

# OPTIMIZED SPARSE ARRAY DESIGN BASED ON THE SUM CO-ARRAY

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## Introduction

### Motivation

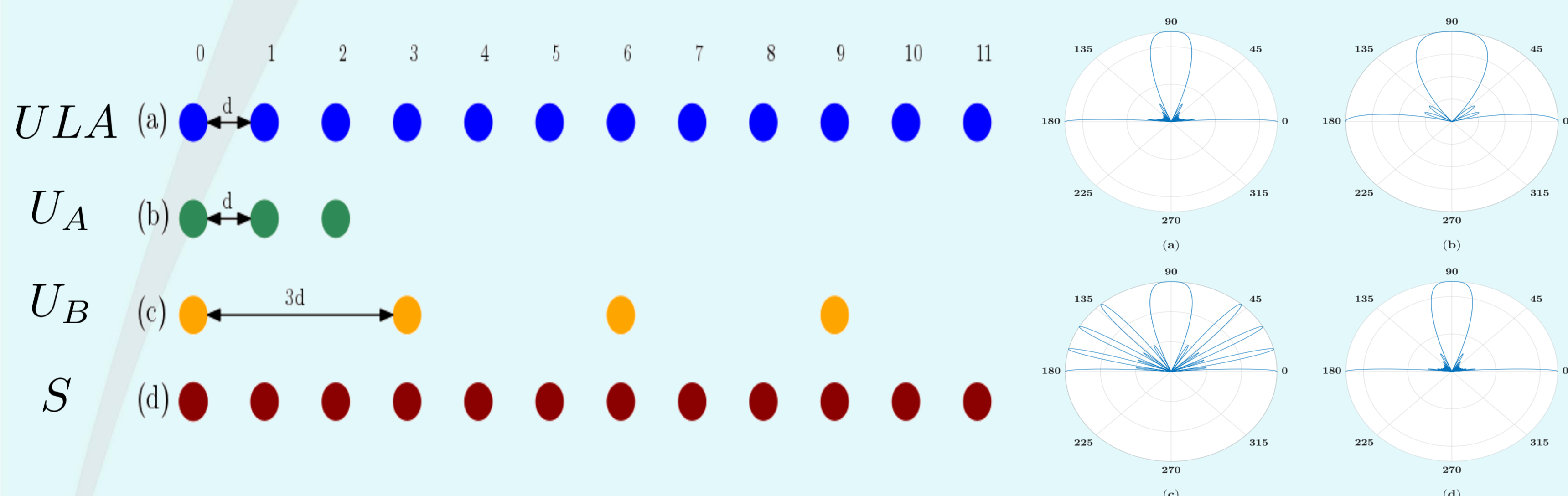
The number of elements of a uniform linear array (ULA) is the main bottleneck in many applications such as radar, communication and ultrasound imaging.

### Main Goal

Reducing the number of elements while preserving the beam pattern, which determines the image quality or detection performance.

### Solution

We propose the use of multiplicative beamforming along with sparse array design based on the sum co-array. We employ multiple sub-arrays, which the product of their beam patterns yields an effective beam pattern, similar to that of the full array. This allows for significant element reduction without compromising the performance.



## Sparse Multiplicative Beamforming

Beam pattern of standard delay and sum (DAS) beamforming:

$$H_{\text{DAS}}(\theta) = \sum_{n=0}^{N-1} \exp(2\pi j \frac{d \sin \theta}{\lambda} n).$$

### Proposed Sparse Arrays

$$U_A = \{0, \dots, A-1\}, \quad U_B = \{nA : n = 0, \dots, B-1\}$$

where  $N = AB$ . **Sum co-array property:**  $S = U_A + U_B = \{0, \dots, N-1\}$ .

