

# Binary Subspace Chirps

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# Motivation/Applications

- Machine-type wireless communication.
  - Signature coding.
  - Unsourced random access.
- Compressive Sensing.
  - Construction of deterministic compressive sensing matrices.
  - Fast decoding/reconstruction algorithms.
- Quantum Computing.
  - Clifford Group/Hierarchy.
  - Stabilizer States.
- Coding Theory.
  - Reed-Muller codes.
  - Rank-metric codes.
    - Kerdock codes.
    - Delsarte-Goethals codes.

# Binary Chirps

- Fix  $m \in \mathbb{N}$  and  $\mathbf{S} \in \text{Sym}(m)$  binary symmetric.
- Put  $N = 2^m$ .  $\mathbb{C}^N$  is indexed with  $\mathbb{F}_2^m$  (all vectors, complex or binary, are column vectors).
- Define a unitary matrix  $\mathbf{U}_{\mathbf{S}} \in \mathbb{C}^{N \times N}$  as

$$\mathbf{U}_{\mathbf{S}}(\mathbf{a}, \mathbf{b}) = \frac{1}{\sqrt{N}} j^{\mathbf{a}^t \mathbf{S} \mathbf{a} + 2\mathbf{b}^t \mathbf{a} \bmod 4}.$$

- A *binary chirp* (BC) is a column  $\mathbf{U}_{\mathbf{S}, \mathbf{b}}$ .
  - There are  $2^m \cdot 2^{m(m+1)/2}$  BCs.
  - If  $\mathbf{S}$  has zero diagonal then  $\mathbf{U}_{\mathbf{S}, \mathbf{b}} \in \mathbb{R}^N$ .
  - There are  $2^m \cdot 2^{m(m-1)/2}$  real BCs.

# Reconstruction Algorithm for BCs<sup>1</sup>

**Problem:** Assume you are given and unknown BC  $\mathbf{w} = \mathbf{U}_{\mathbf{S}, \mathbf{b}} \in \mathbb{C}^N$ .  
How to find  $\mathbf{S}, \mathbf{b}$ ?

**Solution:** “Shift and multiply” technique<sup>1</sup>:

- For a shift  $\mathbf{e}$  compute

$$\begin{aligned}\mathbf{w}_{\mathbf{e}} &:= [\mathbf{w}(\mathbf{a} + \mathbf{e}) \overline{\mathbf{w}(\mathbf{a})}]_{\mathbf{a} \in \mathbb{F}_2^m} \in \mathbb{C}^N \\ &= \frac{1}{N} j^{\mathbf{e}^t \mathbf{S} \mathbf{e} + 2\mathbf{b}^t \mathbf{e} \bmod 4} \cdot [(-1)^{\mathbf{e}^t \mathbf{S} \mathbf{a}}]_{\mathbf{a} \in \mathbb{F}_2^m}.\end{aligned}$$

- The Walsh-Hadamard transform is

$$\mathbf{H}_N = \frac{1}{\sqrt{N}} [(-1)^{\mathbf{b}^t \mathbf{a}}]_{\mathbf{a}, \mathbf{b}} = \frac{1}{\sqrt{N}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}^{\otimes m}.$$

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<sup>1</sup>S. D. Howard, A. R. Calderbank, and S. J. Searle, “A fast reconstruction algorithm for deterministic compressive sensing using second order Reed-Muller codes, 2008 42nd annual conference on information sciences and systems, 2008 March, pp. 11–15.

# Reconstruction Algorithm for BCs **Continued**

- For a basis vector  $\mathbf{e}_i$  one has

$$|(\mathbf{H}_N \mathbf{w}_{\mathbf{e}_i})(\mathbf{a})| = \begin{cases} 1, & \text{if } \mathbf{a} = \mathbf{S}\mathbf{e}_i, \\ 0, & \text{else.} \end{cases}$$

- After  $m$  shifts one recovers  $\mathbf{S}$ .
- To recover  $\mathbf{b}$  one computes

$$[\mathbf{w}(\mathbf{a}) \overline{\mathbf{U}_{\mathbf{S},0}(\mathbf{a})}]_{\mathbf{a} \in \mathbb{F}_2^m},$$

and then applies  $\mathbf{H}_N$ .

