

# Optimizing Minimum Redundancy Arrays for Robustness

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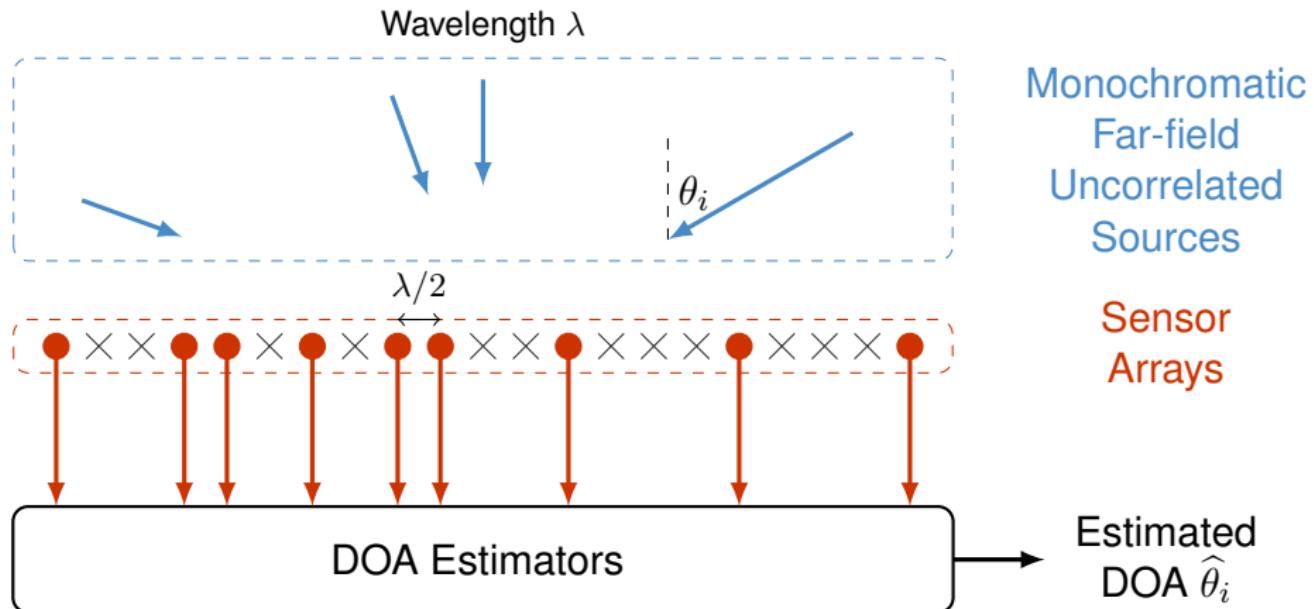
# Outline

- 1 Introduction
- 2 Review of Sparse Arrays and Robustness
- 3 Robust Minimum Redundancy Arrays
- 4 Concluding Remarks

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# Direction-Of-Arrival (DOA) Estimation



<sup>1</sup>Van Trees, *Optimum Array Processing: Part IV of Detection, Estimation, and Modulation Theory*, 2002.

# Physical Array and Difference Coarray

Physical array  $\mathbb{S}$



Difference coarray  $\mathbb{D} \triangleq \{n_1 - n_2 : n_1, n_2 \in \mathbb{S}\}$



<sup>1</sup>Van Trees, *Optimum Array Processing: Part IV of Detection, Estimation, and Modulation Theory*, 2002.

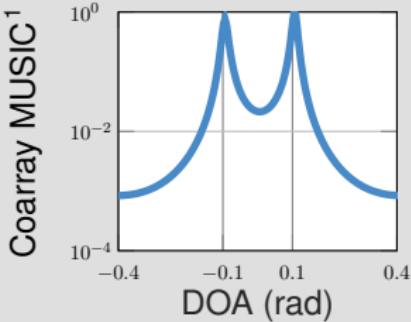
# Sensor Failures

Array #1



5 elements

$$\text{RMSE} = 0.00617$$

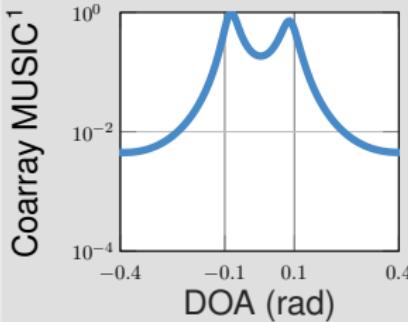


Array #2 (2 fails)



4 elements

$$\text{RMSE} = 0.014367$$



Array #3 (1 fails)



4 elements

Coarray MUSIC is not applicable here!