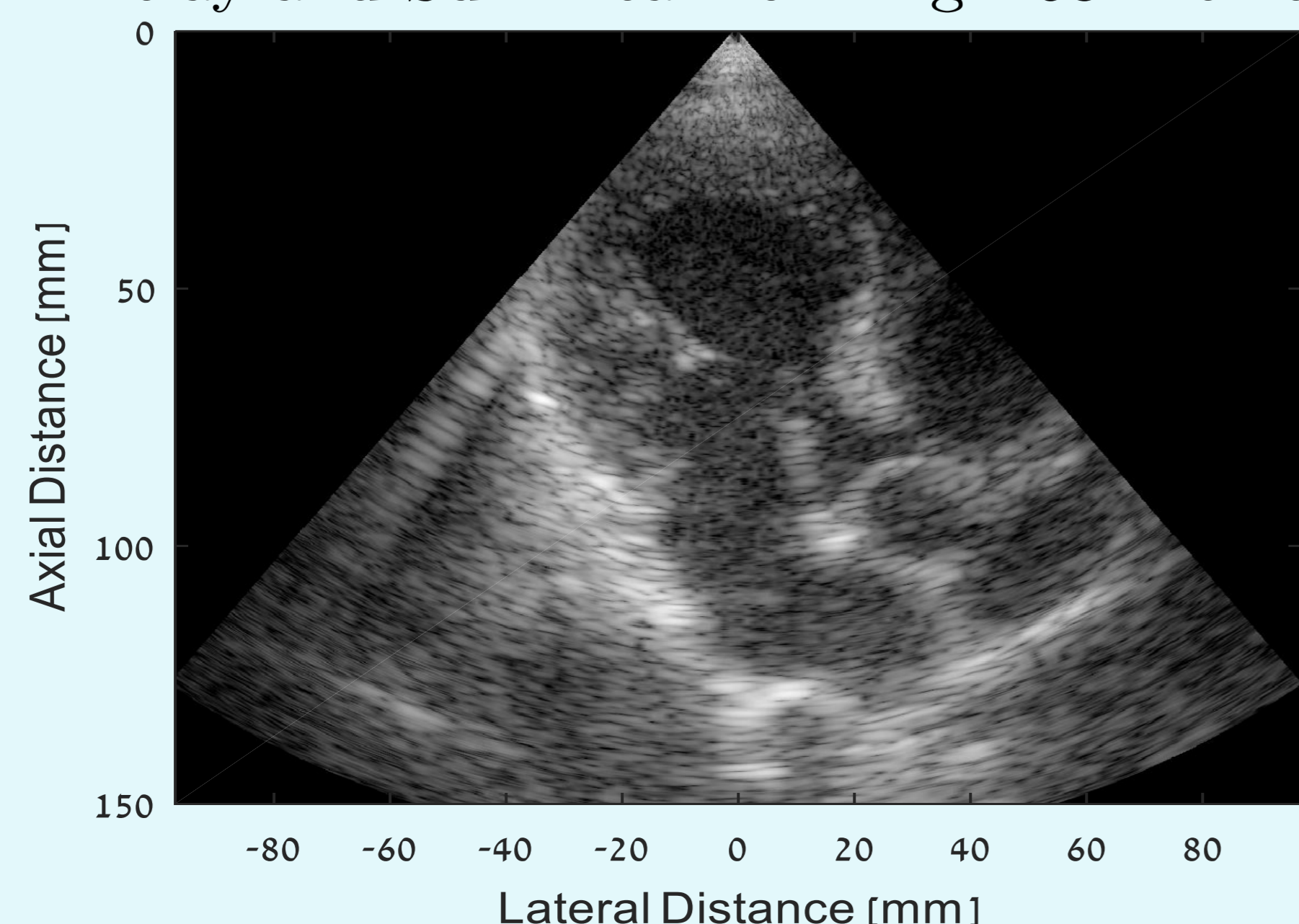
### Multiplicative Beamforming

1. Apply DAS on each sub-array  $U_A$  and  $U_B$  separately.
2. Multiply the two results.

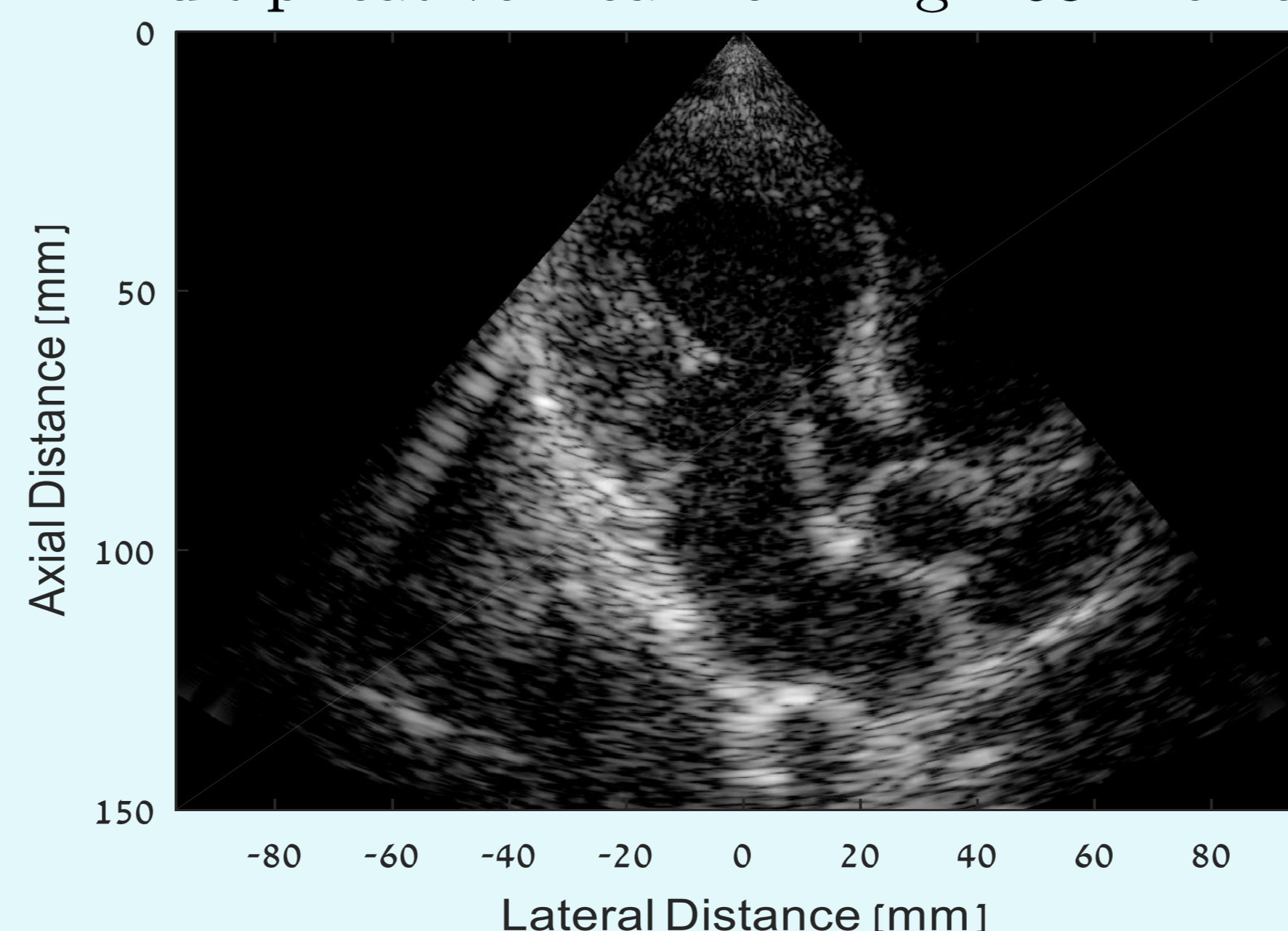
### Effective Beam Pattern

$$H(\theta) = \left( \sum_{n=0}^{A-1} \exp(2\pi j \frac{d \sin \theta}{\lambda} n) \right) \left( \sum_{m=0}^{B-1} \exp(2\pi j \frac{d \sin \theta}{\lambda} mA) \right) = H_{\text{DAS}}(\theta).$$

