

# Sparse Recovery of Strong Reflectors With an Application to Non-Destructive Evaluation

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# Outline

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Motivation

Measurement setup

Problem statement

Proposed approach

- TOF recovery

- TOF matching

- Localization

Results

Conclusion



How can we reduce data rate / increase frame rate?

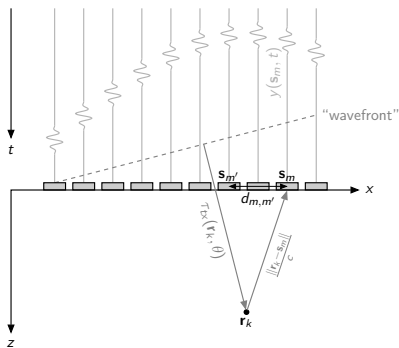
How can we reduce data rate / increase frame rate?

*To what extent can recent signal processing techniques, e.g.:*

- 1. Finite rate of innovation (FRI)*
- 2. Euclidean distance matrices (EDMs)*

*be applied to ultrasound (US) imaging / **localization**?*

# Measurement setup: plane wave insonification



Time-of-flight (TOF) for element at  $\mathbf{s}_m$  and point at  $\mathbf{r}_k = [x_k, z_k]^T$ :

$$\tau(\mathbf{r}_k, \mathbf{s}_m, \theta) = (x_k \sin \theta + z_k \cos \theta)/c + \|\mathbf{r}_k - \mathbf{s}_m\|/c.$$

## Problem statement

Given discrete measurements  $\{y_m[n]\}_{n=0}^{N-1}$  at the element positions  $\{\mathbf{s}_m\}_{m=0}^{M-1}$ , estimate the locations of the reflectors  $\{\mathbf{r}_k\}_{k=0}^{K-1}$ .

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- ▶ Exploit minimum *degrees of freedom* (DOF): least amount of elements  $M$  and samples per element  $N$ .
- ▶ Continuous recovery.

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Existing approaches discretize medium:

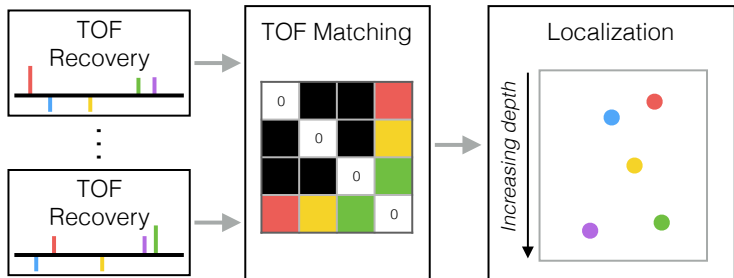
- ▶ Sparse deconvolution.
- ▶ Subspace approach  $\Rightarrow K \leq M$ .<sup>1</sup>

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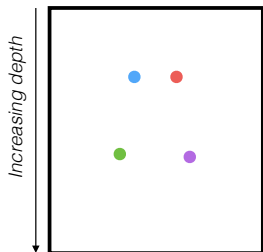
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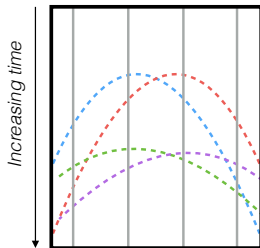
# Proposed approach



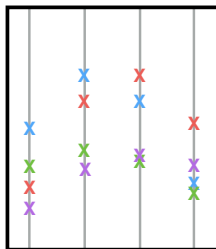
# Simple example



(a) Image domain.



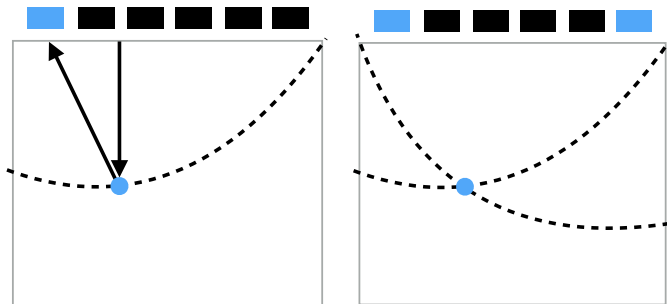
(b) Measurement domain.