<sup>1</sup>Liu and Vaidyanathan, *IEEE Signal Process. Letters*, 2015.

<sup>2</sup>100 snapshots, 0dB SNR,  $D = 2$  sources,  $\theta_1 = -0.1$ ,  $\theta_2 = 0.1$ , equal-power, uncorrelated sources.

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# ULA and Sparse Arrays

## ULA (not sparse)



- Identify at most  $N - 1$  uncorrelated sources.<sup>1</sup>  
( $N$  is the number of sensors)
- Can only find fewer sources than sensors.

## Linear sparse arrays

- 1 Minimum redundancy arrays<sup>2</sup>
  - 2 Nested arrays<sup>3</sup>
  - 3 Coprime arrays<sup>4</sup>
  - 4 Super nested arrays<sup>5</sup>
- Identify  $\mathcal{O}(N^2)$  uncorrelated sources.
  - More sources than sensors!

<sup>1</sup> Van Trees, *Optimum Array Processing: Part IV of Detection, Estimation, and Modulation Theory*, 2002.

<sup>2</sup> Moffet, *IEEE Trans. Antennas Propag.*, 1968.

<sup>3</sup> Pal and Vaidyanathan, *IEEE Trans. Signal Process.*, 2010.

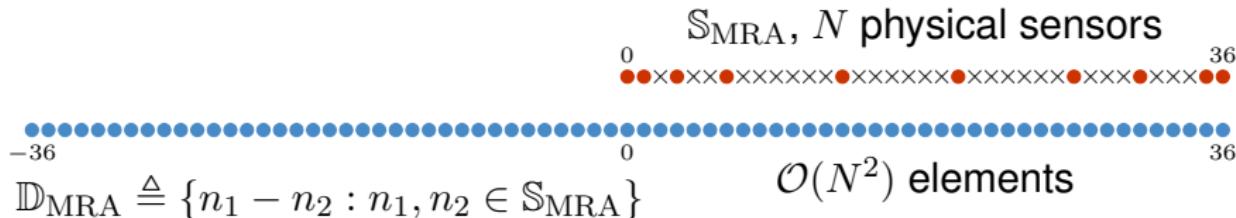
<sup>4</sup> Vaidyanathan and Pal, *IEEE Trans. Signal Process.*, 2011.

<sup>5</sup> Liu and Vaidyanathan, *IEEE Trans. Signal Process.*, 2016.

# Minimum Redundancy Arrays

$$\mathbb{S}_{\text{MRA}} \triangleq \arg \max_{\mathbb{S}} |\mathbb{D}| \quad \text{subject to} \quad |\mathbb{S}| = N, \quad \mathbb{D} = \mathbb{U}.$$

- $N$  physical sensors
- Hole-free  $\mathbb{D}$
- No closed-form expressions for  $\mathbb{S}_{\text{MRA}}$



<sup>1</sup> Moffet, IEEE Trans. Antennas Propag., 1968.

# The Essentialness Property

The sensor  $n \in \mathbb{S}$  is **essential** with respect to  $\mathbb{S}$  if  $\overline{\mathbb{D}} \neq \mathbb{D}$ .

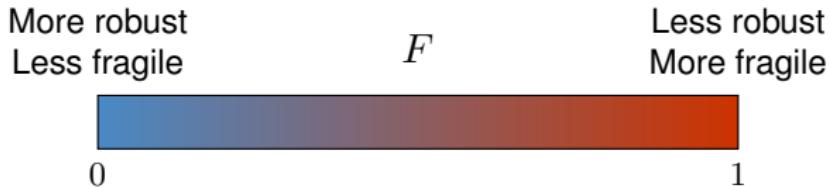
	Physical array	Difference coarray	
(a)	● ● ● ● ● ● ● ● 0 1 2 3 4 5 6 7 8	● ● ● ● ● ● ● ● -8 0 8	
(b)	× ● ● ● ● ● ● ● 1 2 3 4 5 6 7 8	× ● ● ● ● ● ● ● × -7 0 7	0 is essential
(c)	● × ● ● ● ● ● ● 0 2 3 4 5 6 7 8	● ● ● ● ● ● ● ● -8 0 8	1 is inessential
(d)	● ● × ● ● ● ● ● 0 1 3 4 5 6 7 8	● ● ● ● ● ● ● ● -8 0 8	2 is inessential
(e)	● ● ● ● ● ● × ● 0 1 2 3 4 5 6 8	● ● ● ● ● ● ● ● -8 0 8	7 is inessential

<sup>1</sup>Liu and Vaidyanathan, IEEE ICASSP, 2018;  $\mathbb{D}$  is the difference coarray of  $\mathbb{S}$  and  $\overline{\mathbb{D}}$  is the difference coarray of  $\mathbb{S} \setminus \{n\}$ .