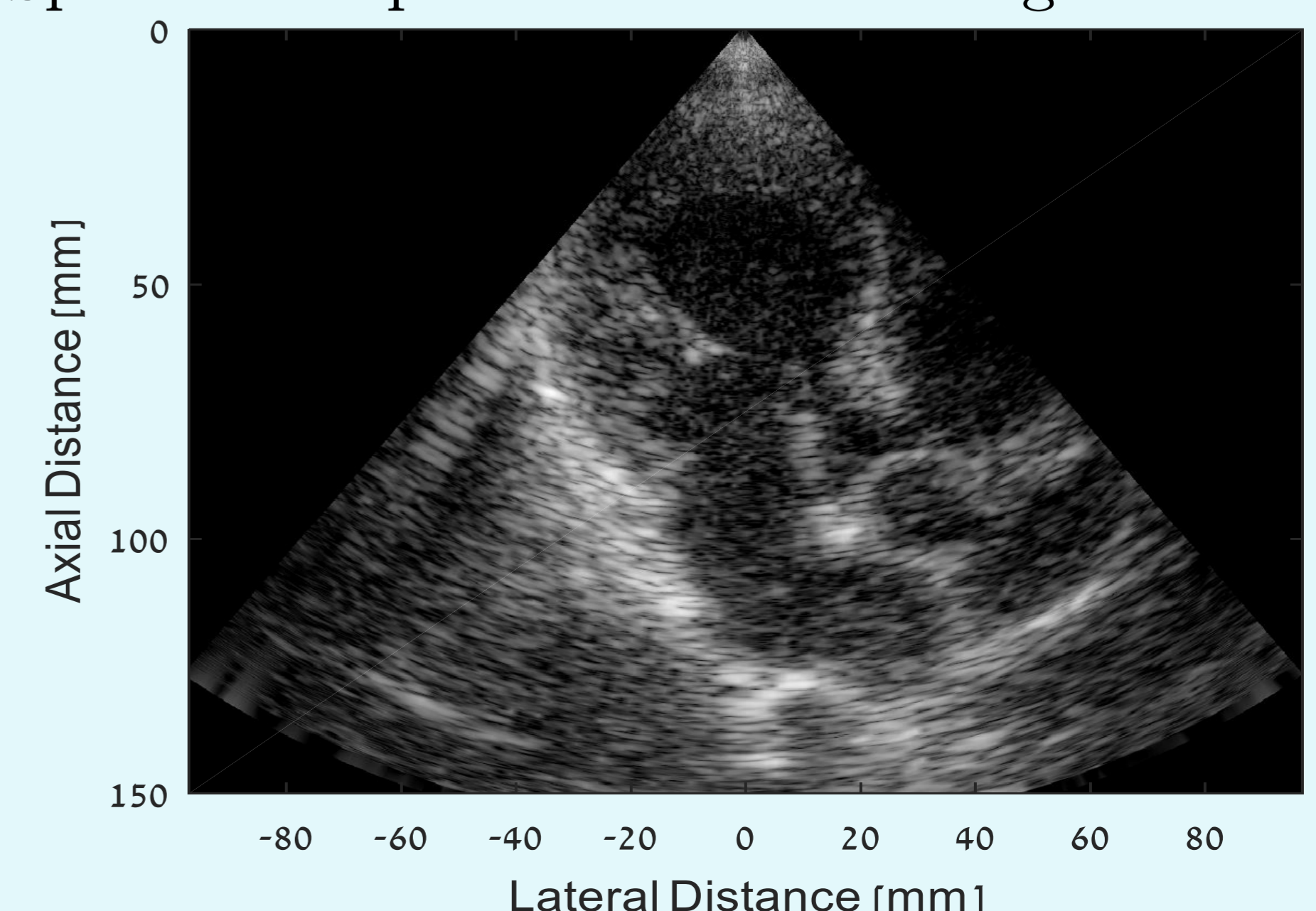
Delay and Sum Beamforming - 63 Elements



Multiplicative Beamforming - 63 Elements



Sparse Multiplicative Beamforming - 21 Elements



### Minimal Number of Elements

$$\min_{A,B} A + B \quad s.t. \quad N = AB \rightarrow A = \sqrt{N}, B = \sqrt{N}$$