## Reconstruction Algorithm for BCs: Example

- Let  $m = 3$  and consider  $\mathbf{S} = \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix}$  and  $\mathbf{b} = \mathbf{0}$ .
- Corresponding BC  $\mathbf{w} = \mathbf{w}_{\mathbf{S},0}$  is  $+++ - + - -$ .
- The three shifted version  $\mathbf{w}_i$ , and their Walsh-Hadamard transform are, the rows/columns of  $\mathbf{S}$  are

$\mathbf{w}_i$	$\mathbf{H}_8 \mathbf{w}_i$	$\mathbf{S}_i$
$++-+-+--$	00010000	4 = 011
$+ - + + - - + -$	00000100	6 = 101
$+ - - - + + + -$	00000010	7 = 110

- Minor problems arise if  $\mathbf{S}$  has equal rows/columns.
- Algorithm works even in presence of noise.

# Multiple BC scenario

- **Problem:** What if you are given a linear combination of multiple BCs

$$\mathbf{w} = \sum_{l=1}^L c_l \mathbf{w}_l, c_l \in \mathbb{C},$$

where  $L$  is small. How to find  $\mathbf{S}_l, \mathbf{b}_l$ ?

- **Compressive sensing prospective:** Concatenate all matrices  $\mathbf{U}_S$  to a long  $2^m \times 2^m \cdot 2^{m(m+1)/2}$  matrix  $\Phi$ . Given

$$\mathbf{y} = \Phi \mathbf{x},$$

where  $\mathbf{x}$  is sparse, how to find  $\mathbf{x}$ ?

# Binary Subspace Chirps

- Fix a rank  $0 \leq r \leq m$  and  $\mathbf{P} \in \text{GL}(m)$ .
- $\mathbf{P}^{-t}$  denotes the inverse transpose.
- Define a unitary matrix  $\mathbf{U}_{\mathbf{P},\mathbf{S}} \in \mathbb{C}^{N \times N}$  as

$$\mathbf{U}_{\mathbf{P},\mathbf{S}}(\mathbf{a}, \mathbf{b}) = \frac{1}{\sqrt{2^r}} i^{(\mathbf{P}^{-1}\mathbf{a})^t \mathbf{S}(\mathbf{P}^{-1}\mathbf{a}) + 2\mathbf{b}^t(\mathbf{P}^{-1}\mathbf{a}) \bmod 4} \cdot f(\mathbf{b}, \mathbf{P}^{-1}\mathbf{a}, r),$$

where

$$f(\mathbf{x}, \mathbf{y}, r) = \prod_{i=r+1}^m (1 + x_i + y_i).$$

- A *binary subspace chirp* (BSC) is a column  $\mathbf{U}_{\mathbf{P},\mathbf{S},\mathbf{b}}$ .
- **Note:** Not all choices of  $\mathbf{P}, \mathbf{S}$  give different BSCs.



# Parametrization of BSCs

**Theorem:** A rank  $r$  BSC is characterized by  $H \in \mathcal{G}(m, r)$  and  $\mathbf{S}_r \in \text{Sym}(r)$ .

- Write  $H = \text{cs}(\mathbf{H}_{\mathcal{I}})$  where  $\mathbf{H}_{\mathcal{I}}$  is in CREF and  $\mathcal{I}$  is the set of pivots. Then put

$$\mathbf{P} = \mathbf{P}_H = [\mathbf{H}_{\mathcal{I}} \quad \mathbf{I}_{\tilde{\mathcal{I}}}]$$

- **Note:**  $\mathbf{P}^{-t} = [\mathbf{I}_{\mathcal{I}} \quad \widetilde{\mathbf{H}}_{\mathcal{I}}]$ , where  $(\mathbf{H}_{\mathcal{I}})^t \widetilde{\mathbf{H}}_{\mathcal{I}} = \mathbf{0}$ .
- For  $\mathbf{S}_r \in \text{Sym}(r)$  we will denote

$$\widetilde{\mathbf{S}}_r = \begin{bmatrix} \mathbf{S}_r & \mathbf{0} \\ \mathbf{0} & \mathbf{0} \end{bmatrix}$$