# The Fragility $F$ : Definition

The fragility  $F \triangleq \frac{\# \text{ of essential sensors}}{\# \text{ of sensors}}$ .

$$\frac{2}{N} \leq F \leq 1, \quad \text{for all } N \geq 4$$

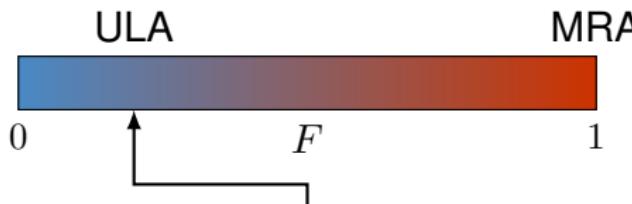


<sup>1</sup>Liu and Vaidyanathan, IEEE ICASSP, 2018.

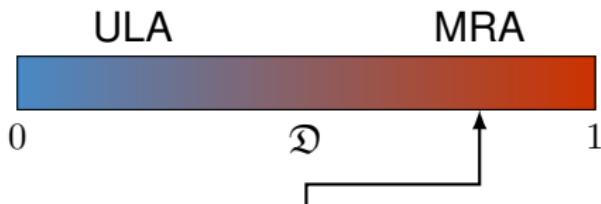
# The Fragility $F$ and the Normalized Size of $\mathbb{D}$

$$F \triangleq \frac{\text{\# of essential sensors}}{\text{\# of all sensors } (N)}$$

$$\mathfrak{D} \triangleq \frac{|\mathbb{D}|}{N^2 - N + 1}$$



- More robust
- Less fragile



- More sources than sensors
- Higher resolution



<sup>1</sup>Liu and Vaidyanathan, IEEE SAM, 2018.

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# RMRA: Definition

$$(P1) : \quad S_{\text{RMRA}} \triangleq \arg \max_S |\mathbb{D}| \quad \text{subject to} \quad (1)$$

$$|S| = N, \quad (2)$$

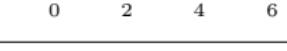
$$\mathbb{D} = \mathbb{U}, \quad (3)$$

$$F = \frac{2}{N}, \quad N \geq 4. \quad (4)$$

- (2):  $N$  physical sensors
- (3): Hole-free  $\mathbb{D}$
- (4): The minimum fragility (the most robust)

<sup>1</sup>Liu and Vaidyanathan, ACSSC, 2018.

# RMRA: Array Geometries Found by Direct Searching

$N$	Array Configuration	Essential	Inessential
4	#1: 	◆ Essential	■ Inessential
5	#1: 		
6	#1:  #2: 		
7	#1:  #2: 		
8	#1:  #2: 		
9	#1: 		
10	#1:  #2: 		

<sup>1</sup>Liu and Vaidyanathan, ACSSC, 2018.

# RMRA: Properties

$$(P1) : \quad \mathbb{S}_{\text{RMRA}} \triangleq \arg \max_{\mathbb{S}} |\mathbb{D}| \quad \text{subject to}$$

$$|\mathbb{S}| = N,$$

$$\mathbb{D} = \mathbb{U},$$

$$F = \frac{2}{N}, \quad N \geq 4.$$


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- The solution to (P1) exists.
- The solution to (P1) is not unique.
- To the best of our knowledge, closed-form expressions for  $\mathbb{S}_{\text{RMRA}}$  is not known.
- The difference coarray of  $\mathbb{S}_{\text{RMRA}}$  has size  $\mathcal{O}(N^2)$ .

<sup>1</sup>Liu and Vaidyanathan, ACSSC, 2018.

# Theorem 1: The Size of the Difference Coarray

## Theorem

Let  $\mathbb{S}_{\text{RMRA}}$  be a solution to (P1) with  $N \geq 4$  physical sensors. The aperture of  $\mathbb{S}_{\text{RMRA}}$  is denoted by  $A_{\text{RMRA}}$ . Then

$$L_{\text{RMRA}} \leq \frac{N^2}{A_{\text{RMRA}}} < U_{\text{RMRA}}, \quad (5)$$

where

$$L_{\text{RMRA}} \triangleq 4 + \frac{4\sqrt{2}}{3\pi}, \quad U_{\text{RMRA}} \triangleq 16. \quad (6)$$

The smaller  $N^2/A$  is,  
the better the array is (in the sense of large hole-free  $\mathbb{D}$ ).

