Reed-Muller (RM) code.

The RM code was introduced by Muller in 1954 [109], and Reed developed shortly after a decoding algorithm decoding up to half its minimum distance [121]. The code construction can be described with a greedy procedure. Consider building a linear code (with block length a power of two); it must contain the all-0 codeword. If one has to pick a second codeword, then the all-1 codeword is the best choice under any meaningful criteria. If now one has to keep these two codewords,

the next best choice to maximize the code distance is the half-0 half-1 codeword, and to continue building a basis sequentially, one can add a few more vectors that preserve a relative distance of half, completing the simplex code, which has an optimal rate for the relative distance half. Once saturation is reached at relative distance half, it is less clear how to pick the next codeword, but one can simply re-iterate the simplex construction on any of the support of the previously picked vectors, and iterate this after each saturation, reducing each time the distance by half. This gives the RM code, whose basis is equivalently defined by the evaluation vectors of bounded degree monomials.

As mentioned, the first order RM code is the augmented simplex code or equivalently the Hadamard code, and the simplex code is the dual of the Hamming code that is “perfect”. This strong property is clearly lost once the RM code order gets higher, but RM codes preserve nonetheless a decent distance (at root block length for constant rate). Of course this does not give a “good” family of codes (i.e., a family of codes with asymptotically constant rate and constant relative distance), and it is far from achieving the distance that other combinatorial codes can reach, such as Golay codes, BCH codes or expander codes [99]. However, once put under the light of random errors, i.e., the Shannon setting, for which the minimum distance is no longer the right figure or merit, RM codes may perform well again. In [77], Levenshtein and co-authors showed that for the binary symmetric channel, there are codes that improve on the simplex code in terms of the error probability (with matching length and dimension). Nonetheless, in the lens of Shannon capacity, RM codes seem to perform very well. In fact, more than well; it is plausible that they achieve the Shannon capacity on any Binary-input Memoryless Symmetric (BMS) channel [1], [2], [43], [89], [90], [122] and perform comparably to random codes on criteria such as the scaling law [70] or the weight enumerator [82]–[84], [99], [127], [142].

The fact that RM codes have good performance in the Shannon setting, and that they seem to achieve capacity, has long been observed and conjectured. It is hard to track back the first appearance of this belief in the literature, but [89] reports that it was likely already present in the late 60s. The claim was mentioned explicitly in a 1993 talk by Shu Lin, entitled “RM Codes are Not So Bad” [95]. It appears that a 1994

paper by Dumer and Farrell contains the earliest printed discussion on this matter [50]. Since then, the topic has become increasingly prevalent¹ [1], [9], [11], [39], [43], [45], [104].

But the research activity has truly sparked with the emergence of polar codes [11]. Polar codes are close relatives of RM codes. They are derived from the same square matrix (the matrix whose rows correspond to evaluations of multilinear monomials) but with a different rule of row selection. The more sophisticated and channel dependent construction of polar codes gives them the advantage of being provably capacity-achieving on any BMS channel, due to the polarization phenomenon. Even more impressive is the fact that they possess an efficient decoding algorithm down to the capacity.

Shortly after the polar code breakthrough, and given the close relationship between polar and RM codes, the hope that RM codes could also be proved to achieve capacity on any BMS started to propagate, both in the electrical engineering and computer science communities. A first confirmation of this was obtained in extremal regimes of the BEC and BSC [1], exploiting new bounds on the weight enumerator [84], and a first complete proof for the BEC at constant rate was finally obtained in [90]. The paper [122] presented a major breakthrough proving that constant-rate RM codes indeed achieve capacity on all BMS channels under bit-MAP decoding. While [122] comes close to proving the conjecture, the question of whether RM codes achieve capacity under block-MAP decoding still remains open.

The papers mentioned in the previous paragraph however did not exploit the close connection between RM and polar codes. This connection was studied in [2] where it was shown that the RM transform is also polarizing and that a third variant of the RM code achieves capacity on any BMS channel. Furthermore [2] conjectured that this variant is indeed the RM code itself.

Polar codes and RM codes can be compared in different ways. In most performance metrics, and putting aside the decoding complexity, RM codes seem to be superior to polar codes [2], [104]. Namely, they seem

¹The capacity conjecture for the BEC at constant rate was posed as one of the open problems at the Information Theory Semester at the Simons Institute, Berkeley, in 2015.

to achieve capacity universally and with an optimal scaling-law, while polar codes have a channel-dependent construction with a suboptimal scaling-law [66], [70], [71]. However, RM codes seem more complex both in terms of obtaining performance guarantees (as evidenced by the long standing conjectures) and in terms of their decoding complexity.

Efficient decoding of RM codes is the second main outstanding challenge. Many algorithms have been proposed since Reed's algorithm [121], such as [20], [48], [49], [51], [64], [125], [141], and newer ones have appeared in the post polar code period [129], [130], [153]. Some of these already show that at various block-lengths and rates that are relevant for communication applications, RM codes are indeed competing or even superior to polar codes [104], [153], even compared to the improved versions considered for 5G [57].

This survey is meant to overview these developments regarding both the performance guarantees (in particular on weight enumerator and capacity) and the decoding algorithms for RM codes. At the end of this survey, we discuss a few applications of RM codes in the areas beyond communication, e.g., applications in low degree testing, private information retrieval, and compressed sensing.

1.1 Outline of the Survey and Differences from a Previous Version

Part of this monograph was taken from a previous survey [4] written by the first author, the third author and the fourth author. At the same time, we have added quite a few new elements and optimized the presentation of the contents from [4]. Below we give the outline of this new survey and discuss the difference from [4].

We start in Section 2 with the main definitions and basic properties of RM codes. Most parts of this section already appeared in [4], e.g., the code parameters, recursive structure, duality, automorphism group, and local properties. We have, however, added two new subsections discussing the cyclic property of punctured RM codes and the nonlinear subcodes of RM codes. In Section 3, we introduce some performance measures and important quantities in channel coding. This is a new section that has not appeared in [4]. We then cover the bounds on the weight enumerator of RM codes in Section 4. In Section 5, we cover

1.1. *Outline of the Survey and Differences from a Previous Version* 7

the capacity-achieving results, using tools from the weight enumerator and sharp thresholds of monotone Boolean functions. In Section 6, we explore the connection between RM codes and polar codes. Although Sections 4–6 have appeared in [4], we have revised the organization of these 3 sections and added some proofs to better explain the results as well as covered results that appeared between the publication time of these two surveys. Section 7 is a new section that describes the finite-length scaling of random codes, RM codes and polar codes. We then cover various decoding algorithms in Section 8, providing pseudo-codes for them. This section is similar to the previous version [4]. Finally, in Section 9, we discuss some applications of RM codes beyond communication and channel coding, which were not covered in the previous version [4].

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