

# Polynomial-Complexity Computation of the *M*-phase Vector that Maximizes a Rank-Deficient Quadratic Form

A.T. Kyrillidis and G.N. Karystinos

Telecommunications Division  
Department of Electronic & Computer Engineering  
Technical University of Crete

November 13, 2010

- Problem Presentation
- Previous Work and Applications
- Problem Reformulation
- Proposed Algorithm
- Results
- Conclusion and Further Work

# Quadratic Form Maximization over the $M$ -ary field

- Optimization problem:

$$\mathbf{s}_{\text{opt}} \triangleq \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \mathbf{s}^H \mathbf{Q} \mathbf{s}$$

where

- $\mathbf{Q} \in \mathbb{C}^{N \times N}$ : positive (semi)definite matrix.
  - $\mathbf{s} \in \mathcal{A}_M^N$ : a  $M$ -PSK  $N$ -tuple vector argument.
  - $\mathcal{A}_M = \left\{ e^{j\frac{2\pi m}{M}} \mid m = 0, 1, \dots, M-1 \right\}$ :  $M$ -phase alphabet.
  - $M \in \{2^k \mid k = 1, 2, \dots\}$ .
- 
- Rank of observation matrix  $\mathbf{Q}$ :  $D \leq N$ .

# Quadratic Form Maximization over the $M$ -ary field

- Optimization problem:

$$\mathbf{s}_{\text{opt}} \triangleq \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \mathbf{s}^H \mathbf{Q} \mathbf{s}$$

where

- $\mathbf{Q} \in \mathbb{C}^{N \times N}$ : positive (semi)definite matrix.
- $\mathbf{s} \in \mathcal{A}_M^N$ : a  $M$ -PSK  $N$ -tuple vector argument.
- $\mathcal{A}_M = \left\{ e^{\frac{j2\pi m}{M}} \mid m = 0, 1, \dots, M-1 \right\}$ :  $M$ -phase alphabet.
- $M \in \{2^k \mid k = 1, 2, \dots\}$ .
- Rank of observation matrix  $\mathbf{Q}$ :  $D \leq N$ .
- Full rank case  $\rightarrow$  known  $\mathcal{NP}$ -hard problem with an obvious approach: *Exhaustive search*.
- Complexity:  $\mathcal{O}(M^N)$ .
- We focus on: rank-deficient case.

- Optimal Approaches - Computational Geometry (CG) Algorithms:
  - Assume rank-deficiency of  $\mathbf{Q}$ .
  - Reverse Search Method [D. Avis et. al.:1996] - only for BPSK and QPSK vector argument.
  - Incremental Algorithm [H.Edelsbrunner et. al.: 1986] - applied for arbitrary  $M$ .
  - A fast algorithm for multiple-symbol differential detection of MPSK [K.M. Mackenthun Jr.: 1994].
- Suboptimal Approaches - Relaxation Algorithms:
  - No assumption of rank-deficiency of  $\mathbf{Q}$ .
  - SemiDefinite Relaxation (SDR) method [S. Zhang and Y. Huang: 2006] - approximation algorithm with randomization (most of the times).
- Our work is based on:
  - Quadratic Form Maximization over the binary field  $\{\pm 1\}$  [G. Karystinos and A. Liavas: 2010].

- A common optimization problem in digital communications.
- Examples:
  - ML block noncoherent MPSK detection in SIMO systems [D.S. Papailiopoulos and G.N. Karystinos: 2008].
  - MLSD noncoherent  $M$ -PSK detection [I. Motedayen et. al.: 2007, V. Pauli et. al.: 2008].
  - Multiuser detection in  $M$ -PSK CDMA systems:
    - [G. Manglani and A.K. Chaturvedi: 2006, W.-K. Ma et. al.: 2002] - SDR over binary field.
    - [P. Tan and L.K. Rasmussen: 2001] - SDR over BPSK and QPSK.
    - [W.-K. Ma et. al.: 2004] - SDR over  $M$ -ary PSK CDMA.
  - Blind ML detection of orthogonal space-time block codes (OSTBC) [W.-K. Ma et.al.: 2006]
  - QAM and PSK codebooks for limited MIMO beamforming [D.J. Ryan et. al.: 2009].
  - etc.

# Quadratic Form Maximization over the $M$ -ary field

- Optimization problem:

$$\mathbf{s}_{\text{opt}} \triangleq \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \mathbf{s}^H \mathbf{Q} \mathbf{s}.$$

- We will show that if  $\text{rank}(\mathbf{Q}) = D$ ,  $D < N$  and  $D$  is independent of  $N$ , then:

# Quadratic Form Maximization over the $M$ -ary field

- Optimization problem:

$$\mathbf{s}_{\text{opt}} \triangleq \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \mathbf{s}^H \mathbf{Q} \mathbf{s}.$$

- We will show that if  $\text{rank}(\mathbf{Q}) = D$ ,  $D < N$  and  $D$  is independent of  $N$ , then:
  - Exhaustive search is not necessary.



# Quadratic Form Maximization over the $M$ -ary field

- Optimization problem:

$$\mathbf{s}_{\text{opt}} \triangleq \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \mathbf{s}^H \mathbf{Q} \mathbf{s}.$$

- We will show that if  $\text{rank}(\mathbf{Q}) = D$ ,  $D < N$  and  $D$  is independent of  $N$ , then:
  - Exhaustive search is not necessary.
  - $\mathbf{s}_{\text{opt}}$  belongs into a polynomial-size set of candidates.

# Quadratic Form Maximization over the $M$ -ary field

- Optimization problem:

$$\mathbf{s}_{\text{opt}} \triangleq \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \mathbf{s}^H \mathbf{Q} \mathbf{s}.$$

- We will show that if  $\text{rank}(\mathbf{Q}) = D$ ,  $D < N$  and  $D$  is independent of  $N$ , then:
  - Exhaustive search is not necessary.
  - $\mathbf{s}_{\text{opt}}$  belongs into a polynomial-size set of candidates.
  - $\mathbf{s}_{\text{opt}}$  can be obtained in polynomial time  $\mathcal{O}((\frac{MN}{2})^{2D})$ .

- Let  $\mathbf{V} \triangleq \begin{bmatrix} \mathbf{v}_1 & \mathbf{v}_2 & \dots & \mathbf{v}_D \end{bmatrix} \in \mathbb{C}^{N \times D}$  such that:

$$\mathbf{Q} = \mathbf{V}\mathbf{V}^H.$$

- $\text{Rank}(\mathbf{V}) = D$ .

