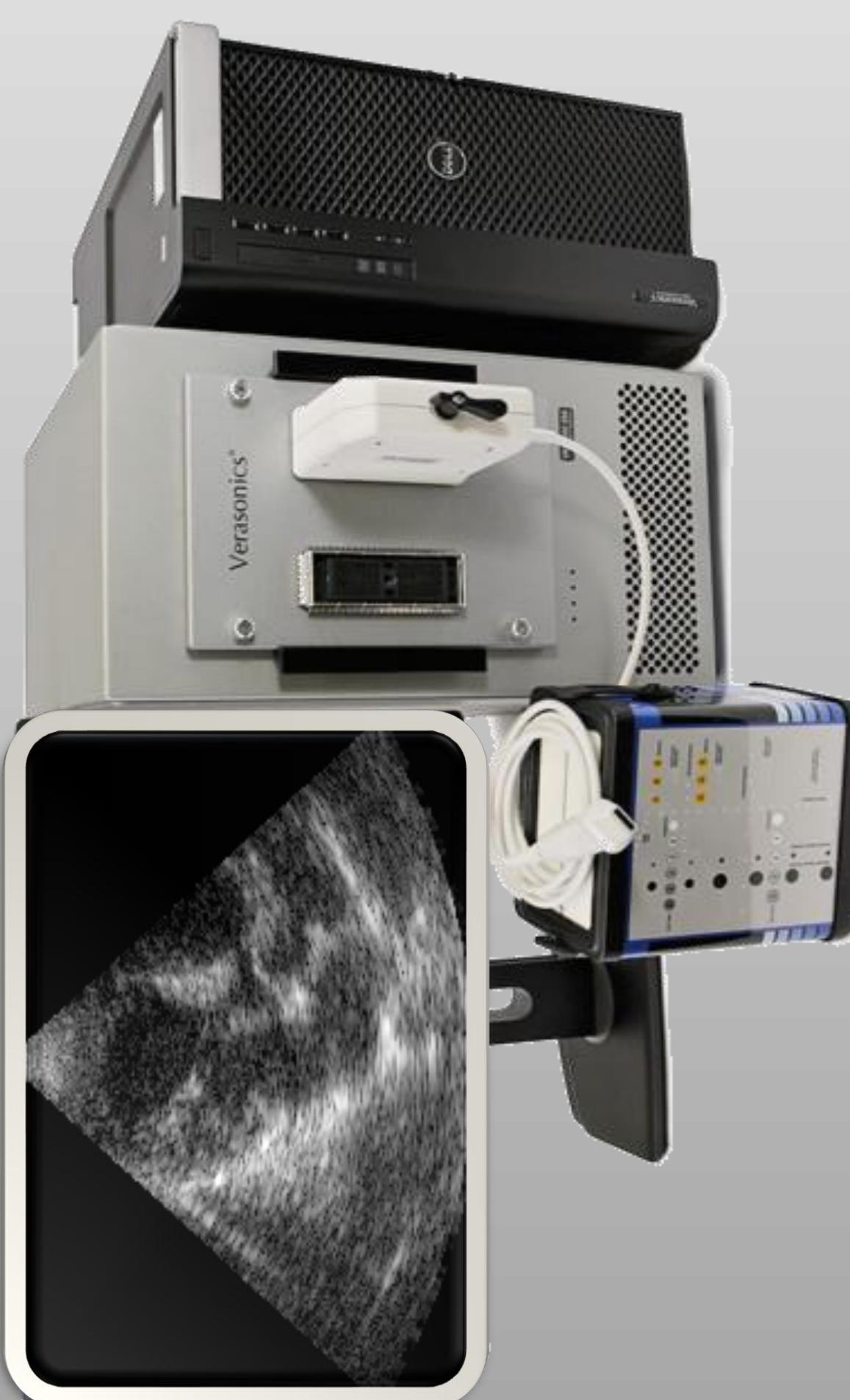


You are welcome to join us and  
take part of our live demonstrations

**Place:** Verasonics booth

**Time:**