### Extension Beyond Two Sub-Arrays

$$U_1 = \{0, \dots, A_1 - 1\} \quad U_k = \{n \prod_{i=1}^{k-1} A_i : n = 0, \dots, A_k - 1, k = 2, \dots, K\}.$$

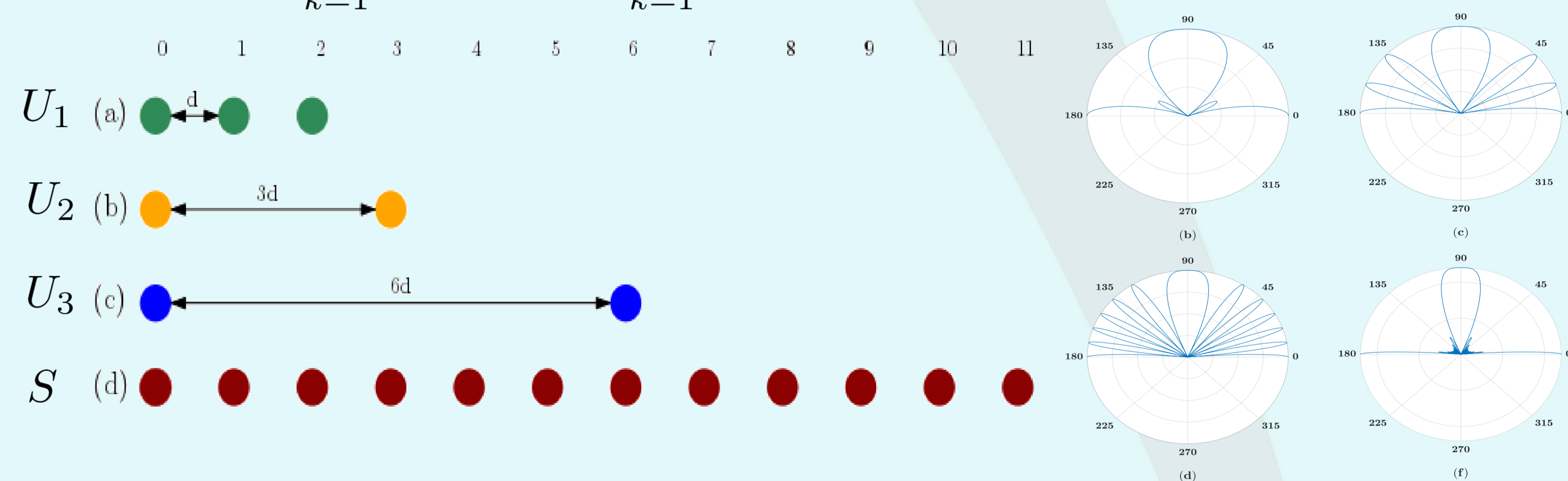
$$\rightarrow S = \sum_{k=1}^K U_k = \{0, \dots, N-1\}$$

### Effective Beam Pattern

$$H(\theta) = \prod_{k=1}^K \left( \sum_{n=0}^{A_k-1} \exp(2\pi j \frac{d \sin \theta}{\lambda} n \prod_{i=1}^{k-1} A_i) \right) = H_{\text{DAS}}(\theta)$$

### Minimal Number of Elements

$$\min_{K, A_1, \dots, A_K} \sum_{k=1}^K A_k \quad s.t. \quad N = \prod_{k=1}^K A_k \rightarrow K = \log_2 N, A_k = 2 \quad k = 1, \dots, K.$$



## Summary

### Sparse Multiplicative Beamforming

- Performing multiplicative beamforming is equivalent to apply standard DAS on the sum co-array.
- We outline a sparse array design composed of two sub-arrays which their sum co-array is full ULA. This allows to obtain the beam pattern of the full array, while using fewer elements on the order of  $2\sqrt{N}$ .
- Extending this approach to numerous sub-arrays enables element reduction where the number of sensors is as low as  $\log_2 N$ .

### Results

- We show in the simulations above that the beam pattern of the full array can be realized using several sub-arrays with fewer elements.
- The proposed method was applied for ultrasound imaging using *in-vivo* cardiac data. The results shown below, prove that the proposed approach is valid and suitable for clinical use.