## Example $m = 5, r = 3$

$$\mathbf{P} = \left[ \begin{array}{ccc|cc} \mathbf{1} & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & \mathbf{1} & 0 \\ 0 & \mathbf{1} & 0 & 0 & 0 \\ 0 & 0 & \mathbf{1} & 0 & 0 \\ 1 & 1 & 0 & 0 & \mathbf{1} \end{array} \right], \mathbf{P}^{-t} = \left[ \begin{array}{ccc|cc} \mathbf{1} & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & \mathbf{1} & 0 \\ 0 & \mathbf{1} & 0 & 0 & 1 \\ 0 & 0 & \mathbf{1} & 0 & 0 \\ 0 & 0 & 0 & 0 & \mathbf{1} \end{array} \right]$$
$$\tilde{\mathbf{S}}_3 = \left[ \begin{array}{ccccc} 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right], \mathbf{P}^{-t} \tilde{\mathbf{S}}_3 \mathbf{P}^{-1} = \left[ \begin{array}{ccccc} 0 & \mathbf{0} & 1 & 1 & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} \\ 1 & \mathbf{0} & 0 & 1 & \mathbf{0} \\ 1 & \mathbf{0} & 1 & 0 & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} \end{array} \right]$$

Example  $m = 5, r = 1$

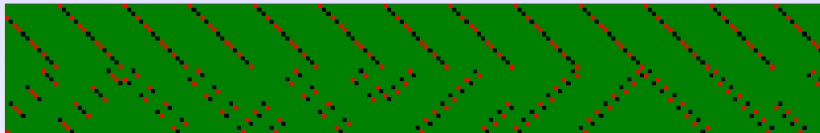


Figure: Rank 1 Real BCs: 1 Red, -1 Black, 0 Green.

# Properties of BSCs

- A rank  $r$  BSC has  $2^r$  nonzero entries.
- The nonzero entries are precisely the BCs in the respective dimensions.
- The total number of BSCs is

$$2^m \cdot \sum_{r=0}^m 2^{r(r+1)/2} \binom{m}{r}_2 = 2^m \cdot \prod_{r=1}^m (2^r + 1).$$

- $|\text{BSC}|/|\text{BC}| \rightarrow 2.384\dots$

# Reconstruction algorithm for BSC

- **Problem:** How to recover  $r, H, \mathbf{S}_r$  for an unknown BSC  $\mathbf{w}$ ?
- First recover  $r$  and  $H$ :  $\mathbf{H}_N[\mathbf{w}(\mathbf{a})\overline{\mathbf{w}(\mathbf{a})}]_{\mathbf{a}} \neq 0$  iff

$$\mathbf{a} \in \left\{ \mathbf{I}_{\tilde{\mathcal{I}}} \mathbf{b}_{m-r} + \mathbf{H}_{\mathcal{I}} \mathbf{x} \mid \mathbf{x} \in \mathbb{F}_2^r \right\}.$$

- To recover rows  $\mathbf{S}_r$  use “shift and multiply” where instead of shifting with the basis vectors  $\mathbf{e}_i$  one shifts with columns of  $\mathbf{H}_{\mathcal{I}}$ .
- The column  $\mathbf{b}$  is recovered exactly as for BCs.
- Complexity  $\mathcal{O}(N \log N)$  (same as for BC reconstruction!).

# Algebraic structure of BSCs

- *Pauli matrices*

$$\mathbf{I}_2, \quad \sigma_x = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \quad \sigma_z = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \quad \sigma_y = i\sigma_x\sigma_z.$$

- *m-qubit Pauli group*

$$\mathcal{P}_m = \langle e_1 \otimes \cdots \otimes e_m \mid e_i \text{ Pauli matrix} \rangle \subset \mathbb{U}(N).$$

- The *m-qubit Clifford group* is the normalizer of  $\mathcal{P}_m$  in  $\mathbb{U}(N)$ .
- BSCs are columns of Clifford matrices, also known as *stabilizer states*.

- Fix a BC  $\mathbf{w}$  which is a column of  $\mathbf{U}_S$ . If  $\mathbf{v}$  is another BC that runs through the columns of  $\mathbf{U}_{S'}$ , then

$$|\langle \mathbf{v} | \mathbf{w} \rangle|^2 = \begin{cases} 1/2^\ell, & 2^\ell \text{ times,} \\ 0, & 2^m - 2^\ell \text{ times,} \end{cases}$$

where  $\ell = \text{rank}(\mathbf{S} + \mathbf{S}')$ .

- If  $\ell = m$ ,  $\mathbf{v}$  and  $\mathbf{w}$  are called *mutually unbiased*.
- A vector space of invertible matrices contains at most  $2^m$  matrices.
  - **Corollary:** This can be used show that there exist precisely  $2^m + 1$  *mutually unbiased bases* in  $2^m$  dimensions.
- $\text{Sym}(m)$  is a disjoint union of  $2^{m(m-1)/2}$  MRD (of type  $[2^m, m]$ ) codes.

Thank You!