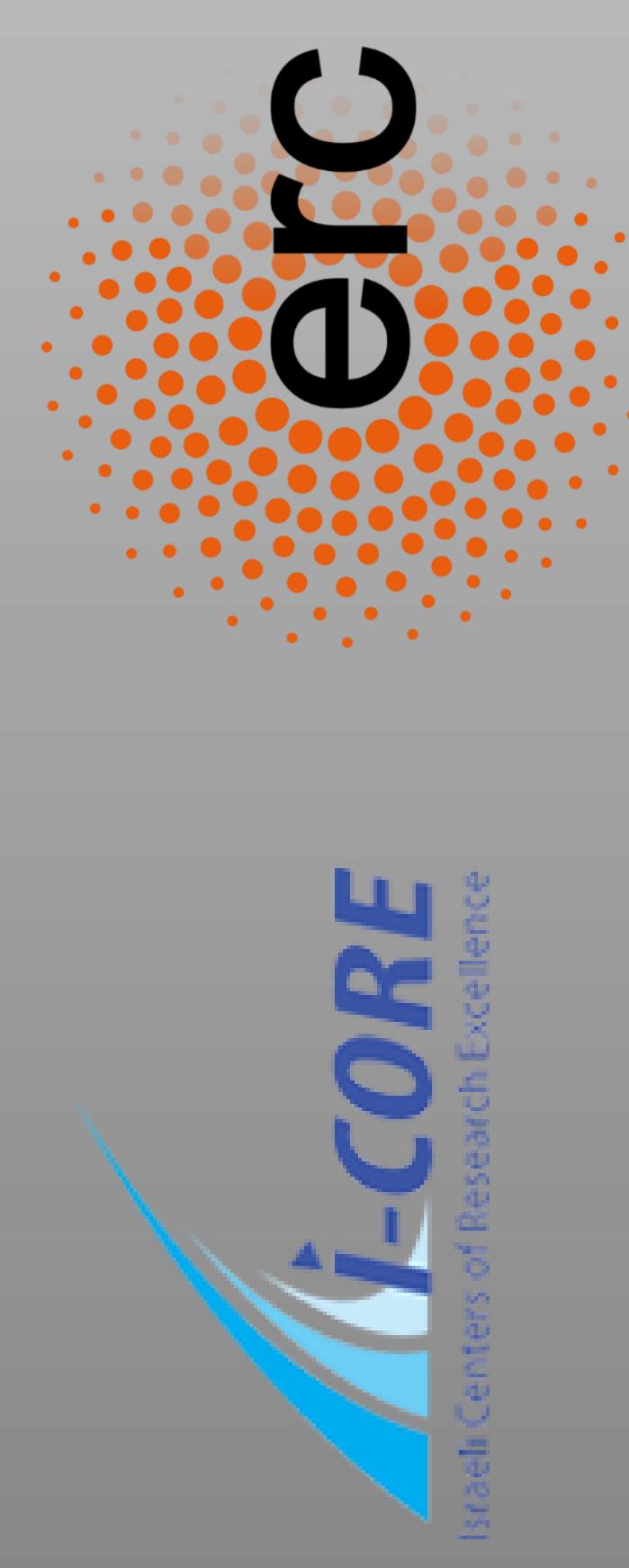
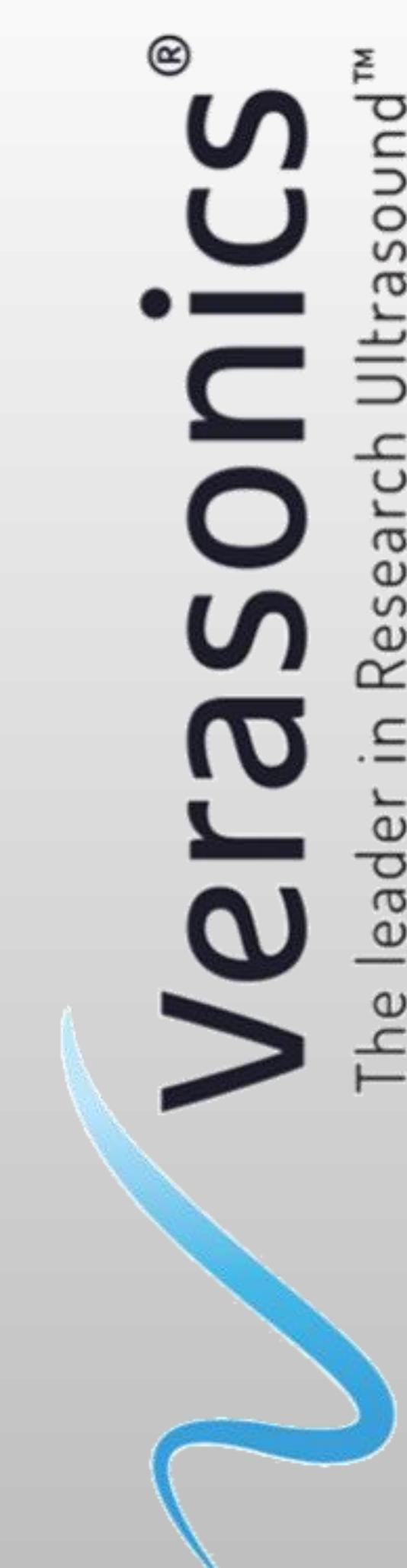
- Tuesday 15:30-16:00
- Wednesday 15:30-16:00
- Thursday 15:30-16:00



