## A Gridless DOA Estimation Method Based on Convolutional Neural Network With Toeplitz Prior

M: # of sensors, L: # of snapshots, K: # of sources

N: position of the last sensor in our sparse array

Aa: MxK manifold matrix of our SLA

A: NxK manifold matrix of our virtual array (ULA)

## Example:

→ The sensor errory

-> The virtual array

$$\mathcal{G}_{\mathcal{R}} = \mathcal{A}_{\mathcal{R}} S + \Lambda$$

$$\mathcal{A}_{\mathcal{R}} = \frac{1}{L} \mathcal{A}_{\mathcal{R}} S S^{H} \mathcal{A}_{\mathcal{R}}^{H} + \mathcal{T}_{\Lambda^{2}}^{2} \cdot \mathcal{I}_{M} = \mathcal{A}_{\mathcal{R}} \mathcal{A}_{\mathcal{R}} \mathcal{A}_{\mathcal{R}}^{H} + \mathcal{T}_{\Lambda^{2}}^{2} \cdot \mathcal{I}_{M}$$

$$= \mathcal{I}_{\mathcal{R}} \mathcal{A}_{\mathcal{R}} \mathcal{A}^{H} \mathcal{I}_{\mathcal{R}}^{H} + \mathcal{T}_{\Lambda^{2}}^{2} \cdot \mathcal{I}_{M}$$

$$= \mathcal{I}_{\mathcal{R}} \mathcal{A}_{\mathcal{R}} \mathcal{A}^{H} \mathcal{I}_{\mathcal{R}}^{H} + \mathcal{T}_{\Lambda^{2}}^{2} \cdot \mathcal{I}_{M}$$

$$= \mathcal{I}_{\mathcal{R}} \mathcal{A}_{\mathcal{R}} \mathcal{A}^{H} \mathcal{I}_{\mathcal{R}}^{H} + \mathcal{T}_{\Lambda^{2}}^{2} \cdot \mathcal{I}_{M}$$

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$$= \mathcal{I}_{\mathcal{R}} \mathcal{A}_{\mathcal{R}} \mathcal{A}^{H} \mathcal{I}_{\mathcal{R}}^{H} + \mathcal{I}_{\Lambda^{2}}^{2} \cdot \mathcal{I}_{\mathcal{R}}$$

$$= \mathcal{I}_{\mathcal{R}} \mathcal{A}_{\mathcal{R}} \mathcal{A}^{H} \mathcal{I}_{\mathcal{R}}^{H} + \mathcal{I}_{\Lambda^{2}}^{2} \cdot \mathcal{I}_{\mathcal{R}}$$

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$$= \mathcal{I}_{\mathcal{R}} \mathcal{A}_{\mathcal{R}} \mathcal{A}^{H} \mathcal{I}_{\mathcal{R}}^{H} + \mathcal{I}_{\Lambda^{2}}^{2} \cdot \mathcal{I}_{\mathcal{R}^{H} + \mathcal{I}_{\Lambda^{2}}^{2} \cdot \mathcal{I}_{\Lambda^{2}}$$

$$= \mathcal{I}_{\mathcal{R}} \mathcal{A}_{\mathcal{R}} \mathcal{A}_{\mathcal{R}} \mathcal{A}^{H} \mathcal{A}_{\mathcal{R}}^{H} + \mathcal{I}_{\Lambda^{2}}^{2} \cdot \mathcal$$

## TDN (detects number of sources)

- -> input: R. (MxM)
- -> output: k (N x 1 vector where all entries are zero except the Kth entry)

## CRN

- -> input: R. (MxM)
- -> output: u (N x 1 vector which is the dirst column of the toeplitz matrix R)

Alter R is composed, root-MUSIC algorithm is applied.