- Let  $\mathbf{V} \triangleq [\mathbf{v}_1 \ \mathbf{v}_2 \ \dots \ \mathbf{v}_D] \in \mathbb{C}^{N \times D}$  such that:

$$\mathbf{Q} = \mathbf{V}\mathbf{V}^H.$$

- $\text{Rank}(\mathbf{V}) = D$ .
- Optimal solution:

$$\begin{aligned}\mathbf{s}_{\text{opt}} &= \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \mathbf{s}^H \mathbf{Q} \mathbf{s} \\ &= \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \mathbf{s}^H \mathbf{V} \mathbf{V}^H \mathbf{s} \\ &= \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \|\mathbf{V}^H \mathbf{s}\|^2 = \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \|\mathbf{V}^H \mathbf{s}\|.\end{aligned}$$

## Problem Reformulation

- Let  $\phi_{i:j} \triangleq [\phi_i, \phi_{i+1}, \dots, \phi_j]^T$ . We introduce:
  - (i)  $2D - 1$  auxiliary hyperspherical coordinates:

$$\phi_{1:2D-1} \in \left(-\frac{\pi}{2}, \frac{\pi}{2}\right]^{2D-2} \times (-\pi, \pi].$$

## Problem Reformulation

- Let  $\phi_{i:j} \triangleq [\phi_i, \phi_{i+1}, \dots, \phi_j]^T$ . We introduce:
  - (i)  $2D - 1$  auxiliary hyperspherical coordinates:

$$\phi_{1:2D-1} \in \left(-\frac{\pi}{2}, \frac{\pi}{2}\right]^{2D-2} \times (-\pi, \pi].$$

- (ii) the hyperspherical real vector:

$$\tilde{\mathbf{c}}(\phi_{1:2D-1}) \triangleq \begin{bmatrix} \sin \phi_1 \\ \cos \phi_1 \sin \phi_2 \\ \cos \phi_1 \cos \phi_2 \sin \phi_3 \\ \vdots \\ \left[ \prod_{i=1}^{2D-2} \cos \phi_i \right] \sin \phi_{2D-1} \\ \left[ \prod_{i=1}^{2D-2} \cos \phi_i \right] \cos \phi_{2D-1} \end{bmatrix}_{2D \times 1}$$

## Problem Reformulation

- Let  $\phi_{i:j} \triangleq [\phi_i, \phi_{i+1}, \dots, \phi_j]^T$ . We introduce:
  - $2D - 1$  auxiliary hyperspherical coordinates:

$$\phi_{1:2D-1} \in \left(-\frac{\pi}{2}, \frac{\pi}{2}\right]^{2D-2} \times (-\pi, \pi].$$

- the hyperspherical real vector:

$$\tilde{\mathbf{c}}(\phi_{1:2D-1}) \triangleq \begin{bmatrix} \sin \phi_1 \\ \cos \phi_1 \sin \phi_2 \\ \cos \phi_1 \cos \phi_2 \sin \phi_3 \\ \vdots \\ \left[ \prod_{i=1}^{2D-2} \cos \phi_i \right] \sin \phi_{2D-1} \\ \left[ \prod_{i=1}^{2D-2} \cos \phi_i \right] \cos \phi_{2D-1} \end{bmatrix}_{2D \times 1}$$

- the  $D \times 1$  hyperspherical complex vector:

$$\begin{aligned} \mathbf{c}(\phi_{1:2D-1}) &\triangleq \tilde{\mathbf{c}}_{2:2:2D}(\phi_{1:2D-1}) + j\tilde{\mathbf{c}}_{1:2:2D-1}(\phi_{1:2D-1}) \\ &= \begin{bmatrix} \cos \phi_1 \sin \phi_2 + j \sin \phi_1 \\ \cos \phi_1 \cos \phi_2 \cos \phi_3 \sin \phi_4 + j \cos \phi_1 \cos \phi_2 \sin \phi_3 \\ \vdots \\ \left[ \prod_{i=1}^{2D-2} \cos \phi_i \right] \cos \phi_{2D-1} + j \left[ \prod_{i=1}^{2D-2} \cos \phi_i \right] \sin \phi_{2D-1} \end{bmatrix} = \begin{bmatrix} c_1(\phi_{1:2}) \\ c_2(\phi_{1:4}) \\ \vdots \\ c_D(\phi_{1:2D-1}) \end{bmatrix}_{D \times 1} \end{aligned}$$

- Using the following inequalities  $\forall \mathbf{a} \in \mathbb{C}^D, \forall \hat{\theta} \in (-\pi, \pi]$  and  $\forall \phi_{2D-1} \in (-\frac{\pi}{2}, \frac{\pi}{2}]^{2D-2} \times (-\pi, \pi]$ :

$$\Re\left\{\mathbf{a}^H \mathbf{c}(\phi_{1:2D-1}) e^{-j\hat{\theta}}\right\} \stackrel{(2)}{\leq} \left|\mathbf{a}^H \mathbf{c}(\phi_{1:2D-1})\right| \stackrel{(1)}{\leq} \|\mathbf{a}\| \underbrace{\|\mathbf{c}(\phi_{1:2D-1})\|}_{=1} = \|\mathbf{a}\|.$$



# Problem Reformulation

- Using the following inequalities  $\forall \mathbf{a} \in \mathbb{C}^D, \forall \hat{\theta} \in (-\pi, \pi]$  and  $\forall \phi_{1:2D-1} \in (-\frac{\pi}{2}, \frac{\pi}{2}]^{2D-2} \times (-\pi, \pi]$ :

$$\Re\left\{\mathbf{a}^H \mathbf{c}(\phi_{1:2D-1}) e^{-j\hat{\theta}}\right\} \stackrel{(2)}{\leq} \left|\mathbf{a}^H \mathbf{c}(\phi_{1:2D-1})\right| \stackrel{(1)}{\leq} \|\mathbf{a}\| \underbrace{\|\mathbf{c}(\phi_{1:2D-1})\|}_{=1} = \|\mathbf{a}\|.$$

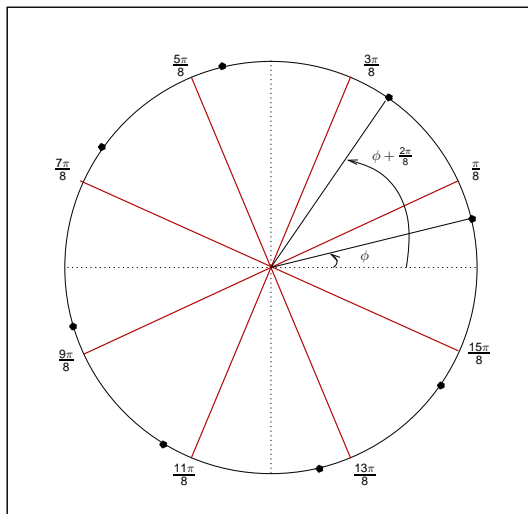
- The maximization problem becomes (substituting  $\mathbf{a} = \mathbf{V}^H \mathbf{s}$  and applying some computations):

$$\begin{aligned} \mathbf{s}_{\text{opt}} &= \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \|\mathbf{V}^H \mathbf{s}\| \stackrel{(1)}{=} \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \max_{\phi_{1:2D-1} \in (-\frac{\pi}{2}, \frac{\pi}{2}]^{2D-2} \times (-\pi, \pi]} \left| \mathbf{s}^H \mathbf{V} \mathbf{c}(\phi_{1:2D-1}) \right| \\ &\stackrel{(2)}{=} \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \max_{\phi_{1:2D-1} \in (-\frac{\pi}{2}, \frac{\pi}{2}]^{2D-2} \times (-\pi, \pi]} \Re\left\{\mathbf{s}^H \mathbf{V} \mathbf{c}(\phi_{1:2D-1})\right\}. \end{aligned}$$