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SAMPL – Headed by Yonina C. Eldar  
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# Sparse Doppler Sensing

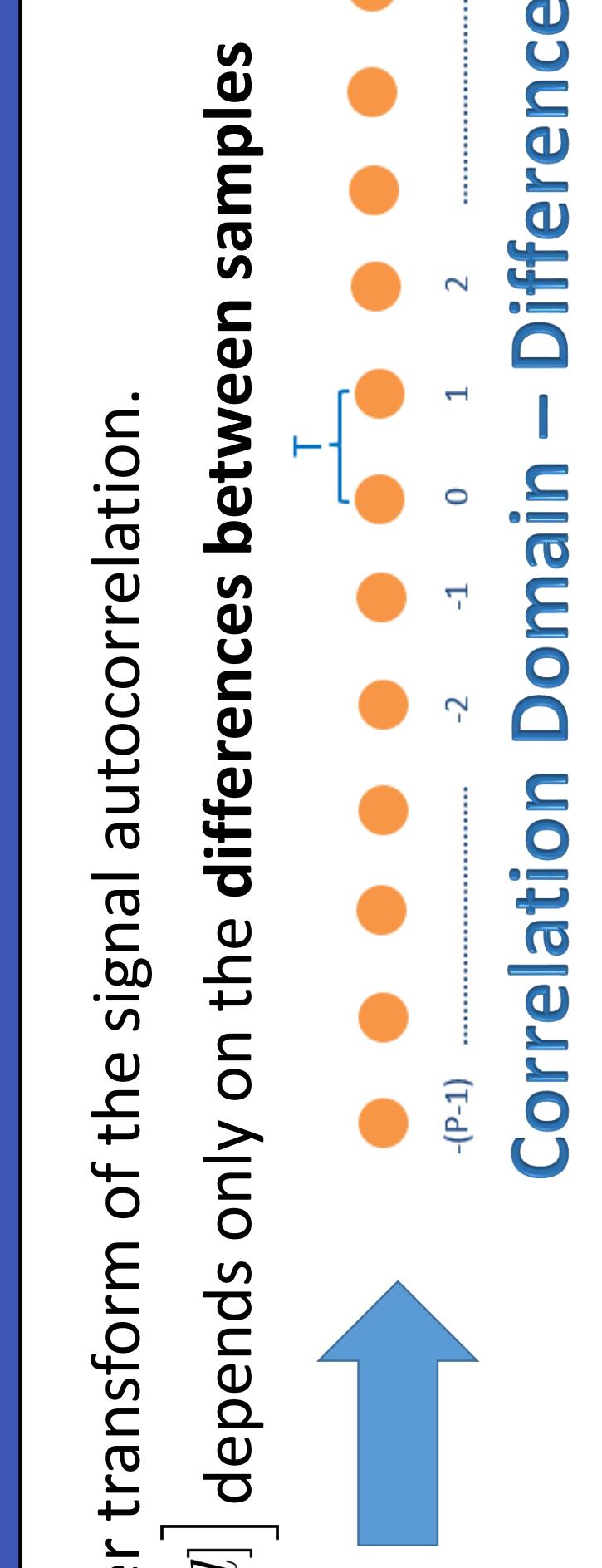
## Challenges in Spectral Doppler

- Spectral Resolution** – Large number of Doppler transmissions is required.
- Alternating Strategy** – Doppler and B-mode both must be displayed at the same time.
- Frame Rate** – We need to identify rapid temporal variations in the blood flow and track tissue movement.
- Lateral Velocity** – flow perpendicular to the beam is not usually measured.
- Spatial Coverage** – In focused acquisition, velocity estimation can be performed only on points on the acquisition line.
- Clutter Removal** – Reflections from the vessel walls degrade our estimation

## Main Goal

Recovering the blood spectrum while reducing the number of transmissions.