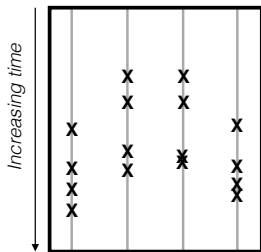
(c) Labeled TOFs

# Localization with labeled TOFs

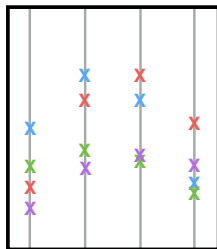
Recover reflector position from TOFs of at least two elements.



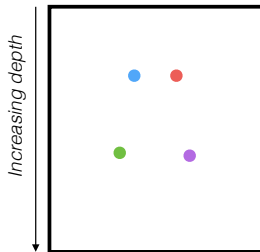
# Proposed approach applied to simple example



(a) Recovered TOFs.



(b) TOF matching.



(c) Localization.

For  $K$  reflectors, we receive the following *pulse stream* at  $\mathbf{s}_m$ :

$$y_m(t) = \sum_{k=0}^{K-1} \underbrace{\frac{a_k}{2\pi \|\mathbf{r}_k - \mathbf{s}_m\|}}_{a_{m,k}} h\left(t - \underbrace{\tau(\mathbf{r}_k, \mathbf{s}_m)}_{\tau_{m,k}}\right),$$

where  $h(t)$  is known pulse shape.

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- ▶  $2K$  DOF:  $\{a_{m,k}\}_{k=0}^{K-1}$  and  $\{\tau_{m,k}\}_{k=0}^{K-1}$ .
- ▶ *Finite rate of innovation* (FRI) sampling and recovery.<sup>2</sup>
- ▶ At least  $N \geq 2K + 1$  samples.

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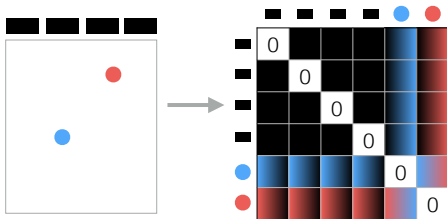


**And Now For Something  
Completely Different**



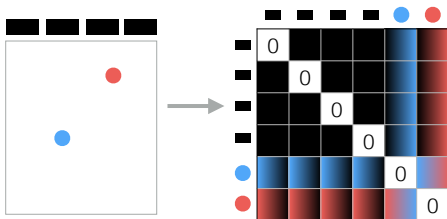
# An overview of *Euclidean distance matrices* (EDMs)

- ▶ Consider  $P$  points (elements and reflectors)  $\{\mathbf{x}_p\}_{p=0}^{P-1}$  in a  $D$ -dimensional Euclidean space.
- ▶ For US,  $P = M + K$  and  $D = 2$ .



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- ▶ For US,  $P = M + K$  and  $D = 2$ .



- ▶ Entry at  $i$ -th row and  $j$ -th column of an EDM  $\mathbf{E} \in \mathbb{R}^{(P \times P)}$ :

$$\mathbf{E}_{(i,j)} = \|\mathbf{x}_i - \mathbf{x}_j\|^2 = \mathbf{x}_i^T \mathbf{x}_i - 2\mathbf{x}_i^T \mathbf{x}_j + \mathbf{x}_j^T \mathbf{x}_j.$$

## An overview of EDMs (cont.)

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- ▶ Matrix formulation with  $\mathbf{X} = [\mathbf{x}_0, \dots, \mathbf{x}_{P-1}]$ :

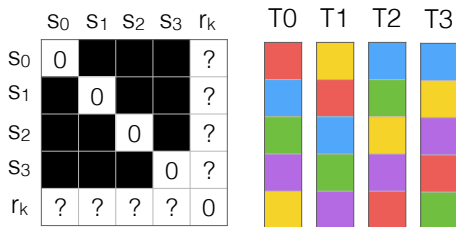
$$\mathbf{E} = \mathbf{1}\text{diag}(\mathbf{X}^T\mathbf{X})^T - 2\mathbf{X}^T\mathbf{X} + \text{diag}(\mathbf{X}^T\mathbf{X})\mathbf{1}^T.$$

# An overview of EDMs (cont.)

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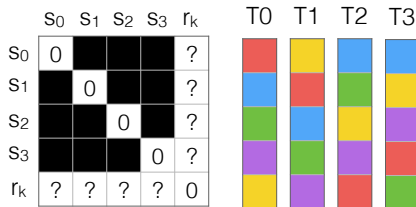
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- ▶ For  $P \geq D \Rightarrow \text{rank}(\mathbf{X}^T \mathbf{X}) \leq D \Rightarrow \text{rank}(\mathbf{E}) \leq D + 2$ .
- ▶ Assuming perfect TOF recovery, we have entries of EDM but need to determine their position.