- Maximization over the whole range of  $\phi_{1:2D-1}$  generates  $M$ -ary equivalent duplicates of candidate vectors  $\mathbf{s}$ .
- We observed and proved that:
  - We can divide the space  $(-\frac{\pi}{2}, \frac{\pi}{2}]^{2D-2} \times (-\pi, \pi]$  into  $M$  equivalent subspaces.
  - The division is performed by separating the range of values of  $\phi_{2D-1}$  into  $M$  consecutive parts.
  - E.g.  $\phi_{2D-1} \in \left\{ \left( -\frac{\pi}{M}, \frac{\pi}{M} \right], \left( \frac{\pi}{M}, \frac{3\pi}{M} \right], \dots, \left( \frac{(2M-3)\pi}{M}, \frac{(2M-1)\pi}{M} \right] \right\}$ .
- Without loss of generality, we choose  $\phi_{2D-1} \in (-\frac{\pi}{M}, \frac{\pi}{M}]$ .

# Problem Reformulation

- Illustrative example:  $\text{Rank}(\mathbf{Q}) = D = 1$ , arbitrary  $N$ ,  $M = 8$ .



- The optimization problem becomes:

$$\mathbf{s}_{\text{opt}} = \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \max_{\phi_{1:2D-1} \in (-\frac{\pi}{2}, \frac{\pi}{2}]^{2D-2} \times (-\frac{\pi}{M}, \frac{\pi}{M}]} \Re \left\{ \mathbf{s}^H \mathbf{V} \mathbf{c}(\phi_{1:2D-1}) \right\}.$$

- The optimization problem becomes:

$$\mathbf{s}_{\text{opt}} = \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \max_{\phi_{1:2D-1} \in (-\frac{\pi}{2}, \frac{\pi}{2}]^{2D-2} \times (-\frac{\pi}{M}, \frac{\pi}{M}]} \Re \left\{ \mathbf{s}^H \mathbf{V} \mathbf{c}(\phi_{1:2D-1}) \right\}.$$

- Interchanging maximizations:

$$\max_{\phi_{1:2D-1} \in \Phi^{2D-2} \times (-\frac{\pi}{M}, \frac{\pi}{M}]} \sum_{n=1}^N \max_{s_n \in \mathcal{A}_M} \Re \left\{ s_n^* \mathbf{V}_{n,1:D} \mathbf{c}(\phi_{1:2D-1}) \right\}, \quad \Phi \triangleq \left( -\frac{\pi}{2}, \frac{\pi}{2} \right].$$

## Decision Functions and Candidate Set $\mathcal{S}(\mathbf{V}_{N \times D})$

- Element-by-element decision rule for fixed set of  $\phi_{1:2D-1}$ :

$$s(\mathbf{v}^T; \phi_{1:2D-1}) \triangleq \arg \max_{s \in \mathcal{A}_M} \Re\{s^* \mathbf{v}^T \mathbf{c}(\phi_{1:2D-1})\}.$$

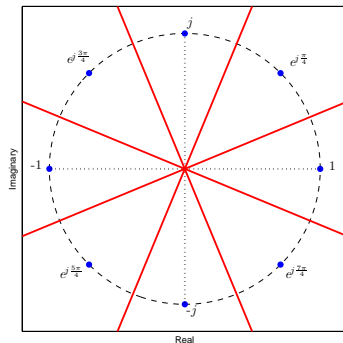
- Sequence decision rule for fixed set of  $\phi_{1:2D-1}$ :

$$\begin{aligned} \mathbf{s}(\mathbf{V}_{N \times D}; \phi_{1:2D-1}) &\triangleq \begin{bmatrix} s(\mathbf{V}_{1,1:D}; \phi_{1:2D-1}) \\ s(\mathbf{V}_{2,1:D}; \phi_{1:2D-1}) \\ \vdots \\ s(\mathbf{V}_{N,1:D}; \phi_{1:2D-1}) \end{bmatrix} \\ &= \arg \max_{\mathbf{s} \in \mathcal{A}_M^N} \Re\{\mathbf{s}^H \mathbf{V} \mathbf{c}(\phi_{1:2D-1})\}. \end{aligned}$$

- Computing  $\mathbf{s}(\mathbf{V}_{N \times D}; \phi_{1:2D-1})$  for  $\forall \phi_{1:2D-1} \in \Phi^{2D-2} \times (-\frac{\pi}{M}, \frac{\pi}{M}]$ :

$$\mathcal{S}(\mathbf{V}_{N \times D}) \triangleq \bigcup_{\phi_{1:2D-1} \in \Phi^{2D-2} \times (-\frac{\pi}{M}, \frac{\pi}{M}]} \left\{ \mathbf{s}(\mathbf{V}_{N \times D}; \phi_{1:2D-1}) \right\} \subseteq \mathcal{A}_M^N.$$

# Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

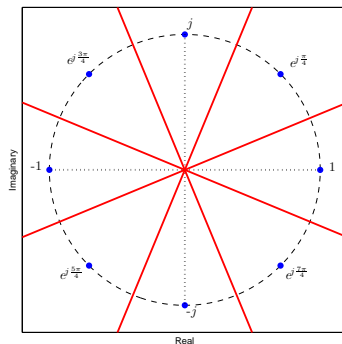


- We rewrite the previous expression:

$$\max_{\phi_{1:2D-1} \in \Phi^{2D-2} \times (-\frac{\pi}{M}, \frac{\pi}{M}]} \sum_{n=1}^N \max_{s_n \in \mathcal{A}_M} \Re \left\{ s_n^* \mathbf{V}_{n,1:D} \mathbf{c}(\phi_{1:2D-1}) \right\}, \quad \Phi \triangleq \left( -\frac{\pi}{2}, \frac{\pi}{2} \right].$$

- As  $\phi_{1:2D-1}$  vary, the decision in favor of  $s_n$  is maintained as long as a decision boundary is not crossed.

# Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$



- The  $\frac{M}{2}$  decision boundaries for the determination of  $s_n$  are given by

$$\mathbf{V}_{n,1:D} \mathbf{C}(\phi_{1:2D-1}) = A e^{j\pi \frac{2k+1}{M}}, \quad A \in \mathbb{R}, \quad k = 0, 1, \dots, \frac{M}{2} - 1,$$

or equivalently

$$\Im \left\{ e^{-j\pi \frac{2k+1}{M}} \mathbf{V}_{n,1:D} \mathbf{C}(\phi_{1:2D-1}) \right\} = 0, \quad k = 0, 1, \dots, \frac{M}{2} - 1.$$



# Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

- In matrix form:

$$\Im \left( \underbrace{\begin{bmatrix} e^{-j\frac{\pi}{M}} \mathbf{V}_{N \times D} \\ \vdots \\ e^{-j\pi \frac{M-1}{M}} \mathbf{V}_{N \times D} \end{bmatrix}}_{\hat{\mathbf{V}}_{\frac{MN}{2} \times D}} \mathbf{c}(\phi_{1:2D-1}) \right) = \mathbf{0}_{\frac{MN}{2} \times 1} \Rightarrow \dots$$

$$\tilde{\mathbf{V}}_{l,1:2D} \tilde{\mathbf{c}}(\phi_{1:2D-1}) = 0, \quad l = 1, 2, \dots, \frac{MN}{2}$$

where

$$\tilde{\mathbf{V}}_{:,1:2D-1} = \Re(\hat{\mathbf{V}}) \quad \text{and} \quad \tilde{\mathbf{V}}_{:,2:2D} = \Im(\hat{\mathbf{V}}).$$

# Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

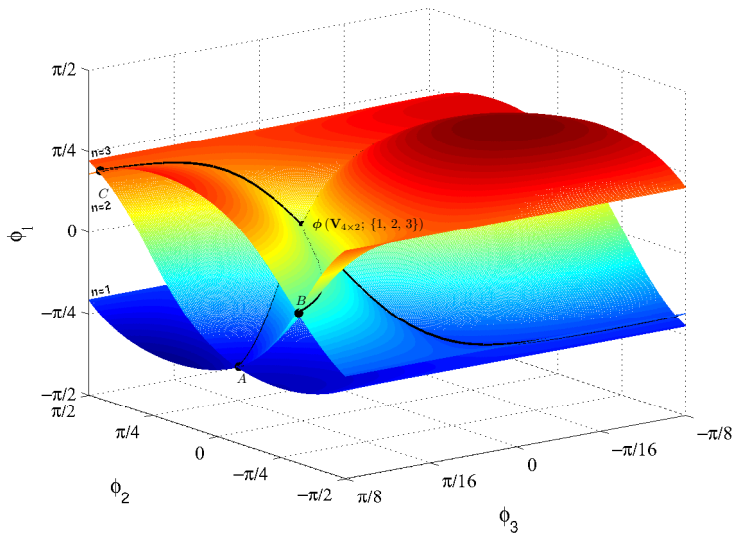
- From the construction of  $\tilde{\mathbf{V}}_{\frac{MN}{2} \times 2D}$ , we observe:
  - each row of  $\mathbf{V}$  is rotated by each of the  $\frac{M}{2}$  exponentials  $e^{-j\pi \frac{2k+1}{M}}$ .
  - We can define  $N$  different groups  $\mathcal{G}^{(n)}$ ,  $n \in \{1, 2, \dots, N\}$ :

$$\mathcal{G}^{(n)} \triangleq \begin{bmatrix} \tilde{\mathbf{V}}_{n,1:2D} \\ \tilde{\mathbf{V}}_{(n+N),1:2D} \\ \vdots \\ \tilde{\mathbf{V}}_{(n+(\frac{M}{2}-1)N),1:2D} \end{bmatrix}_{\frac{M}{2} \times 2D}$$

- The set of  $\frac{MN}{2}$  hypersurfaces  $\mathcal{H} = \{\mathcal{H}(\tilde{\mathbf{V}}_{1,1:2D}), \mathcal{H}(\tilde{\mathbf{V}}_{2,1:2D}), \dots, \mathcal{H}(\tilde{\mathbf{V}}_{\frac{MN}{2},1:2D})\}$ :
  - Separate the  $\Phi^{2D-2} \times (-\frac{\pi}{M}, \frac{\pi}{M}]$  space into distinct regions.
  - Each of which is associated with a different  $M$ -PSK candidate vector  $\mathbf{s}$ .

# Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

- Illustrative example:  $D = 2$ ,  $N = 4$  and  $M = 8$ .



# Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

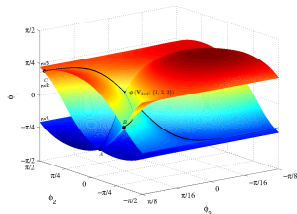
- Let  $\mathcal{H} = \{\mathcal{H}(\tilde{\mathbf{V}}_{1,1:2D}), \mathcal{H}(\tilde{\mathbf{V}}_{2,1:2D}), \dots, \mathcal{H}(\tilde{\mathbf{V}}_{\frac{MN}{2},1:2D})\}$  the set of  $\frac{MN}{2}$  hypersurfaces.

## Proposition 2a

Each subset of  $\mathcal{H}$  that consists of  $2D - 1$  hypersurfaces has either a single or uncountably many intersections in  $\Phi^{2D-2} \times (-\frac{\pi}{M}, \frac{\pi}{M}]$ .

## Proposition 2b

Each combination of  $2D - 1$  hypersurfaces from the set  $\mathcal{H}$  has a unique intersection point that constitutes a vertex of a cell if and only if no more than two hypersurfaces originate from the same row of the matrix  $\mathbf{V}$ .



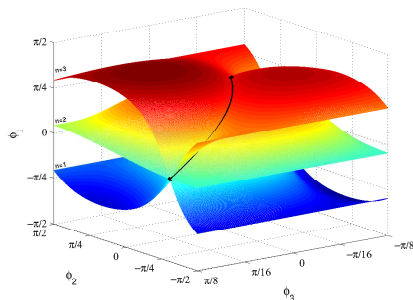
## Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

- Let  $\mathcal{I}_{2D-1} \triangleq \{i_1, i_2, \dots, i_{2D-1}\} \subset \{1, 2, \dots, \frac{MN}{2}\}$  be the indices of  $2D - 1$  hypersurfaces  $\mathcal{H}(\tilde{\mathbf{V}}_{i_1, 1:2D}), \mathcal{H}(\tilde{\mathbf{V}}_{i_2, 1:2D}), \dots, \mathcal{H}(\tilde{\mathbf{V}}_{i_{2D-1}, 1:2D})$ . We detect the following cases:
  - (a) At most two surfaces originate from the same row of  $\mathbf{V}$ .
  - (b) At least three surfaces originate from the same row of  $\mathbf{V}$ .

# Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

- Let  $\mathcal{I}_{2D-1} \triangleq \{i_1, i_2, \dots, i_{2D-1}\} \subset \{1, 2, \dots, \frac{MN}{2}\}$  be the indices of  $2D - 1$  hypersurfaces  $\mathcal{H}(\tilde{\mathbf{v}}_{i_1, 1:2D}), \mathcal{H}(\tilde{\mathbf{v}}_{i_2, 1:2D}), \dots, \mathcal{H}(\tilde{\mathbf{v}}_{i_{2D-1}, 1:2D})$ . We detect the following cases:
  - (a) At most two surfaces originate from the same row of  $\mathbf{V}$ .
  - (b) At least three surfaces originate from the same row of  $\mathbf{V}$ .
- Combinations of the form (b) do not have a unique intersection point.