## Sparse Doppler Transmission Strategy based on the Difference Array

- The blood spectrum is given by the Fourier transform of the signal autocorrelation.
  - The autocorrelation  $R[d] = E[y[p]\bar{y}[p-d]]$  depends only on the differences between samples
- 
- Correlation Domain – Difference Array**
- Time Domain**

- We introduce a sparse transmission scheme whose **difference array is full**

$$U = [U_A \ U_B] \rightarrow U_A = \{1, \dots, A\}, \quad U_B = \{n(A+1) : n = 1, \dots, B\}.$$

- This allows to recover the autocorrelation while reducing the number of transmission to  $A + B$ :

$$\min_{A,B} A + B \quad s.t. \quad P = (A+1)B \rightarrow A = \sqrt{P} - 1, B = \sqrt{P}.$$

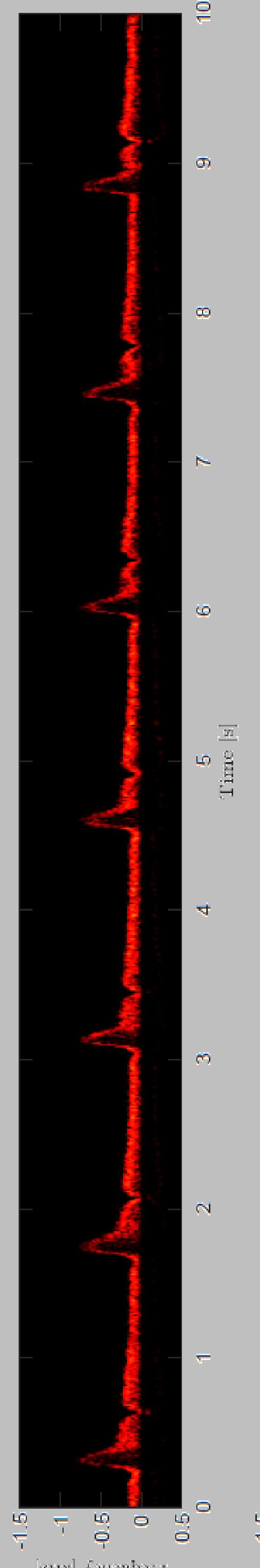
- The Doppler gaps created can be used for B-mode or for other Doppler sequences at different directions.

- For more details:**

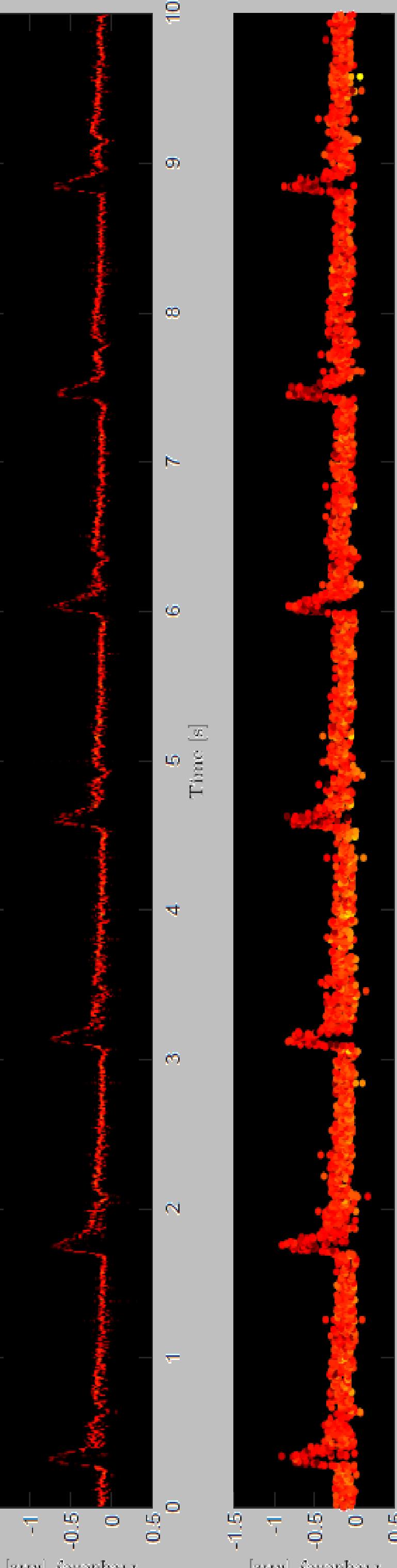
Cohen, R. and Eldar, Y.C. “**Sparse Convolutional Beamforming for Ultrasound Imaging**”. To appear in IEEE Transactions on Ultrasonics, Ferroelectrics, and Frequency Control.

