## **Codes for Input-Constrained Channels**

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## **Channel Models and Constraints**

- Consider the setting of the transmission of information across input-constrained binary-input channels.
- We work with both stochastic and adversarial noise models.





► The constraints we work with find application in a number of domains:

1. Runlength-limited (RLL) constraints: Alleviate ISI in magneto-optical recording



2. Subblock composition constraints: Aid in energy-harvesting in communication



3. Charge constraints: Ensure spectral nulls (DC-freeness) in frequency spectrum





Explicit Codes over RLL Input-Constrained Binary Memoryless Symmetric (BMS) Channels

• Our interest is in the  $(d, \infty)$ -RLL constraint, i.e., there must be at least d 0s between successive 1s. Example: When d = 2,





BSC with crossover probability p

1 - p

BEC with erasure probability  $\epsilon$ 

 $\triangleright$  We construct coding schemes, using Reed-Muller (RM) codes of rate R, over such input-constrained BMS channels of unconstrained capacity  $C \in (0, 1)$ .







## Computing the Sizes of Constrained Subcodes of General Linear Codes

- Using a trick from Boolean Fourier analysis, we convert the problem into a question on the structure of the dual code:
  - $N(\mathcal{C};\mathcal{A}) = \sum_{\mathsf{x} \in \{0,1\}^n} \mathbb{1}\{\mathsf{x} \in \mathcal{A}\} \cdot \mathbb{1}\{\mathsf{x} \in \mathcal{C}\} \stackrel{(Plancherel's)}{=} 2^n \cdot \sum_{\mathsf{s} \in \{0,1\}^n} \widehat{\mathbb{1}_{\mathcal{A}}}(\mathsf{s}) \cdot \widehat{\mathbb{1}_{\mathcal{C}}}(\mathsf{s}) = |\mathcal{C}| \cdot \sum_{\mathsf{s} \in \mathcal{C}^{\perp}} \widehat{\mathbb{1}_{\mathcal{A}}}(\mathsf{s}). \qquad [\{0,1\}^n \supseteq \mathcal{A} \leftarrow \text{Set of constrained sequences}]$
- For many constraints, the Fourier transform  $\widehat{\mathbb{1}_{A}}$  is computable!

1.  $(d, \infty)$ -RLL constraint  $(S^d)$ : Theorem: For  $n \ge d + 2$  and for  $\mathbf{s} = (s_1, \dots, s_n) \in \{0, 1\}^n$ , it holds that  $\widehat{\mathbb{1}_{S^d}}^{(n)}(\mathbf{s}) = c_1 \cdot \widehat{\mathbb{1}_{S^d}}^{(n-1)}(s_2^n) + c_2(s_1) \cdot \widehat{\mathbb{1}_{S^d}}^{(n-d-1)}(s_{d+2}^n)$ , for  $c_1, c_2(s_1) \in \mathbb{R}$ .

2. Subblock composition constraint  $(C_z^p)$ : Theorem: For  $\mathbf{s} \in \{0, 1\}^n$  with  $\mathbf{s} = (\mathbf{s}_1 \mid \mathbf{s}_2 \mid \dots \mid \mathbf{s}_p)$ , we have that  $2^n \cdot \widehat{\mathbb{1}_{C_z^p}}(\mathbf{s}) = \prod_{\ell=1}^p \underbrace{\mathcal{K}_z^{(n/p)}}_{\text{Krawtchouk poly.}}(w(\mathbf{s}_\ell)).$ 

## 3. 2-charge constraint $(S_2)$ :

**Theorem:** There exists a vector space V such that for  $s \in V$ ,

$$\widehat{\mathbb{1}_{S_2}}(\mathbf{s}) = (-1)^{\delta(\mathbf{s})} \cdot 2^{\lfloor \frac{n}{2} \rfloor - n},$$
  
and  $\widehat{\mathbb{1}_{S_2}}(\mathbf{s}) = \mathbf{0}$ , otherwise, with  $\delta(\mathbf{s}) \in \{\mathbf{0}, \mathbf{1}\}.$ 

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Upper Bounds on the Sizes of Constrained Codes Over Adversarial Channels

▶ The bit-flip error-correcting capability (with zero-error) of (constrained) codes is determined by the minimum Hamming distance, d.

- ▶ We propose a version of Delsarte' linear program for constrained systems, for upper bounding the sizes of constrained codes with a prescribed *d*.
- ▶ Our bounds beat the state-of-the-art generalized sphere packing bounds of Fazeli, Vardy, and Yaakobi (2015).

			Our LP <b>Del</b> $(n, d; A)$ is given on the left, for any constraint represented by $A \subseteq \{0, 1\}^n$ .	d	$Del(n = 13, d; S_2)$	$GenSph(n = 13, d; S_2)$	Del(n=13,d)
maximize $f: \{0,1\}^n \to \mathbb{R}$ $\sum_{x \in \{0,1\}^n} f(x)$	(Obj′)		$\triangleright$ Delsarte's (unconstrained) LP is <b>Del(</b> $n, d$ <b>)</b> .	3	45.255	64	512
$x \in \{0,1\}^{\prime\prime}$			Our upper bound is precisely val( $Del(n, d; A)$ ) <sup>1/2</sup> .	4	45.255	64	292.571
$f(x) > 0  \forall x \in [0, 1]^n$	(D1)		<b>Del(</b> $n$ , $d$ ; $A$ ) can be "symmetrized" to yield an LP with much smaller numbers of variables+constraints (sometimes only polynomially many!)	5	22.627	64	64
$f(\mathbf{x}) \geq 0,  \forall \mathbf{x} \in \{0, 1\},$	(D1)			6	17.889	64	40
$f(\mathbf{s}) \geq 0, \ \forall \ \mathbf{s} \in \{0, 1\}^n,$	(D2)	D2) D3)		7	5.657	32	8
$f(\mathbf{x}) = 0, \text{ if } 1 \leq w(\mathbf{x}) \leq d-1,$	(D3)			8	4.619	32	5.333
$f(0^n) \leq \operatorname{val}(\operatorname{Del}(n, d)),$	(D4)		It holds that	9	2.828	16	3.333
$f(x) \leq 2^n \cdot (\mathbb{1}_{\mathcal{A}} \star \mathbb{1}_{\mathcal{A}})(x), \ \forall \ x \in \{0,1\}^n$	'. (D5)	5)	$val(Del(n,d;\mathcal{A}))^{1/2} \leq \min\{ \mathcal{A} , val(Del(n,d))\}.$	10	2.619	16	2.857

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