- Illustrative example:  $D = 2, N = 4$  and  $M = 8$ .
- $2D - 1$  hypersurfaces originating from the same row of  $\mathbf{V}$ .



## Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

- Let  $\mathcal{I}_{2D-1} \triangleq \{i_1, i_2, \dots, i_{2D-1}\} \subset \{1, 2, \dots, \frac{MN}{2}\}$  be the indices of  $2D - 1$  hypersurfaces  $\mathcal{H}(\tilde{\mathbf{V}}_{i_1, 1:2D}), \mathcal{H}(\tilde{\mathbf{V}}_{i_2, 1:2D}), \dots, \mathcal{H}(\tilde{\mathbf{V}}_{i_{2D-1}, 1:2D})$ . We detect the following cases:
  - (a) At most two surfaces originate from the same row of  $\mathbf{V}$ .
  - (b) At least three surfaces originate from the same row of  $\mathbf{V}$ .

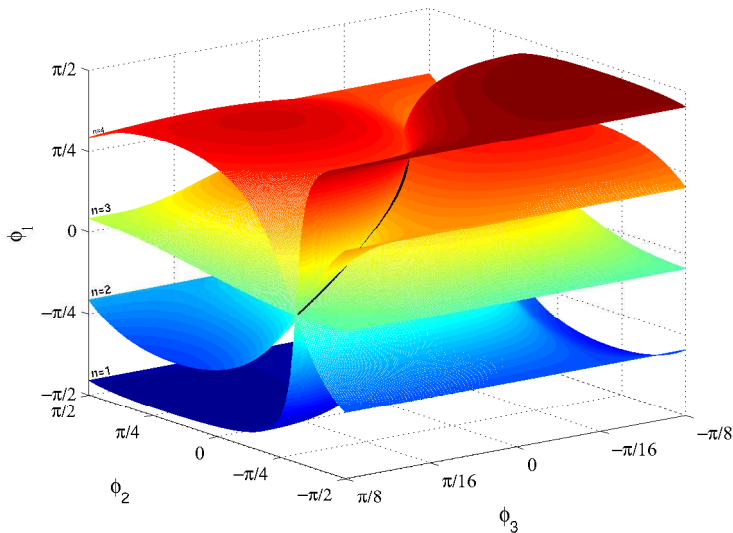
### Corollary

All  $\frac{M}{2}$  hypersurfaces originating from the same row of  $\mathbf{V}$  intersect to a common axis.

- The dimensionality of the common axis depends on the rank  $D$  of the observation matrix  $\mathbf{V}$  and equals to  $2(D - 1)$ .
  - E.g. common one-dimensional line if  $D = 2$ , common four-dimensional hyperplane if  $D = 3$ , etc.

# Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

- Illustrative example:  $D = 2$ ,  $N = 4$  and  $M = 8$ .





## Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

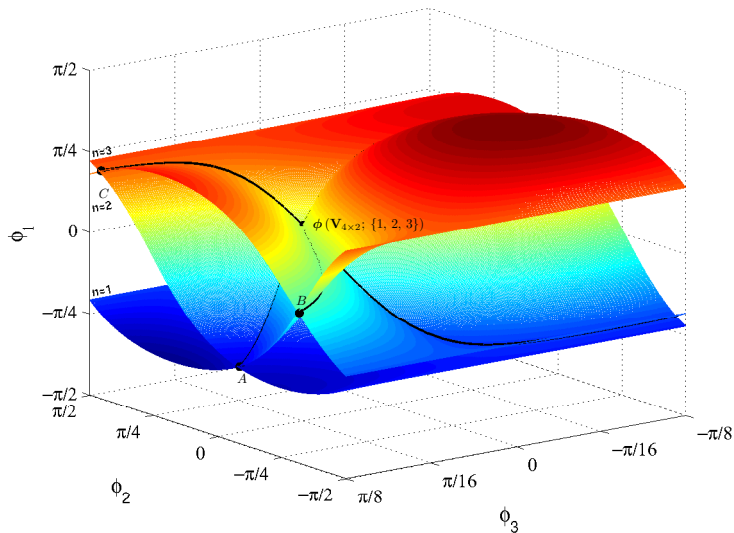
- Let  $\mathcal{I}_{2D-1} \triangleq \{i_1, i_2, \dots, i_{2D-1}\} \subset \{1, 2, \dots, \frac{MN}{2}\}$  be the indices of  $2D - 1$  hypersurfaces  $\mathcal{H}(\tilde{\mathbf{V}}_{i_1, 1:2D}), \mathcal{H}(\tilde{\mathbf{V}}_{i_2, 1:2D}), \dots, \mathcal{H}(\tilde{\mathbf{V}}_{i_{2D-1}, 1:2D})$ . We detect the following cases:
  - (a) At most two surfaces originate from the same row of  $\mathbf{V}$ .
  - (b) At least three surfaces originate from the same row of  $\mathbf{V}$ .
- Combinations of the form (a) have a unique intersection point  $\phi(\mathbf{V}_{N \times D}; \mathcal{I}_{2D-1}) \in \Phi^{2D-2} \times (-\frac{\pi}{M}, \frac{\pi}{M}]$  such that:
  - “Leads”  $\mathcal{Q}$  cells, say  $C_q(\mathbf{V}_{N \times D}; \mathcal{I}_{2D-1})$ , where  $q = 1, 2, \dots, \mathcal{Q}$  and  $\mathcal{Q} \in \left\{ (\frac{M}{2} - 1)^0, (\frac{M}{2} - 1)^1, \dots, (\frac{M}{2} - 1)^{D-1} \right\}$ .

## Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

- Let  $\mathcal{I}_{2D-1} \triangleq \{i_1, i_2, \dots, i_{2D-1}\} \subset \{1, 2, \dots, \frac{MN}{2}\}$  be the indices of  $2D - 1$  hypersurfaces  $\mathcal{H}(\tilde{\mathbf{V}}_{i_1, 1:2D}), \mathcal{H}(\tilde{\mathbf{V}}_{i_2, 1:2D}), \dots, \mathcal{H}(\tilde{\mathbf{V}}_{i_{2D-1}, 1:2D})$ . We detect the following cases:
  - (a) At most two surfaces originate from the same row of  $\mathbf{V}$ .
  - (b) At least three surfaces originate from the same row of  $\mathbf{V}$ .
- Combinations of the form (a) have a unique intersection point  $\phi(\mathbf{V}_{N \times D}; \mathcal{I}_{2D-1}) \in \Phi^{2D-2} \times (-\frac{\pi}{M}, \frac{\pi}{M}]$  such that:
  - “Leads”  $\mathcal{Q}$  cells, say  $C_q(\mathbf{V}_{N \times D}; \mathcal{I}_{2D-1})$ , where  $q = 1, 2, \dots, \mathcal{Q}$  and  $\mathcal{Q} \in \left\{(\frac{M}{2} - 1)^0, (\frac{M}{2} - 1)^1, \dots, (\frac{M}{2} - 1)^{D-1}\right\}$ .
  - Each cell is associated with a distinct  $M$ -phase candidate vector  $\mathbf{s}_q(\mathbf{V}_{N \times D}; \mathcal{I}_{2D-1})$  in the sense that:
    - (i)  $\mathbf{s}_q(\mathbf{V}_{N \times D}; \phi_{1:2D-1}) = \mathbf{s}_q(\mathbf{V}_{N \times D}; \mathcal{I}_{2D-1})$  for all  $\phi_{1:2D-1} \in C_q(\mathbf{V}_{N \times D}; \mathcal{I}_{2D-1})$ .
    - (ii)  $\phi(\mathbf{V}_{N \times D}; \mathcal{I}_{2D-1})$  is a single point of  $C_q(\mathbf{V}_{N \times D}; \mathcal{I}_{2D-1})$  where  $\phi_{2D-1}$  is minimized.

# Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

- Illustrative example:  $D = 2$ ,  $N = 4$  and  $M = 8$ .

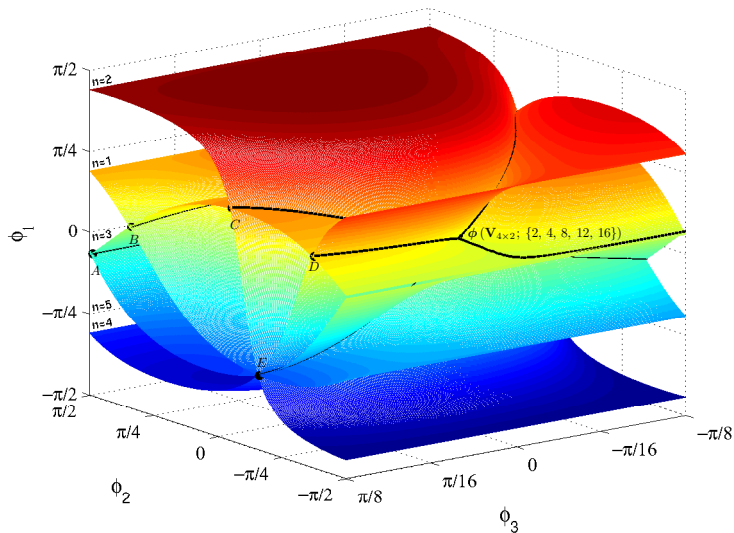


## Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

- What's the value of  $\mathcal{Q}$ ? Depends on the number of participating pairs of hypersurfaces that originate from the same row of matrix  $\mathbf{V}$ .
- E.g.
  - No pairs (i.e. all hypersurfaces originate from different rows of  $\mathbf{V}$ )  $\rightarrow \phi(\mathbf{V}_{N \times D}; \mathcal{I}_{2D-1})$  "leads" only one cell,
  - One pair  $\rightarrow \phi(\mathbf{V}_{N \times D}; \mathcal{I}_{2D-1})$  "leads"  $(\frac{M}{2} - 1)$  cells,
  - $D - 1$  pairs  $\rightarrow \phi(\mathbf{V}_{N \times D}; \mathcal{I}_{2D-1})$  "leads"  $(\frac{M}{2} - 1)^{D-1}$  cells,

# Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

- Illustrative example:  $D = 2, N = 4$  and  $M = 8$ .



## Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

- Each cell is associated with a distinct  $M$ -PSK candidate vector.
- We can collect all these vectors into:

$$\mathcal{J}(\mathbf{V}_{N \times D}) \triangleq \bigcup_{\substack{\mathcal{I}_{2D-1} \subset \{1, 2, \dots, \frac{MN}{2}\}, \\ \phi(\mathbf{V}_{N \times D}; \mathcal{I}_{2D-1}) \in \Phi^{2D-2} \times (-\frac{\pi}{M}, \frac{\pi}{M}]}} \left\{ \mathbf{s}(\mathbf{V}_{N \times D}; \mathcal{I}_{2D-1}) \right\} \subseteq \mathcal{A}_M^N$$

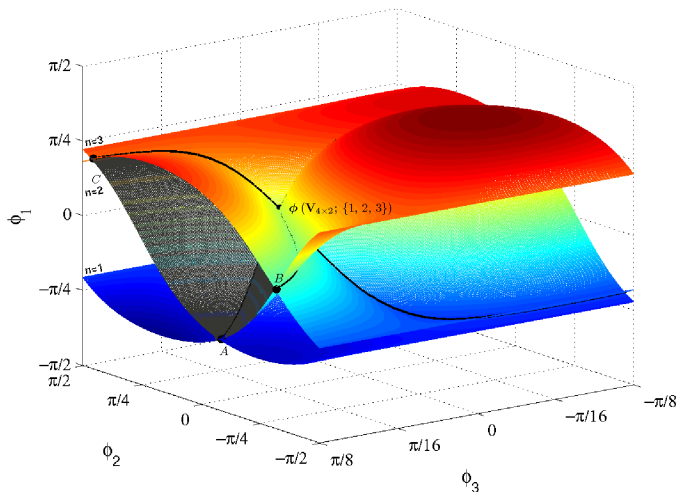
with cardinality

$$|\mathcal{J}(\mathbf{V}_{N \times D})| = \sum_{i=0}^{D-1} \binom{N}{i} \binom{N-i}{2(D-i)-1} \frac{M^{2(D-i)-2}}{2} \left( \frac{M}{2} - 1 \right)^i.$$

- Are there any regions missing?

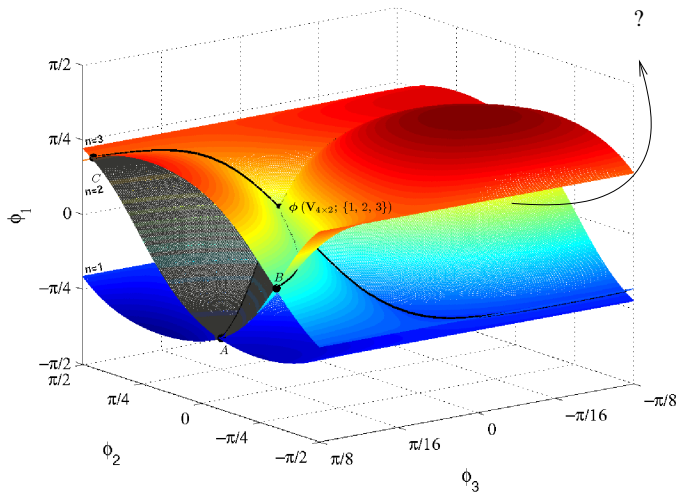
# Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

- Illustrative example:  $D = 2, N = 4$  and  $M = 8$ .



# Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

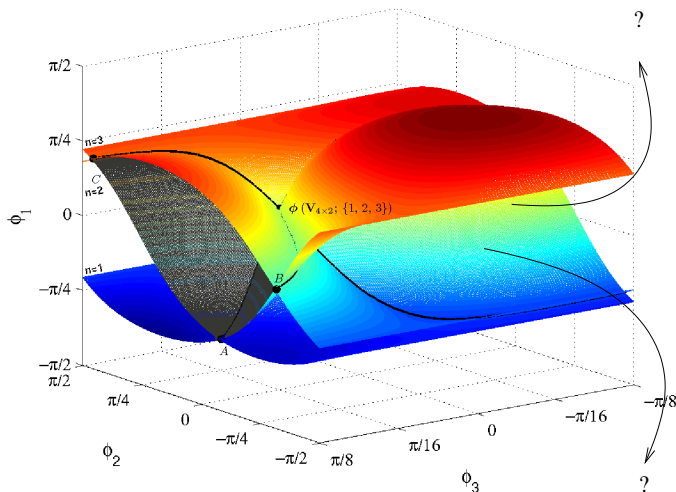
- Illustrative example:  $D = 2, N = 4$  and  $M = 8$ .





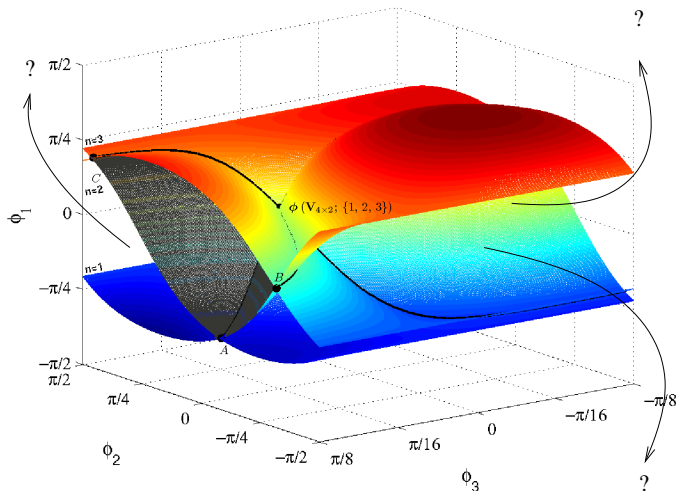
# Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

- Illustrative example:  $D = 2$ ,  $N = 4$  and  $M = 8$ .



# Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

- Illustrative example:  $D = 2, N = 4$  and  $M = 8$ .



## Proposition

For any  $\phi_{1:2D-1} \in \Phi^{2D-2} \times (-\frac{\pi}{M}, \frac{\pi}{M}]$  the following hold true:

- (i)  $\mathbf{s}(\mathbf{V}_{N \times D}; \phi_{1:2D-2}, -\frac{\pi}{M}) = e^{j\frac{2\pi}{M}} \mathbf{s}(\mathbf{V}_{N \times D}; \phi'_{1:2D-2}, \frac{\pi}{M})$  for some  $\phi'_{1:2D-2} \in \Phi^{2D-2}$ ,
- (ii)  $\mathbf{s}(\mathbf{V}_{N \times D}; \phi_{1:2D-3}, \frac{\pi}{2}, \phi_{2D-1}) = \mathbf{s}(\mathbf{V}_{N \times (D-1)}; \phi_{1:2D-3})$ ,
- (iii)  $\mathbf{s}(\mathbf{V}_{N \times D}; \phi_{1:2D-3}, -\frac{\pi}{2}, \phi_{2D-1}) = -\mathbf{s}(\mathbf{V}_{N \times D}; -\phi_{1:2D-3}, \frac{\pi}{2}, \phi'_{2D-1})$ ,  $\forall \phi'_{2D-1} \in (-\frac{\pi}{M}, \frac{\pi}{M}]$ ,
- (iv)  $\mathbf{s}(\mathbf{V}_{N \times D}; \phi_{1:2D-3}, \pm\frac{\pi}{2}, \phi_{2D-1}) = \mathbf{s}(\mathbf{V}_{N \times D}; \phi_{1:2D-3}, \pm\frac{\pi}{2}, \phi'_{2D-1})$ ,  $\forall \phi'_{2D-1} \in (-\frac{\pi}{M}, \frac{\pi}{M}]$ .

- Hence,  $\mathcal{S}(\mathbf{V}_{N \times D}) = \mathcal{J}(\mathbf{V}_{N \times D}) \cup \mathcal{S}(\mathbf{V}_{N \times (D-1)})$  and thus:

$$\begin{aligned} \mathcal{S}(\mathbf{V}_{N \times D}) &= \mathcal{J}(\mathbf{V}_{N \times D}) \cup \mathcal{J}(\mathbf{V}_{N \times (D-1)}) \cup \dots \cup \mathcal{J}(\mathbf{V}_{N \times 1}) \\ &= \bigcup_{d=0}^{D-1} \mathcal{J}(\mathbf{V}_{N \times (D-d)}). \end{aligned}$$

# Hypersurfaces and Cardinality of $\mathcal{S}(\mathbf{V}_{N \times D})$

- Cardinality of  $\mathcal{S}(\mathbf{V}_{N \times D})$ :

$$\begin{aligned} |\mathcal{S}(\mathbf{V}_{N \times D})| &= |\mathcal{J}(\mathbf{V}_{N \times D})| + |\mathcal{J}(\mathbf{V}_{N \times (D-1)})| + \cdots + |\mathcal{J}(\mathbf{V}_{N \times 1})| \\ &= \sum_{d=1}^D \sum_{i=0}^{d-1} \binom{N}{i} \binom{N-i}{2(d-i)-1} \left(\frac{M}{2}\right)^{2(d-i)-2} \left(\frac{M}{2} - 1\right)^i \\ &= \mathcal{O}\left(\left(\frac{MN}{2}\right)^{2D-1}\right). \end{aligned}$$

- If  $\mathbf{Q}$  is full rank,

## Remark

If  $D = N$ , then the computation of  $\mathbf{s}_{\text{opt}}$  is  $\mathcal{NP}$ -hard and can be implemented by applying exhaustive search or the proposed algorithm among all elements of  $\mathcal{A}_M^{N-1}$  since  $|\mathcal{A}_M^{N-1}| = |\mathcal{S}(\mathbf{V}_{N \times D})| = M^{N-1}$ .

