EFFICIENTLY DECODING REED MULLER CODES FROM RANDOM ERRORS

Ben Lee Volk

Joint work with Ramprasad Saptharishi Amir Shpilka Tel Aviv University









Given the truth-table of a polynomial $f \in \mathbb{F}[x_1, ..., x_m]$ of degree $\leq r$, with 1/2 - o(1) of the entries flipped, recover f efficiently.

0	1	1	0	1	0	0	1	1	0	0	0	1	1	1	1

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This talk is about decoding **Reed-Muller codes** from **random** errors.

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A linear code, with

• Block Length: $2^m := n$

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- Dimension: $\binom{m}{0} + \binom{m}{1} + \dots + \binom{m}{r} := \binom{m}{\leq r}$
- Rate: dimension/block length = $\binom{m}{< r}/2^m$

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Worst Case Errors: Up to d/2 (*d* is minimal distance).



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Gopalan-Klivans-Zuckerman08, Bhowmick-Lovett15: List decoding radius = *d*.

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- An ongoing research endeavor: how do Reed-Muller perform in Shannon's random error model?

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MODELS FOR RANDOM CORRUPTIONS (CHANNELS)

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Shannon48: max rate that enables decoding (w.h.p.) is 1 - p (for BEC) and 1 - H(p) (for BSC). Codes achieving bound called **capacity achieving**.



Category:Capacity-achieving codes

From Wikipedia, the free encyclopedia

Pages in category "Capacity-achieving codes"

This category contains only the following page. This list may not reflect recent changes (learn more).

Ρ

• Polar code (coding theory)

Categories: Error detection and correction

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- **Question:** Can we correct $(1-o(1))\binom{m}{< r}$ erasures?
- if yes, RM(m, m r 1) achieves capacity for the BEC

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sanity check: r = 1 is good

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Open Problem: Prove for every degree *r*.

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Corollary #3: If $RM(m, \left(\frac{1+\rho}{2}\right)m)$ achieves capacity, efficient decoding algo for $2^{h\left(\frac{1-\rho}{2}\right)m}$ random errors in $RM(m, \rho m)$.

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PROOF IDEA

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recall: $\{u_1, \dots, u_t\}$ correctable from erasures iff $\{u_1^r, \dots, u_t^r\}$ are linearly independent.

DUAL POLYNOMIALS

Fact: If $\{\mathbf{u}_1^r, \dots, \mathbf{u}_t^r\}$ lin. indep., \exists polys $\{f_1, \dots, f_t\}$ of deg $\leq r$ such that

$$f_i(\mathbf{u}_j) = \begin{cases} 1 & i = j \\ 0 & i \neq j. \end{cases}$$

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Our approach would be to find those polynomials.

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 $\forall \text{ monomial } M, \text{ deg } M \leq r:$ $\sum_{i=1}^{t} f(\mathbf{u}_i) = f(\mathbf{v}) = 1$

$$(f \text{ non-trivial})$$

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If U lin. indep. and $\mathbf{v} = \mathbf{u}_i \in U$, f_i is a solution. Conversely, if solvable and U lin. indep., can show $\mathbf{v} \in U$.

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Input to the algo: $\mathbf{y} = \mathbf{c} + \mathbf{e}$ with $\mathbf{c} \in RM(m, m - 2r - 2)$, and \mathbf{e} characteristic vector of $U = {\mathbf{u}_1, \dots, \mathbf{u}_t}$.

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Syndrome of y: $E(m, 2r+1) \cdot \mathbf{y} = E(m, 2r+1) \cdot \mathbf{e}$.

(recall: E(m, 2r + 1) is PCM of RM(m, m - 2r - 2))

Every coefficient in the system is of the form $\sum_{i=1}^{t} g(\mathbf{u}_i)$ for poly g of degree $\leq 2r + 1$.

How to compute the coefficients?

Input to the algo: $\mathbf{y} = \mathbf{c} + \mathbf{e}$ with $\mathbf{c} \in RM(m, m - 2r - 2)$, and \mathbf{e} characteristic vector of $U = {\mathbf{u}_1, \dots, \mathbf{u}_t}$.

Syndrome of y: $E(m, 2r + 1) \cdot \mathbf{y} = E(m, 2r + 1) \cdot \mathbf{e}$.

(recall: E(m, 2r+1) is PCM of RM(m, m-2r-2))

Corollary: The syndrome of **y** is a $\binom{m}{\leq 2r+1}$ long vector α , where $\alpha_M = \sum_{i=1}^{t} M(\mathbf{u}_i)$, for every monom M, deg $M \leq 2r + 1$.

BACK TO BEC

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