<sup>1</sup>Liu and Vaidyanathan, ACSSC, 2018.

# Comparison between MRA and RMRA

**MRA** [ErdősGál1948, RédeiRényi1948, Leech1956]

For MRA with  $N$  sensors,

$$L_{\text{MRA}} \leq \lim_{N \rightarrow \infty} \frac{N^2}{A_{\text{MRA}}} \leq U_{\text{MRA}}, \quad (7)$$

where

$$L_{\text{MRA}} \triangleq 2.434\dots, \quad U_{\text{MRA}} \triangleq 3.348\dots \quad (8)$$

For sufficiently large  $N$ , we have

$$0.2987N^2 \leq A_{\text{MRA}} \leq 0.4108N^2,$$

Due to the above

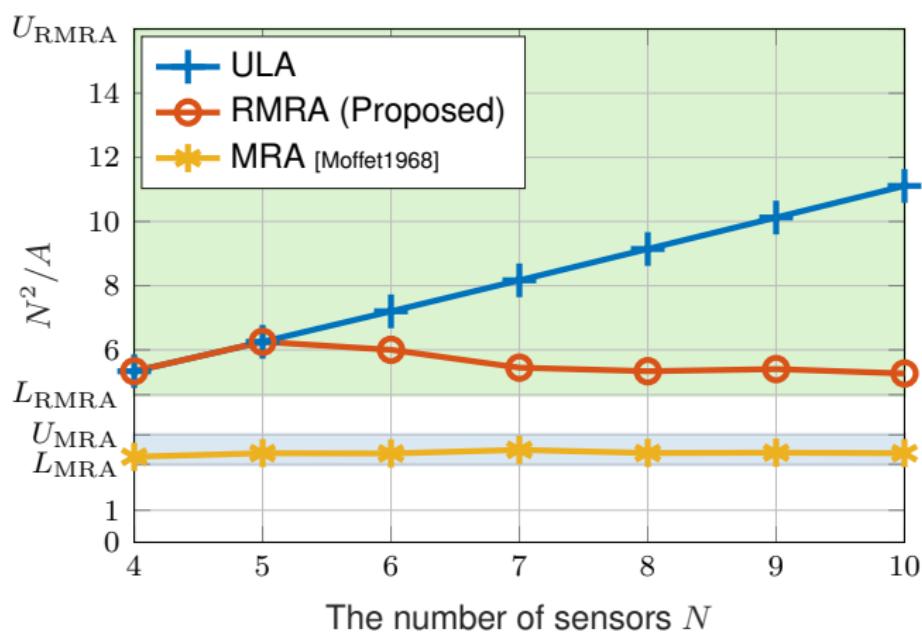
$$0.0625N^2 < A_{\text{RMRA}} \leq 0.2174N^2.$$

This talk

<sup>1</sup>Liu and Vaidyanathan, ACSSC, 2018.

<sup>2</sup>Erdős and Gál, Indagationes Mathematicae, 1948; Rédei and Rényi, Recueil Mathématique, 1948; Leech, J. London Math. Soc., 1956, Moffet, IEEE Trans. Antennas Propag., 1968.

# A Numerical Example of Theorem 1



ULA

$$F = \frac{2}{N}$$

$$|\mathbb{D}| = \mathcal{O}(N)$$

RMRA  
(Proposed)

$$F = \frac{2}{N}$$

$$|\mathbb{D}| = \mathcal{O}(N^2)$$

MRA [Moffet1968]

$$F = 1$$

$$|\mathbb{D}| = \mathcal{O}(N^2)$$

$$L_{\text{RMRA}} = 4 + \frac{4\sqrt{2}}{3\pi} \approx 4.6002, \quad U_{\text{RMRA}} = 16, \quad L_{\text{MRA}} = 2.434\dots, \quad U_{\text{MRA}} = 3.348\dots$$

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# Concluding Remarks

- Robust array geometries with large difference coarray
- Robust minimum redundancy array (RMRA) with  $N$  sensors
  - Minimum fragility ( $F = 2/N$ )
  - Hole-free difference coarray
  - $|\mathbb{D}_{\text{RMRA}}| = \mathcal{O}(N^2)$
- Future work
  - Suboptimal solutions to RMRA with large  $N$

Thank you!