- We have utilized  $2D - 1$  auxiliary hyperspherical coordinates.
- Partitioned the hypercube  $\Phi^{2D-2} \times (-\frac{\pi}{M}, \frac{\pi}{M}]$  into a finite number of cells that are associated with distinct  $M$ -phase vectors.
- $|\mathcal{S}(\mathbf{V}_{N \times D})| = \mathcal{O}\left(\left(\frac{MN}{2}\right)^{2D-1}\right) \subseteq \mathcal{A}_M^N$  and proved that  $\mathbf{s}_{\text{opt}} \in \mathcal{S}(\mathbf{V}_{N \times D})$ .
- Complexity to build  $\mathcal{S}(\mathbf{V}_{N \times D})$ :  $\mathcal{O}\left(\left(\frac{MN}{2}\right)^{2D}\right)$ .
- But how?

## Algorithm

- For each one of  $|\mathcal{J}(\mathbf{V}_{N \times D})|$  intersections, find the coordinates  $\phi_1, \dots, \phi_{2D-1}$  and the corresponding complex hyperspherical vector  $\mathbf{c}(\phi_{1:2D-1})$  solving the system of hypersurfaces:

$$\tilde{\mathbf{V}}_{\mathcal{I}_{2d-1}, 1:2d} \tilde{\mathbf{c}}(\phi_{1:2d-1}) = \mathbf{0}_{(2d-1) \times 1}.$$

- Build candidate vectors using:

$$\mathbf{s}(\mathbf{V}_{N \times D}; \phi_{1:2D-1}) \triangleq \begin{bmatrix} s(\mathbf{V}_{1,1:D}; \phi_{1:2D-1}) \\ s(\mathbf{V}_{2,1:D}; \phi_{1:2D-1}) \\ \vdots \\ s(\mathbf{V}_{N,1:D}; \phi_{1:2D-1}) \end{bmatrix}$$

**Note:** vector elements associated with  $2D - 1$  intersecting hypersurfaces lead to ambiguous decisions.

- Use disambiguation rules for the ambiguous decisions.

## Disambiguation rules

- Let  $\mathcal{N}_{\mathcal{I}_{2d-1}}$  the set of indices of rows from  $\mathbf{V}$  participating in the intersection  $\mathcal{I}_{2d-1}$ .

- (i) For any  $n \in \{1, 2, \dots, N\}$  and  $n \notin \mathcal{N}_{\mathcal{I}_{2d-1}}$ :

$$s_{q,n}(\mathbf{V}_{N \times d}; \mathcal{I}_{2d-1}) = s(\mathbf{V}_{n,1:d}; \phi(\mathbf{V}_{N \times d}; \mathcal{I}_{2d-1})), \forall q \in \{1, 2, \dots, Q\}$$

- (ii) For any  $n \in \mathcal{N}_{\mathcal{I}_{2d-1}}$  such that there is only one hypersurface related with the  $n$ -th row of  $\mathbf{V}$ :

$$s_{q,n}(\mathbf{V}_{N \times d}; \mathcal{I}_{2d-1}) = s(\mathbf{V}_{n,1:d}; \phi(\mathbf{V}_{N \times d}; \mathcal{I}_{2d-1} - \{i_k\})), \forall q \in \{1, 2, \dots, Q\}$$

where  $\Im\{\mathbf{V}_{n,d}\} = 0$ .

- (iii) For any  $n \in \mathcal{N}_{\mathcal{I}_{2d-1}}$  such that there is a pair of hypersurfaces say  $\mathcal{H}(\tilde{\mathbf{V}}_{i_k,1:2d})$ ,  $\mathcal{H}(\tilde{\mathbf{V}}_{i_m,1:2d})$  related with the  $n$ -th row of  $\mathbf{V}$ :

$$s_{q,n}(\mathbf{V}_{N \times d}; \mathcal{I}_{2d-1}) = s\left(\mathbf{V}_{n,1:d}; \phi(\mathbf{V}_{N \times d}; \mathcal{I}_{2d-1} - \{i_k, i_m\}) + \{i'_k\}\right) \cap s\left(\mathbf{V}_{n,1:d}; \phi(\mathbf{V}_{N \times d}; \mathcal{I}_{2d-1} - \{i_k, i_m\}) + \{i'_m\}\right),$$

where  $\Im\{\mathbf{V}_{n,d}\} = 0$  for  $\forall q \in \{1, 2, \dots, Q\}$ .

## Algorithm

- For each one of  $|\mathcal{J}(\mathbf{V}_{N \times D})|$  intersections, find the coordinates  $\phi_1, \dots, \phi_{2D-1}$  and the corresponding complex hyperspherical vector  $\mathbf{c}(\phi_{1:2D-1})$  solving the system of hypersurfaces:

$$\tilde{\mathbf{V}}_{\mathcal{I}_{2d-1,1:2d}} \tilde{\mathbf{c}}(\phi_{1:2d-1}) = \mathbf{0}_{(2d-1) \times 1}.$$

- Build candidate vectors using:

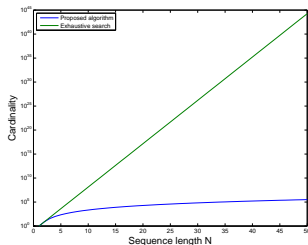
$$\mathbf{s}(\mathbf{V}_{N \times D}; \phi_{1:2D-1}) \triangleq \begin{bmatrix} s(\mathbf{V}_{1,1:D}; \phi_{1:2D-1}) \\ s(\mathbf{V}_{2,1:D}; \phi_{1:2D-1}) \\ \vdots \\ s(\mathbf{V}_{N,1:D}; \phi_{1:2D-1}) \end{bmatrix}$$

**Note:** vector elements associated with  $2D - 1$  intersecting hypersurfaces lead to ambiguous decisions.

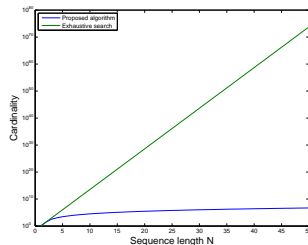
- Use disambiguation rules for the ambiguous symbols.
- Set  $D \leftarrow D - 1$  (drop the rightmost column of complex matrix  $\mathbf{V}$ ) and repeat steps.



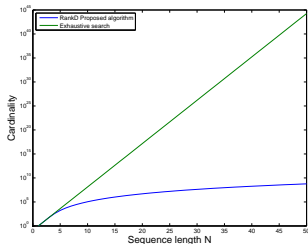
# Complexity Study



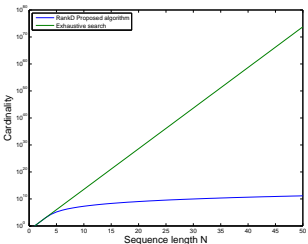
(a)  $D = 2, M = 8$



(b)  $D = 2, M = 32$



(c)  $D = 3, M = 8$



(d)  $D = 3, M = 32$

Figure: Proposed algorithm cardinality (blue line) Vs. exhaustive search cardinality (green line).

## Conclusions:

- Maximization of a rank-deficient form with  $M$ -phase argument proved to be polynomially solvable.
- Implementable algorithm even for large sequence lengths.
- Time and memory efficient, fully parallelizable and rank-scalable algorithm.