## In-vivo Results – Axial Flow of Carotid Artery

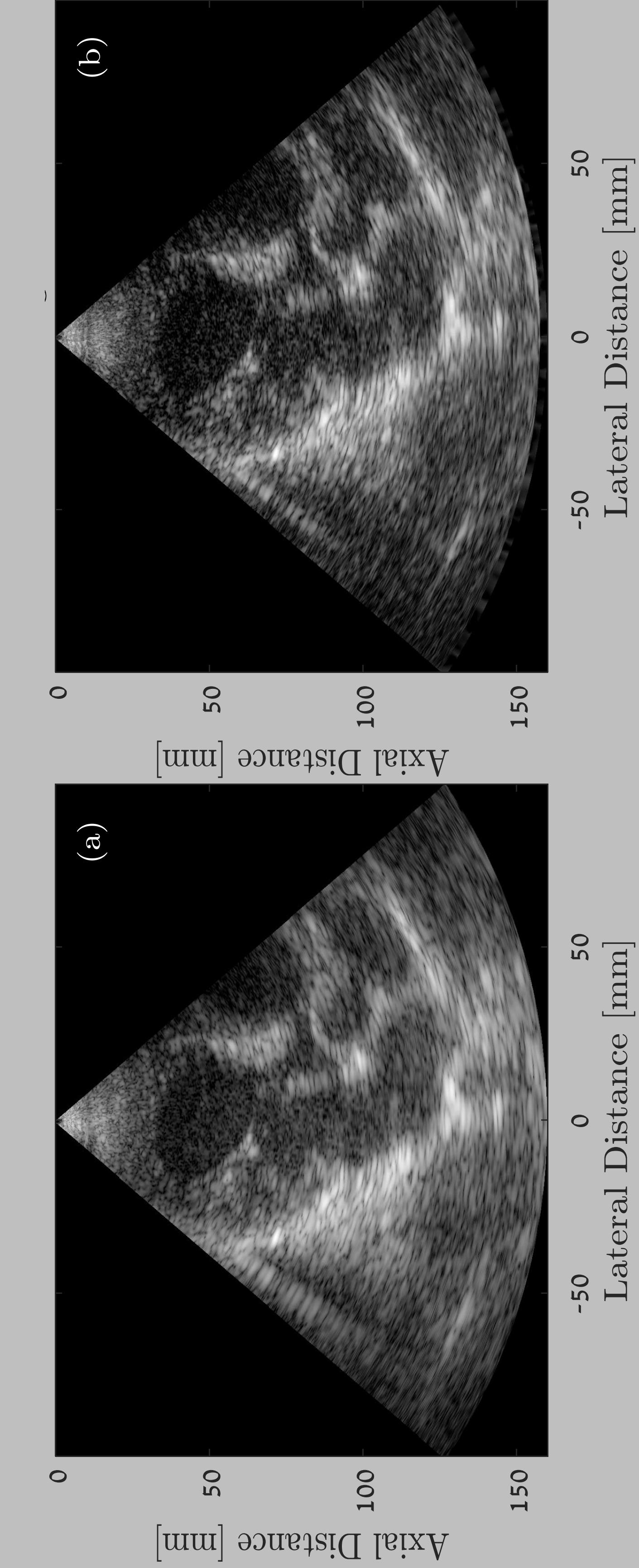
### Welch – 128 Pulses



### NEST – 35 Pulses



### DAS – 64 Elements



## In-vivo Cardiac Data

### SCOBA – 16 Elements



# Sparse Convolutional Beamforming

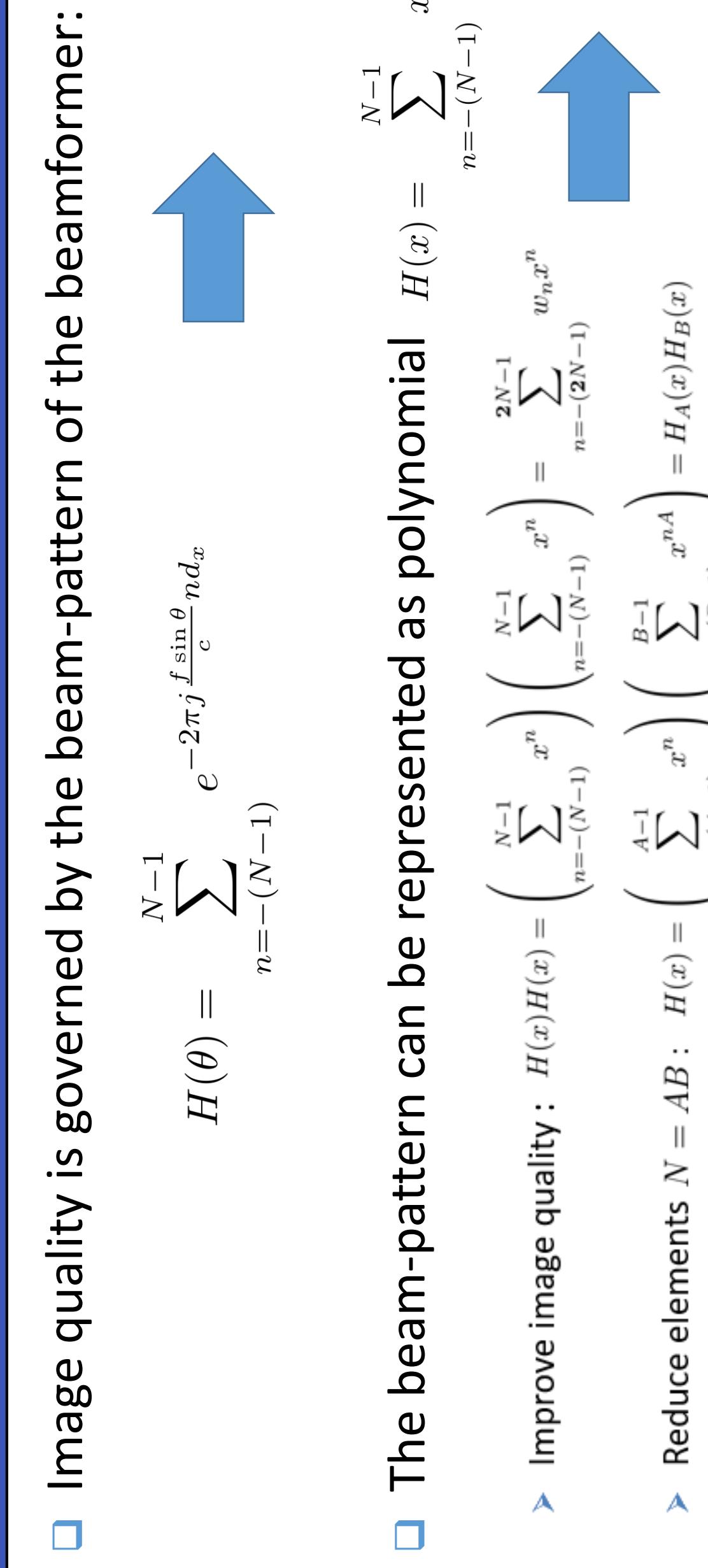
## Main Limitations of Standard Delay and Sum (DAS) Beamforming

- Low Image Resolution.
- Low Image Contrast.
- Large Number of Elements (Receive Electronics) – increase in size, power and cost.

## Main Goal

Reducing the number of receive elements while improving image quality.

## Sparse Convolutional Beamforming based on Cantor Arrays

- Image quality is governed by the beam-pattern of the beamformer:
- 
- The beam-pattern can be represented as polynomial  $H(x) = \sum_{n=-N-1}^{N-1} e^{-2\pi f \sin \theta / c n d_x} x^n$
- Product of polynomials = discrete convolution
- Convolution of received signals!**

- The beam-pattern can be represented as polynomial  $H(x) = \left( \sum_{n=-(N-1)}^{N-1} x^n \right) \left( \sum_{n=-(N-1)}^{N-1} x^n \right) = \sum_{n=-(2N-1)}^{2N-1} w_n x^n$
- Improve image quality:  $H(x)H(\bar{x}) = \left( \sum_{n=-(N-1)}^{N-1} x^n \right)^2 = \sum_{n=-(2N-1)}^{2N-1} w_n^2 x^n$
- Reduce elements  $N = AB : \quad H(x) = \left( \sum_{n=-(A-1)}^{A-1} x^n \right) \left( \sum_{n=-(B-1)}^{B-1} x^{nA} \right) = H_A(x)H_B(x)$
- We propose a non-linear beamformer which is based on the convolution of received signals. This allows
  - Significant element reduction.
  - Improved image quality – enhanced contrast and twice the standard resolution.
- For more details:

Cohen, R. and Eldar, Y.C. “**Sparse Convolutional Beamforming for Ultrasound Imaging**”. To appear in IEEE Transactions on Ultrasonics, Ferroelectrics, and Frequency Control.