# TOF matching

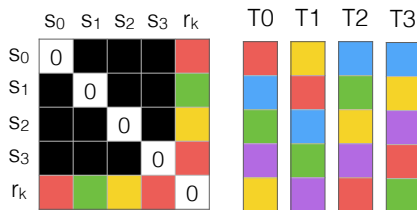
1. With  $D + 2$  elements form an EDM with maximum rank.
2. Augment EDM with different combos of recovered TOFs.<sup>3</sup>
  - ▶ Estimate / remove transmit time and multiply with  $c$ .
  - ▶ Incorrect combos will increase rank.
  - ▶ Correct ones will not!



<sup>3</sup>I. Dokmanić, R. Parhizkar, J. Ranieri, and M. Vetterli, "Euclidean distance matrices: essential theory, algorithms, and applications," 2015.

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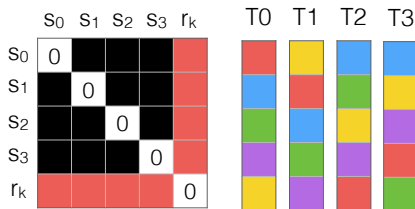
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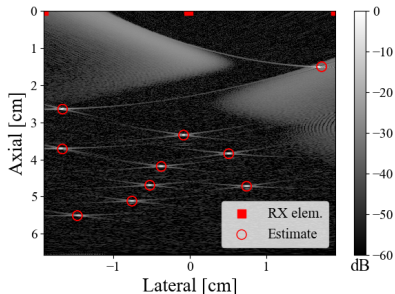
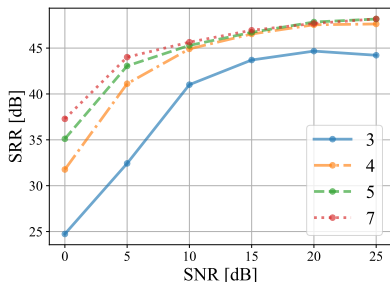
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$K$  TOFs,  $M$  channels  $\Rightarrow K^M$  combinations.

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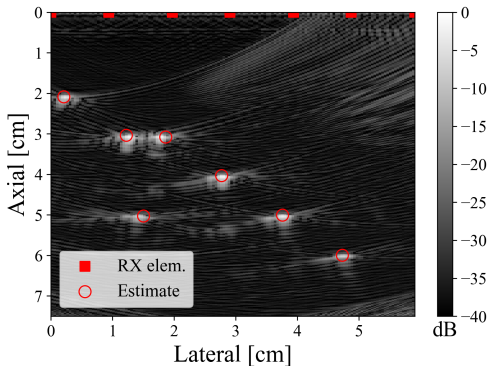
# Simulation with Field II



- ▶ 50 unique configurations of 10 reflectors at varying SNR.
- ▶ 128 transmit elements, single-cycle square wave excitation, and Gaussian-modulated sinusoidal impulse response ( $f_c = 5.208$  MHz and  $bw = 67\%$ ).



# In-vitro non-destructive evaluation scenario



- ▶ Aluminum block with drilled holes.
- ▶ 64 transmit elements,  $f_c = 5$  MHz.

- ▶ Exploit minimum DOF for localization.
  - ▶ FRI along each element  $\Rightarrow N \geq 2K + 1$  samples.
  - ▶ EDM across elements  $\Rightarrow M \geq 3$  elements (using Gram matrix).

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<sup>4</sup>More detail in: E. Bezzam, "Sampling at the rate of innovation of ultrasound imaging and localization," 2018.

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- ▶ In the paper:<sup>4</sup>
  - ▶ More on TOF recovery.
  - ▶ Dealing with noise for TOF recovery and matching.
  - ▶ Easing the combinatorial matching problem.

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# THANK YOU FOR YOUR ATTENTION!

Eric Bezzam

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